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# Safety properties and memory bound analysis in a functional language without a garbage collector

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## PhD Thesis

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# Abstract

In most functional languages, memory management is delegated to the runtime system. There usually exists a *garbage collector* in charge of determining, at runtime, which parts of the memory are no longer needed and can be disposed of. The main advantage of this approach is that programmers are not bothered about memory management issues. However, this purely implicit approach makes it difficult to predict, at compile time, the lifetimes of the data structures at runtime. Other languages delegate the task of memory management to the programmer. In this way, the lifetimes of data structures are known at compile time. However, this is an error-prone approach, since the programmer might try to access to memory areas which have been disposed of previously (*dangling pointers*), mainly due to unexpected sharing between data structures.

*Safe* is a first-order functional language that provides a semiexplicit approach to memory management by means of two mechanisms: regions (inferred by the compiler), and destructive pattern matching (controlled by the programmer). In this way, we can do without a garbage collector to destroy the data structures that are no longer used. As a consequence, the behaviour of programs (regarding memory usage) becomes more predictable, and eases the task of developing a static analysis to infer upper bounds on the memory needs of a program.

The incorporation of explicit mechanisms for memory destruction implies the risk of accessing dangling pointers. The first goal of this thesis is to develop a static analysis for guaranteeing the safety of a program with regard to its memory accesses (specifically, the absence of dangling pointers). Once the compiler has checked the presence of this property for a given program, our next goal is to formalize this property as a certificate, so the latter can be automatically checked by the code receiver. Finally, we will develop an abstract interpretation-based analysis for inferring upper bounds to the memory needs of a program. Since the final aim of the *Safe* project is the certification of pointer-safety properties and bounded memory execution, we have put emphasis on the formal correctness of each analysis.



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# Chapter 1

## Introduction

This thesis describes the automatic analysis of pointer-safety properties and memory bounds for a first-order functional language called *Safe*. This language has been developed in the last few years as a research platform for analysing and formally certifying properties of programs, with regard to memory usage. It was introduced for investigating the suitability of functional languages for programming small devices and embedded systems with strict memory requirements. In this chapter we will motivate the existence of *Safe*, and specify the goals of this thesis.

### 1.1 Motivation

Computer systems are not fully reliable. They just do what a programmer tells them to do, and programmers make mistakes. As software plays an increasingly relevant role on computer systems, its reliability becomes a major concern, especially in those cases when economic damages or human losses might arise as a consequence of a malfunction. The broad research area of *formal methods* deals with the development of mathematically rigorous techniques for the specification of desirable properties of programs, and their verification. Among the set of desirable properties of a program, the most decisive ones are those related with its correctness, which ensures that a program does what the programmer expects it to do. This class of properties is commonly known as *functional properties*.

Without regard to the correctness of the results of a program, there are some other desirable properties that are relevant to the safety of software systems. These are called *non-functional properties*. An example is the fact that a program performs its task in a given amount of time, or that its memory needs do not exceed a given limit. These two examples become part of a broader research field, whose name is *resource analysis*. In this framework, a program is conceived as a resource consumer (resource may be understood as time, memory, energy, etc.) and the aim is to compute an upper bound to the resources being consumed by every possible execution of the program. In this work, we are particularly interested in the analysis of *memory bounds*. Memory consumption is specially relevant to several scenarios: for instance, when programming embedded devices, it is necessary to make sure that the programs running in these devices do not stop working because they try to use more memory than it is available. It is also useful to know in advance how much memory would be needed by the program, in order to reduce hardware and energy costs.

We believe that the choice of the programming paradigm plays a special role in the reliability of software systems: the friendliest languages (from the programmer's point of view) are those that include concepts from the problem to be solved, rather than those of the machine in which that problem

is solved. *Declarative paradigms* fall into this category, since they emphasize the specification of the computations, instead of how these computations are carried out. *Functional programming* belongs to this class of paradigms. Functional programs are based on the mathematical concept of a function, defined as a set of equations. This paradigm abstracts the way in which these functions are evaluated. The main advantage of pure functional programming is the absence of side effects, from which the property of *referential transparency* follows: a function is defined exclusively by a correspondence between its input and its output (in other words, the result of a function solely depends on its input parameters). This implies that, in a pure functional setting, an expression is always evaluated to the same value, without regard to the execution context where the evaluation takes place.

The absence of the notion of a state makes the task of reasoning about the functional properties of a program easier. As a consequence, functional languages are, in general, better suited to several analyses, because of this lack of state. However, the inference of memory bounds requires special attention. In most functional programming languages, memory management is delegated to the runtime system, which allocates memory as it is needed by the program, provided there is enough space available. A *garbage collector* is in charge of determining, at runtime, which parts of the memory are no longer needed, and can be safely disposed of. The main advantage of this approach is that programmers do not bother about low-level details on memory management, but there are also some drawbacks. On the one hand, the time delay introduced by garbage collection may prevent a program from providing an answer in a required reaction time, which may be unacceptable in the context of real-time systems. On the other hand, garbage collection makes it difficult to predict at compile time the lifetimes of data structures, specially in those cases where the runtime system does not specify under which conditions a garbage collection takes place. This does not necessarily prevent the development of a static analysis for inferring memory bounds: a liveness analysis can be used to predict the part of the heap that may still be accessed by a given program, whereas the total amount of allocated heap is not relevant, as dead memory will be eventually reclaimed by a garbage collector. However, it is difficult to know the actual allocated memory *at a given execution time*. This is specially inconvenient when several programs are executed in the same device: if one of them runs out of memory, the remaining ones might not activate their garbage collectors unless they also need more memory. In order to compute memory bounds for functional programs, it is desirable to set out a memory management approach different from garbage collection. In the next sections we shall explore some possibilities.

### 1.1.1 Region-based memory management

In region-based languages, each allocated object is created in a region. Regions are created and disposed of in a stack-like fashion: the last region being created is the first one being destroyed. Allocation and deallocation of regions take place in constant time. The compiler has to determine: (1) when regions are created and destroyed, and (2) in which region objects must be created at runtime. The main improvement of region-based memory management on garbage collection approaches is that they bring forward to compile-time the decision on when a given data structure will be created or destroyed.

Earlier works in this memory management technique were targeted towards imperative languages. These works include that of Ruggieri and Murtagh [102], in which each region is associated with the activation record of a procedure. Each region is created at the beginning of a function call and it is removed when this function call finishes. This mechanism is improved in the seminal work of Tofte and Talpin [115, 116], which provides a type-based approach to region inference in a functional language (ML), including higher-order functions, and region-polymorphic recursion. This work have been im-

plemented in the MLKit Compiler [113]. Regions have also been applied to object-oriented languages [28, 27]. Moreover, the Real Time Specification for Java [22] includes a model for providing memory areas that are managed using regions, and are not subject to garbage collection.

In absence of garbage collection, region-based memory management is well-suited to the automatic inference of memory bounds, since the allocation/deallocation of regions is determined at compile-time, as well as the regions in which the different data structures are located. Regions are also time efficient [17], since both allocation and deallocation take place in constant time. However, a major disadvantage of regions is their strict nested discipline. Regions are created and destroyed in a LIFO basis, which does not always correspond to the lifetimes of data structures. As described by Berger, Zorn, and McKinley [17]:

Regions provide high-performance but force the programmer to retain all memory associated with a region until the last object in the region dies [...]. We show that the performance gains of regions (up to 44%) can come at the expense of excessive memory retention (up to 230%). More importantly, the inability to free individual objects within regions greatly complicates the programming of server applications like Apache which rely on regions to avoid resource leaks.

There exist many approaches for breaking this strict nested discipline of regions: region resetting [19], dissociating region allocation/deallocation from its definition [2], annotating the program with primitives for allocating/deallocating regions [54, 21], and linear regions [45], among others. These approaches require, to a greater or lesser extent, a deep knowledge about how regions are inferred by the compiler, and some of them require help from the programmer in the form of annotations. Alternatively, regions can be combined with garbage collection in order to reduce the impact of memory leaks [114].

### 1.1.2 Explicit cell deallocation

Another alternative to garbage collectors, in contrast to the fully automatic approach of region-based memory management, is to delegate the task of deallocating memory to the programmer. This approach can be mainly found in some imperative languages (being C and C++ well-known examples of this), but it is very rare in declarative languages. In presence of explicit deallocation, the compiler always knows when data structures are allocated or deallocated at runtime, since this information is contained within the source code. This makes the task of analysing memory bounds easier than in presence of garbage collection.

However, explicit deallocation is an error-prone approach. The execution of a program may result in abortion if it tries to access to those parts of memory which have been previously disposed of. These parts of memory are commonly known as *dangling pointers*. These kind of accesses to dead memory may arise as a consequence of unexpected sharing between the different data structures. A program is said to be *pointer-safe* if dead memory is never accessed during its execution. Moreover, explicit deallocation-based approaches are not exempt from memory leaks (i.e. unused memory that is not released by the program), since the programmer may forget to release the data structures involved in the program.

There exist several type-based techniques for avoiding dangling pointers in a functional setting. In [57] Hofmann describes LFPL, a linearly-typed [118] functional language that allows the in-place modification of heap-allocated data structures. In this language, the positions of the heap are made explicit in the program. This approach is refined in [10, 11], by making a distinction between three

different usage aspects. This scheme, while still being safe, allows more pointer-safe programs than its predecessor.

In the context of imperative languages, *separation logic* [100] provides an extension of Hoare logic, which allows specifying properties about programs which manage shared mutable data structures. These ideas have been applied to Space Invader [24], which uses abstract interpretation-based techniques for proving assertions on separation logic. A combination of separation logic with the work on LFPL is developed by Konečný in [69, 67], which distinguishes different layers in the data structures, and infers separation assertions between layers.

All these systems can prove the absence of dangling pointers, but they cannot prove their presence. This means that there may be pointer-safe programs being rejected by the compiler, since the decision problem of determining whether a pointer is dangling or not is undecidable. Current analyses can only compute an approximation of this property. In case a pointer-safe program is (wrongly) considered unsafe by a static analysis, the corresponding implementation may proceed in two ways: either it warns the programmer about a potential dangling pointer, but allows the execution of the program, or it aborts the generation of the object code, so ill-typed programs cannot be executed. In the first case, the pointer-safety property of the program is no longer guaranteed. The second possibility, which makes more sense in the context of safety-critical systems, the programmer is forced to modify his/her program in order to remove those deallocations that make the program be rejected by the analysis. This may result in memory leaks. Nevertheless, and as in the case of region-based memory management, explicit destruction can be combined with garbage collection-related approaches.

### 1.1.3 *Safe's* approach to memory management

In those scenarios where pointer-safety is a must, both of the mechanisms shown in previous sections are able to infer an approximation of this property. This approximation is done in such a way that pointer safety is guaranteed, although some pointer-safe programs are rejected. Over-approximation is inherent to the inference of undecidable properties. In the context of pointer-safety, over-approximation in data structure lifetimes may lead to memory leaks.

*Safe's* memory management is a combination of region-based memory management and explicit deallocation via destructive pattern matching (in the style of Hofmann and Jost's LF language [58]). Pointer-safety is guaranteed by means of a type system. The problem of over-approximating this property is still insurmountable, but the combination of these mechanisms can alleviate its consequences. Explicit deallocation provides a way to selectively destroy data structures inside a region, without having to wait the whole region to be deallocated. In those cases where the compiler forbids the explicit deallocation of some data structure, region-based memory management guarantees that the memory occupied by this data structure will be eventually reclaimed when the deallocation of the corresponding region takes place. As a consequence, each approach provides an escape mechanism to the limitations of the other.

Given the above, we say that *Safe* provides a *semi-explicit approach* to memory management, since it features implicit (i.e. automatically inferred) regions, and explicit deallocation. A drawback of this approach is that the programmer needs some understanding on how regions are created and destroyed, in order to be able to judge the need for explicit deallocation in his/her program. In *Safe* this is alleviated by defining a model of region-based memory management which is simpler than the original system of Tofte and Talpin [116]. There exists a single region per function call, so the programmer only has to think whether a given data structure is local to that function call.

A major advantage of combining regions with explicit deallocation is that a garbage collector is not needed. As a consequence, the lifetime of data structures is completely determined at compile-time<sup>1</sup>, so *Safe* programs are amenable to analyses on memory consumption.

## 1.2 Certification of safety properties

Certifying program properties consists in providing mathematical evidence about them. In a *Proof Carrying Code* (PCC) environment, these proofs should be mechanically checked by an appropriate tool. The research line of PCC started with the seminal work of Necula [88]. The PCC paradigm aims to create an infrastructure in which programs are associated with formal proofs guaranteeing the presence of some desirable properties. These formal proofs make up a *certificate*. The code producer generates both the compiled program and its certificate, and the code consumer has to check that the certificate is consistent with the program being supplied, and that the proofs contained within the certificate are correct. The certificate checking is done in an automatic way, and usually with the help of a proof assistant, such as Isabelle/HOL [93].

From the code producer's point-of-view, there are several approaches to the certificate generation. In the standard PCC approach, the properties are determined and certified at the object-code level. In those cases where the properties being certified are obtained as a result of a static analysis, the latter would have to be carried out at the object-code level. Performing static analysis on object code usually involves a loss of accuracy in the property being determined, since part of the information of the source code is lost when the latter is translated to lower-level code. Another possibility involves determining the certified properties at the source code level, and then generating the certificate at that level. Then, both the code and the certificate are translated into their low-level counterparts, and sent to the code consumer. The main drawback of having low-level certificates is that it requires a huge effort for formalizing the invariants and properties of the elements that take part in the runtime system (stack frames, program counters, a.o.), in contrast to the high-level certificates, whose properties are proved with respect to the higher-level language semantics, much simpler. Moreover, low-level certificates are of considerable size, since they include the proofs of these additional properties and invariants, and this results in longer checking times.

In *Safe*, certificates are generated at the source-code level, and they are sent, together with the source code, to the code consumer. The latter checks the validity of the certificate, and then performs the translation of the source code into its lower-level counterpart. This translation is certified [39, 38], so it guarantees the preservation of the properties being certified, when translating from higher-level code to low-level code. A notable advantage of this approach is that the translation has to be certified only once by the code consumer. Once done, the certified compiler is valid for every program. This has the additional advantage that certificates are of shorter size, since they only include the program-specific proofs. This comes at the cost of supplying the source code to the consumer, which might not be an option in some PCC scenarios.

Among the first publications devoted to certifying compilers we have that of [89, 91, 30], in which the authors develop a certifying compiler for a subset of Java bytecode. The *MRG (Mobile Resource Guarantees)* project [103] aims at the generation of mobile code with a certificate proving bounded heap memory consumption. Certificates are generated at the level of a Java bytecode-alike language called *Grail* [9]. More recently, the *Mobius (Mobility, Ubiquity, and Security)* project [14] aims to extend the PCC

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<sup>1</sup>The lifetime of a data structure is determined by both the programmer (when using explicit deallocation), and the compiler (when assigning regions to data structures).

paradigm to more general scenarios, such as distributed and concurrent systems, resource analysis, security attacks, etc. It is targeted toward mobile devices satisfying the MIDP (*Mobile Information Device Profile*) specification of Java ME.

### 1.3 Memory consumption analysis

Although relatively new, the field of resource analysis has gained considerable attention in the last years, mainly due to the application of mathematical techniques (such as linear programming, or recurrence solving) to programming languages. In particular, the inference of memory bounds is a very complex task that involves several auxiliary analyses, each one a challenge by itself. The first results on memory consumption analysis were targeted towards the functional programming paradigm. The techniques developed were subsequently adapted to mainstream languages, such as Java or C++.

Hughes and Pareto introduce in [61] a first-order functional language with a type and effect system guaranteeing termination and execution in bounded space. This system is a combination of Tofte and Talpin’s approach to regions with sized types [62, 96].

The first fully automatic way to infer closed-form memory bounds is due to Hofmann and Jost [58]. Their analysis, based on a type system with resource annotations, can infer linear heap memory bounds on first-order functional programs with explicit deallocation. These techniques have been applied to subsets of imperative languages, such as Java [59] (without type inference), and C [48] (for computing worst-case execution time). The annotated type system also serves as a basis for a stack consumption analysis due to Campbell [25, 26]. Hofmann and Jost’s approach is extended in [64, 63] to higher-order programs. The latter work provides a general framework that can accommodate different notions of cost. More recently, Hoffmann and Hofmann have extended the initial work of [58] to polynomial memory bounds [56, 55].

The classical approach to resource analysis, due to Wegbreit [119] involves the generation of a recurrence relation from the program being analysed, and, in a second phase, the computation of a closed-form expression (without recursion) equivalent to that recurrence relation. Vasconcelos and Hammond pursue this approach in [117], which is fully automatic in the generation of recurrence equations, but requires the use of an external solver for obtaining a closed form. The COSTA System [5] follows a similar approach, but it provides its own recurrence relation solver, PUBS [4], which can handle multivariate, non-deterministic recurrence relations. COSTA is an abstract interpretation-based analyser which works at the level of Java bytecode, and supports several notions of cost, such as the number of bytecode instructions executed, heap consumption, and number of calls to a particular method. Since memory management in Java is based on garbage collection, their approach to memory consumption is parametric on the behaviour of the garbage collector [7]. The bounds computed by this system go beyond linear expressions; it can compute polynomial, logarithmic, and exponential bounds.

Although not directly aimed at memory bounds, the SPEED project [49] computes non-linear symbolic bounds by combining techniques such as loop transformations, counter instrumentation, and invariant generation.

The problem of inferring memory bounds is closely related to the inference of *size relations* between data structures. The seminal work on type-based sizes is the above mentioned of Hughes, Pareto, and Sabry [62], which is restricted to type checking. The problem of inference is addressed by Chin and Khoo [29]. Another approach involves using abstract interpretation-based techniques on the domain of convex polyhedra, as done by Benoy and King [16]. Abstract interpretation is also applied in [108] to

the approximation of the height<sup>2</sup> of data structures. All the techniques described so far are restricted to infer linear size relations. The work of Shkaravska, van Eekelen and van Kesteren [105] is able to infer polynomial size relations. It provides a type system, in which checking is decidable under certain syntactic conditions. Type inference is performed by a combination of testing and polynomial interpolation-based techniques. More recently, these techniques have been applied to the computation of polynomial loop-bound functions for Java programs [106].

## 1.4 Goals and structure of the work

When considered as a whole, the *Safe* project aims to study the appropriateness of functional languages to devices with strict memory requirements. The goals of this thesis are intended to get closer to this general objective. These goals are the following:

### 1. Develop an efficient, fully automatic, type-based, static analysis for ensuring pointer safety.

Given the memory model of *Safe*, informally explained in Section 1.1.3, the analysis must accommodate region-based memory management and explicit destruction. This analysis will be based on a type system, which will be proven correct with respect to the operational semantics of the language.

With regard to type inference, we will consider regions and explicit destruction separately:

- (a) A region inference algorithm will annotate the source program with region variables, in such a way that the resulting program is well-typed w.r.t. the above mentioned type system. This annotation process will be done in an optimal way, in the sense that every data structure will be placed in the region with the shortest lifetime that does not compromise the pointer safety of this data structure. The algorithm will support polymorphic-recursive functions, and will be proved correct with respect to the type system.
- (b) A mark inference algorithm will ensure that the explicit destruction specified by the programmer does not generate dangling pointers at runtime. In this case, the sharing relations between the different data structures play a fundamental role. The approximation of these relations given by an already existing sharing analysis will serve as a basis for the inference of safe destruction. The inference algorithm will also be proven correct with respect to the type system.

### 2. Certify pointer-safety in an automatic way.

Assuming that a *Safe* program is well-typed with regard to the type system described above, we will describe how the compiler generates a certificate that can be mechanically checked by the Isabelle proof assistant. This certificate proves the pointer-safety property of the program being compiled. The process of certificate generation can be split into two tasks:

- (a) The first task involves proving that well-typed programs do not access dangling pointers. This task involves a formalization of the type system's correctness in a proof assistant. The corresponding theorems are generic, in the sense that they are proved once for all, and reused in every certificate.

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<sup>2</sup>The height of a data structure is the longest chain of pointers that can be followed from a given one.

- (b) The second task involves proving that the specific program being certified is well-typed. By combining this part with the results of the first task, the certificate proves that the program being certified does not access dangling pointers.

In this thesis we shall deal only with the latter point: the generation of the program-specific part of the certificate. The details of the generic part can be found in [37, 35].

### 3. Develop a memory cost model for *Safe* programs.

A memory cost model provides a way to formally describe the memory needs of each syntactical construction of the language. Our cost model will be included in the *Safe* big-step operational semantics, and it will take both heap and stack memory needs into account. This model will be formally derived from the translation of *Safe* source code to the code of the imperative SVM (*Safe* virtual machine), and it will be proven correct with respect to this translation.

### 4. Develop an abstract interpretation-based analysis for inferring heap and stack memory bounds.

Given the cost model of the previous point, we aim to use abstract interpretation-based techniques [33] for inferring non-linear, monotonic, closed-form expressions bounding the heap and stack memory costs of a program. The resulting bounds will be functions on the sizes of the program's input.

Even if we restrict ourselves to a first-order functional language, like *Safe*, the inference of safe memory bounds is a very complex task, which involves considering several preliminary results, such as size analysis, and call-tree size analysis. Each one of these analysis could be, by itself, a subject of a PhD thesis. Therefore, we will focus on the inference of heap and stack memory bounds by assuming that the size and call-tree information is given externally.

We will formally prove the correctness of the results of the analysis with respect to the memory cost model. We will also specify certain conditions on the externally-given call tree information, which allows to prove the *reductivity* of the resulting bounds. The advantage of having reductive bounds is that they will allow us to obtain a decreasing sequence of bounds, as a result of iterating the analysis several times for the input program.

In this work we have put emphasis on the formal correctness of each of the analyses. We believe that, in general, formal proofs play an important role when developing languages devoted to safety-critical systems, but, in particular, they are very relevant to the *Safe* project, since its goals include the formal certification of the properties being proved. In this work we only deal with the certification of pointer-safety properties, but much work has also been done in *Safe* for certifying memory bounds [40, 35].

The structure of the thesis is as follows:

- In Chapter 2 we introduce, through several examples, the basic concepts of the *Safe* language. Then we formalize the operational semantics of the language, which serve as a basis for deriving and abstract virtual machine for *Safe* programs (SVM). The translation of *Safe* into instructions of SVM will allow us to derive a memory cost model, whose correctness will be proved with respect to the translation. With this cost model we will have achieved the third goal shown above. This chapter is an extended version of the work described in [82, 86].
- Chapter 3 presents a type system guaranteeing that regions and explicit destruction are done in a safe way. There we assume that the source program is already annotated with regions, and show that well-typed programs do not access dangling pointers. The problem of inferring correct

typing derivations is deferred to the next two chapters. This chapter is an improved version of the work in [84].

- Chapter 4 addresses the problem of region inference. Given a *Safe* program, the algorithm decorates its abstract syntax tree with region variables, in such a way that the resulting program is well-typed. The correctness and optimality of the algorithm are formally proved. Our algorithm has been published in [85].
- Chapter 5 deals with explicit destruction. In this chapter we develop an algorithm guaranteeing the safety of the explicit deallocations specified by the programmer. It is proved correct and complete with respect to the type system of Chapter 3. This chapter is an extended version of the publication [81].
- The work of Chapters 3, 4, and 5 is aimed at obtaining typing derivations for a *Safe* program (first goal). Once these derivations have been obtained, Chapter 6 explains how to represent them formally as Isabelle/HOL proof scripts, in order to certify pointer-safety (second goal). It corresponds to the publication [36].
- In Chapter 7 we develop an abstract interpretation-based analysis for inferring heap and stack memory bounds for *Safe* programs (fourth goal). Firstly we define the abstract domain, and the abstraction function which maps program expressions to elements of the abstract domain. This function is proved correct with respect to the cost model of Chapter 2. This chapter extends our previous work on space consumption analysis [83].
- Finally, in Chapter 8 we discuss the obtained results, and draft some future research directions.

### 1.4.1 Notation and basic assumptions

We denote by  $\mathbb{N}$  the set of natural numbers (including zero), whereas  $\mathbb{R}^+$  denotes the set of non-negative real numbers.

The notation  $\overline{x_i^n}$  abbreviates the sequence  $x_1 x_2 \dots x_n$ . More generally, if  $e(i)$  is an expression which depends on the  $i$  variable,  $\overline{e(i)^n}$  denotes the sequence  $e(1) e(2) \dots e(n)$ . With this sequence notation, we normally use  $i$  and  $j$  to denote the bound variable that ranges from 1 to  $n$ . When this notation occurs nested,  $i$  stands for the bound variable of the outermost sequence, whereas  $j$  is the bound variable of the innermost one. For instance, the abbreviation **case**  $x$  **of**  $\overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i^n}$  denotes the following expression:

$$\begin{array}{ll} \mathbf{case\ } x \mathbf{ of} & C_1 x_{11} x_{12} \dots x_{1,n_1} \rightarrow e_1 \\ & C_2 x_{21} x_{22} \dots x_{2,n_2} \rightarrow e_2 \\ & \vdots \\ & C_n x_{n1} x_{n2} \dots x_{n,n_n} \rightarrow e_n \end{array}$$

The superscript denoting the number of elements will be left out when it is not relevant (for instance, as in  $\overline{x_i}$ ). The sequence notation can also be applied when defining a set with a finite number of elements. For instance,  $\{\overline{\rho_i^m}\}$  denotes the set  $\{\rho_1, \dots, \rho_m\}$ .

If  $A$  and  $B$  are sets, we use  $A \rightarrow B$  to denote the set of *partial* functions (or *partial mappings*) which map elements from  $A$  to elements of  $B$ . By abuse of terminology, these partial functions will be denoted simply functions, or mappings. If  $f$  belongs to  $A \rightarrow B$ , the domain of  $f$  is denoted by  $\text{dom } f$ . By convention, the application of a function  $f(x)$  implicitly assumes that  $x \in \text{dom } f$ . Our partial functions

may be considered as a set of bindings of the form  $x \mapsto y$ , where  $x \in A$  and  $y \in B$ , provided that no element of  $A$  appears bound to two different values in the same function. This allows us to use the usual notation of sets with mappings. For instance:

- $\Gamma_1 \subseteq \Gamma_2$  denotes that  $\text{dom } \Gamma_1 \subseteq \text{dom } \Gamma_2$ , and that  $\forall x \in \text{dom } \Gamma_1. \Gamma_1(x) = \Gamma_2(x)$ .
- $\overline{[a_i \mapsto b_i]^n}$  abbreviates the function  $[a_1 \mapsto b_1, \dots, a_n \mapsto b_n]$ .
- $[x \mapsto 0 \mid x \in D]$  is the function which maps every element of  $D$  to 0.
- If  $f, g \in A \rightarrow B$  and  $\text{dom } f \cap \text{dom } g = \emptyset$ , the function  $f \uplus g$  has  $(\text{dom } f \cup \text{dom } g)$  as its domain, and is defined as follows:

$$(f \uplus g)(x) = \begin{cases} f(x) & \text{if } x \in \text{dom } f \\ g(x) & \text{if } x \in \text{dom } g \end{cases}$$

- The notation  $\Gamma \setminus x$  denotes the restriction of  $\Gamma$  to the set  $(\text{dom } \Gamma) \setminus \{x\}$ .

In the context of type environments, we use the more familiar notation  $x_i : y_i$  for denoting their bindings, and we use  $+$  instead of  $\uplus$  to denote the disjoint union of two environments.

*Remark.* Deviating from other authors, the notation  $f[x \mapsto y]$  specifies that the binding  $x \mapsto y$  *already belongs to*  $f$ . On the contrary,  $f \uplus [x \mapsto y]$  denotes the function which maps  $x$  to  $y$ , and behaves like  $f$  for the rest of elements in its domain.

## Chapter 2

# Syntax and resource-aware semantics of *Safe*

### 2.1 Introduction

As it was pointed out in the last chapter, a novel feature in *Safe* is the combination of two mechanisms for replacing garbage collection: region-based memory management and destructive pattern matching. In this chapter we will make this idea more precise. We start with an informal description of these mechanisms from a programmer’s point of view. Then we formally define the syntax and semantics of the language, and in a series of successive formal steps we derive an abstract virtual machine for executing *Safe* programs. This allows us to obtain a *memory model*, which will be useful when inferring upper-bounds to memory consumption (Chapter 7). This chapter provides the theoretical foundations which will serve as a basis for the correctness results of next chapters.

This chapter is an extended version of the work described in [82]. It also includes some implementation aspects of the *Safe* runtime system and its compiler. These were taken from [80]. Firstly we present in Section 2.2 a high-level view of the language *Safe*, and how its memory facilities are implemented. After this, a desugared variant of *Safe* is formally defined via its syntax (Section 2.3) and semantics (Section 2.4). In Section 2.5 we present the abstract machine in which *Safe* programs are run and the translation process between *Safe* programs and the code being executed by the abstract machine (Section 2.6). The abstract machine will serve as a reference implementation which will allow us to determine the memory consumption of a *Safe* program. This will be made apparent by enriching our semantics with a resource vector specifying how much memory is needed by a program in order to be executed (Section 2.7). Finally, Section 2.9 gives a broad overview of the *Safe* compiler’s implementation and Section 2.10 concludes.

### 2.2 Language concepts: *Safe* by example

*Safe* is a first-order polymorphic functional language, whose syntax is similar to that of (first-order) Haskell or ML, but with some facilities to manage memory. Polymorphic data types are defined in the same way as in Haskell. As an example, we have the following **data** declarations of binary search trees and lists:

$$\begin{aligned} \mathbf{data} \text{ BSTree } \alpha &= \text{Empty} \mid \text{Node } (\text{BSTree } \alpha) \alpha (\text{BSTree } \alpha) \\ \mathbf{data} [\alpha] &= [] \mid (\alpha : [\alpha]) \end{aligned}$$

Functions are defined as a set of equations with the same syntax as Haskell functions. For example,

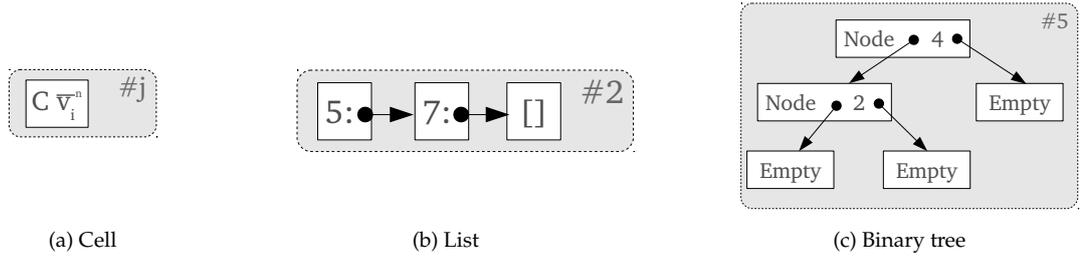


Figure 2.1: Graphical representation of cells and regions.

the following definitions compute the length of a list and the number of nodes in a tree:

$$\begin{aligned}
 \text{length} &:: [\alpha] \rightarrow \text{Int} \\
 \text{length} \ [] &= 0 \\
 \text{length} \ (x : xs) &= 1 + \text{length} \ xs \\
 \\ 
 \text{size} &:: \text{BSTree } \alpha \rightarrow \text{Int} \\
 \text{size} \ \text{Empty} &= 0 \\
 \text{size} \ (\text{Node } i \ x \ d) &= 1 + \text{size } i + \text{size } d
 \end{aligned}$$

*Safe's* memory model is based on *heap regions*. Regions are disjoint parts of the heap where data structures are built. A region can be created and disposed of in constant time.

A *cell* is a piece of memory big enough to hold a data constructor with its parameters. In implementation terms, a cell contains the identifier of a data constructor, and a representation of the values to which this constructor is applied. These values can be either basic (integers or booleans), or pointers to other cells. With the term “big enough” we mean that a cell being disposed of the heap may be immediately reused by the runtime system. A naive implementation would define this size as the space taken by the biggest constructor (i.e. with the highest number of parameters). In a more efficient approach we would have a fixed number of cell sizes, all of them multiples of the smallest one. In any case, the aim is to be able to reuse a cell in constant time.

We represent regions and cells as in Figure 2.1. A cell is depicted as a white square which contains the constructor and its arguments to which it is applied. Pointers are represented via arrows between cells, whereas basic values are shown in the cell itself. Shaded rectangles correspond to regions, which are labelled with a number identifying them (see Figure 2.1a). As an example, Figure 2.1b shows a list of integers, in which the constructor ( $:$ ) is shown in infix form.

Cells are combined in order to build *data structures*. A data structure (DS in the following) is the set of cells that results from taking a particular cell (the *root*) and following the transitive closure of the relation  $C_1 \rightarrow C_2$ , which denotes that  $C_1$  and  $C_2$  are cells *of the same type*, and there is a pointer in  $C_1$  to  $C_2$ . This relation will be described more precisely in the next chapter, as well as the concept of type. An important thing to note, however, is that we only consider as part of a DS the set of cells with the same type as the root cell. For instance, if we have a list of lists (type  $[[\alpha]]$ ), the cells that make up the recursive spine of the outer list constitute a DS, to which the inner lists do not belong, even when there are pointers from the outer list to them. Each one of the inner lists constitute a separate DS on its own.

During the design of the language several decisions were taken. The first one involves the correspondence between DSs and regions.

**Axiom 2.1.** *A DS completely resides in a single region.*

This decision poses a constraint to the data constructors: the recursive children of a cell (i.e. those with the same type) must belong to the region of the father.

**Axiom 2.2.** *A DS can be part of another DS, and two DS may share a third DS.*

As an example, consider the binary tree of Figure 2.1c. The left and the right subtrees of the root are separate DS, which belong to the whole binary tree, which is another DS.

**Axiom 2.3.** *Basic values (integers and booleans) occurring in the heap do not belong to any region by themselves. They are contained within cells.*

## 2.2.1 Region-based memory management in *Safe*

A distinctive aspect of *Safe* is the way in which regions are created and destroyed:

**Axiom 2.4.** *Allocation of regions takes place at function calls. Deallocation of regions takes place when a function call finishes.*

This implies that new regions are created as functions are called, so there exists a correspondence between the function call stack and regions, which are also created and disposed of in a stack-like fashion. Since function calls have nested lifetimes (for instance, if  $f$  calls to a function  $g$ , the execution of the latter begins after the execution of  $f$  has started, and it finishes before the execution  $f$  has finished), regions also have nested lifetimes. That is why they are stored in a stack-like fashion: the last region being created is the first being destroyed.

The region associated to a given function call  $f$  is its *working region*. The function may create DSs in this region, provided these DS are not accessed outside the function's context, since they will be destroyed when the function finishes. A function may also access the working regions of the function calls situated below it in the call stack. These regions must be passed as parameters by the functions calling  $f$ . Each region existing at a given execution point is uniquely identified by a natural number ranging from 0 (which identifies the bottommost region in the stack) to the number  $k$  of active regions minus one (which identifies the topmost one).

An important point is the fact that regions are not handled directly by the *Safe* programmer. The compiler determines which DSs will be created in the working region and which regions should be passed as parameters between functions. However, in order to get an idea on how regions are inferred, we will consider a syntactically-extended version of *Safe*, which we call *Safe with regions*. In this version regions become apparent. The main syntactical additions of *Safe with regions* include the following:

- A function definition may have additional region parameters  $r_1 \dots r_m$  separated by a @ from the rest of formal parameters. As an example, we may have the following function definition:

$$f \ x_1 \ x_2 \ x_3 \ @ \ r_1 \ r_2 = \dots$$

These extra parameters will contain, at runtime, the identifiers of the regions in which  $f$  will build its output. These regions are, in general, different from the regions in which the DSs given as input live. However, in those cases when the result is build upon a DS passed as parameter, at least one output region will be the same as the one in which that DS lives. The compiler determines, in each call to  $f$ , whether output regions are equal or different from the input regions.

- The working region is referred to by the identifier *self*.
- When calling a function with these extra parameters, the region arguments are also separated from the rest of arguments by the @ symbol. For example:

$$f\ 4\ x\ z\ @\ self\ r_1$$

where  $r_1$  is a region variable in scope.

- Each constructor expression is attached a region variable which contains, at runtime, the identifier of the region where the resulting cell will be built. For example:

$$[]@r_2 \quad (4 : []@self)@self$$

In the latter example, the outermost *self* annotates the application of the list constructor (:).

**Example 2.5.** Consider a function *append* for concatenating two lists. The following is *Safe* code, as written by the programmer:

$$\begin{aligned} \text{append} &:: [\alpha] \rightarrow [\alpha] \rightarrow [\alpha] \\ \text{append} \ [] \quad \quad \quad ys &= ys \\ \text{append} \ (x : xs) \quad ys &= x : \text{append} \ xs \ ys \end{aligned}$$

This function is annotated by the compiler as follows:

$$\begin{aligned} \text{append} \ [] \quad \quad \quad ys @ r &= ys \\ \text{append} \ (x : xs) \quad ys @ r &= (x : \text{append} \ xs \ ys @ r) @ r \end{aligned}$$

There is a new region parameter  $r$ , which is used to build the resulting list, and is passed to the subsequent recursive calls. Assume this function is called in the context of a function  $f$ :

$$f \ xs = \dots \text{append} \ xs \ [3] \dots$$

and that its definition is annotated as follows:

$$f \ xs = \dots \text{append} \ xs \ ((3 : []@self)@self) @ self \dots$$

Figure 2.2a shows the correspondence between call and region stacks when evaluating the call  $f \ [1, 2]$ . Assuming that the identifier of the working region corresponding to this call is number 1, Figure 2.2a shows the regions created when the execution reaches the base case of *append*. This function receives the region identifier bound to the *self* region of  $f$ , that is, 1. Each recursive call to *append* has its own working region (regions 2, 3 and 4), but nothing is built there, since the result is built in region 1, which is the value of the region parameter being propagated through the recursive calls to *append*. When the initial call to *append* finishes, we get the situation of Figure 2.2b. Notice that the result is forced to be built in the same region as the list passed as second parameter. This is because this parameter is reused in the base case of *append*, and a DS must be contained within a single region. The compiler takes this into account when annotating the call to *append*.  $\square$

The working region *self* of a function is used to build temporary DSs which are not part of the result. An example of a function with this kind of behaviour is *treesort*.

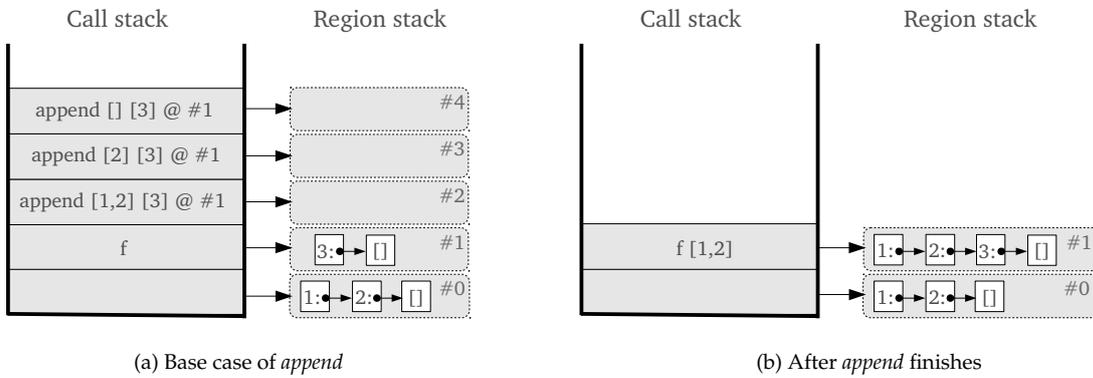


Figure 2.2: Call and region stacks of the *append* function.

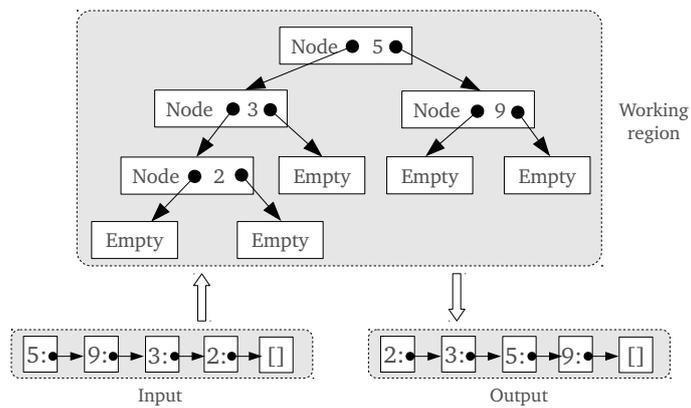


Figure 2.3: DSs involved in the *treesort* function. The working region contains the intermediate representation of the list as a binary tree.

**Example 2.6.** A tree sort algorithm builds a binary search tree from the input list to be sorted. Then it does an inorder traversal of the tree, so that the elements come out in sorted order. Figure 4.12 en la página 156 shows the full *Safe* code with regions. For the purposes of this example, the *treesort* function is enough:

$$\begin{aligned} \textit{treesort} &:: [\alpha] \rightarrow [\alpha] \\ \textit{treesort} \textit{ xs} &= \textit{inorder} (\textit{mkTree} \textit{ xs}) \end{aligned}$$

where *mkTree* builds a binary search tree from the list given as parameter, and *inorder* performs an inorder traversal of a binary search tree by adding the visited elements to a list that is returned as result. Now we show the *Safe* code with regions:

$$\textit{treesort} \textit{ xs} @ r = \textit{inorder} (\textit{mkTree} \textit{ xs} @ \textit{self}) @ r$$

Both functions *inorder* and *mkTree* receive a region parameter specifying where to build the resulting list (resp. tree). The *mkTree* function is given the *self* identifier, so the tree will be built in the working region. The *inorder* function receives the parameter given to *treesort*, which is the output region in which the sorted list will be built (Figure 2.3). When *treesort* finishes its working region will disappear from the heap, together with the temporary tree.  $\square$

*Safe* provides a built-in facility for copying data structures: the @ notation. The expression *ys@* returns a copy of the DS pointed to by *ys*. The copy of the data structure will be located in a (possibly) different region, if this does not contradict Axiom 2.1. The copy facility is useful when the programmer does not want to build DS upon already existing ones.

**Example 2.7.** The *append* function of Example 2.5 forces the resulting list to be located in the same region as the list passed as second parameter. This is because the result is linked to this parameter, so that the latter becomes part of the former, and, by Axiom 2.1, they must live in the same region. Let us consider the following variant in which the result is built upon a *copy* of the list passed as second parameter:

$$\begin{aligned} \textit{appendC} [] \quad \textit{ys} &= \textit{ys}@ \\ \textit{appendC} (x : \textit{xs}) \quad \textit{ys} &= x : \textit{appendC} \textit{xs} \textit{ys} \end{aligned}$$

The compiler annotates every copy expression with the region variable in which the copy will be returned. In the case of *appendC* function, it produces the following code with regions:

$$\begin{aligned} \textit{appendC} [] \quad \textit{ys} @ r &= \textit{ys} @ r \\ \textit{appendC} (x : \textit{xs}) \quad \textit{ys} @ r &= (x : \textit{appendC} \textit{xs} \textit{ys} @ r) @ r \end{aligned}$$

The copy of *ys* is created in the output region *r*, which may now be different from the region of the second parameter *ys*. (Figure 2.4).  $\square$

**Example 2.8.** The copy of a DS might not be able to live in a region different from that of the DS being copied. In the following example:

$$\textit{duplicate} \textit{t} = \textit{Node} \textit{t} 0 (\textit{t}@)$$

The original binary tree *t* and its copy *t@* are forced to live in the same region, as they belong to the same DS.  $\square$

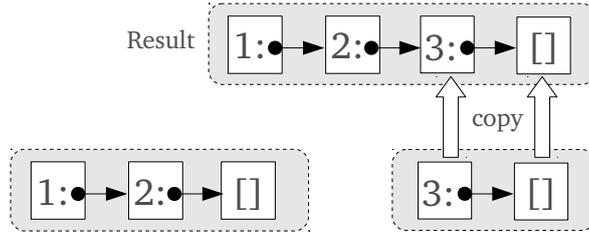


Figure 2.4: Runtime behaviour of *appendC* [1, 2] [3]: the list passed as second parameter is copied, so it does not share cells with the result, with may be built in an independent region.

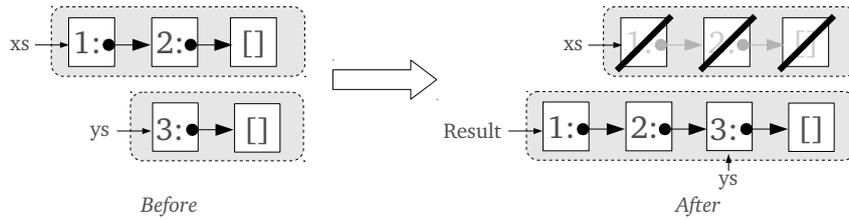


Figure 2.5: Regions of the input parameters before and after a call to *appendD* [1, 2] [3]. Every cell of the input list is removed in each recursive call to *appendD*.

## 2.2.2 Destructive pattern matching

Destructive pattern matching allows the selective disposal of a DS inside a region, without the need of waiting the whole region to be disposed of. This allows the programmer to break the strict nested discipline of nested regions, an issue which was discussed in Section 1.1.1.

Destructive pattern matching, denoted by (!) or a *case!* expression, deallocates the cell corresponding to the outermost constructor of the DS being matched against. In this case we say that the DS involved in the destructive pattern matching is *condemned*.

**Axiom 2.9.** Every function has the following capabilities over their input data structures:

- A function may only read a DS which is not a condemned parameter.
- A function may read (before destroying) and destroy a DS which is a condemned parameter.

As an example, we will consider a destructive variant of *append*.

**Example 2.10.** Assume the following definition:

$$\begin{aligned} \text{appendD} \ []! \quad ys &= ys \\ \text{appendD} \ (x : xs)! \quad ys &= x : \text{appendD} \ xs \ ys \end{aligned}$$

The (!) mark in the first parameter specifies that the cell to which the pattern matching is done will be destroyed at runtime. The function destroys the first cons cell of the list passed as first parameter. The remaining cells will be destroyed in the subsequent recursive calls to *appendD*, until we reach the base case in which the empty list matches the first equation, and it is also destroyed (Figure 2.5).

This version of *append* needs no additional heap space: a cell is destroyed and another cell is built in each recursive call. The destruction of the first parameter is reflected in the type of the function:

$$\text{appendD} :: [\alpha]! \rightarrow [\alpha] \rightarrow [\alpha]$$

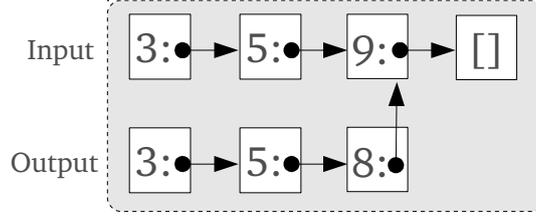


Figure 2.6: Inserting the number 8 in the list [3,5,9]. The part of the list before the new cell must be reconstructed.

□

Destructive pattern matching allows the programmer to define functions requiring constant additional heap space. (see Chapter 3 for more examples). This facility is also very useful for breaking the restriction (imposed by regions) of having DSs with nested lifetimes.

**Example 2.11.** Given a list of elements, the *insertion sort* algorithm starts with an empty list and does successive ordered insertions in it with the elements of the input list. The function for inserting an element in an ordered list is defined as follows:

$$\begin{aligned}
 \text{insert } x \ [] &= [x] \\
 \text{insert } x \ (y : ys) & \\
 & \quad | x \leq y = x : y : ys \\
 & \quad | x > y = y : \text{insert } x \ ys
 \end{aligned}$$

The compiler produces the following region-annotated version:

$$\begin{aligned}
 \text{insert } x \ [] \ @ r &= (x : [] \ @ r) \ @ r \\
 \text{insert } x \ (y : ys) \ @ r & \\
 & \quad | x \leq y = (x : (y : ys) \ @ r) \ @ r \\
 & \quad | x > y = (y : \text{insert } x \ ys \ @ r) \ @ r
 \end{aligned}$$

In the equation guarded by  $x \leq y$  the result is built upon the input  $ys$ , so the region of the output list must be the same as that of the input list (Figure 2.6). As a result, the following *inssort* function,

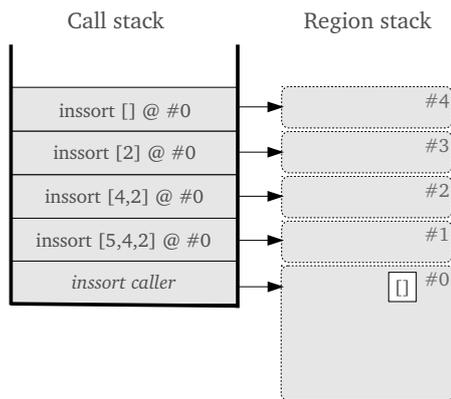
$$\begin{aligned}
 \text{inssort} \ [] &= [] \\
 \text{inssort} \ (x : xs) &= \text{insert } x \ (\text{inssort } xs)
 \end{aligned}$$

whose regions are inferred by the compiler in the following way,

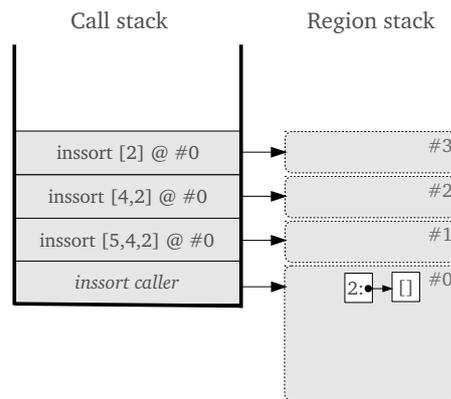
$$\begin{aligned}
 \text{inssort} \ [] \ @ r &= [] \ @ r \\
 \text{inssort} \ (x : xs) \ @ r &= \text{insert } x \ (\text{inssort } xs \ @ r) \ @ r
 \end{aligned}$$

builds every intermediate result in the region of the empty list being created in the base case, that is, the output region  $r$ , whereas the working regions of the calls to *inssort* remain unused. This implies that the *inssort* function has a  $\mathcal{O}(n^2)$  worst-case space complexity, where  $n$  is the number of elements of the input list (Figure 2.7).

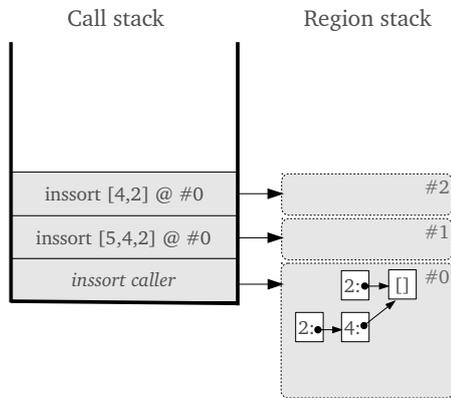
We could return a copy of  $ys$  in the *insert* function. In this way the result is built upon that copy,



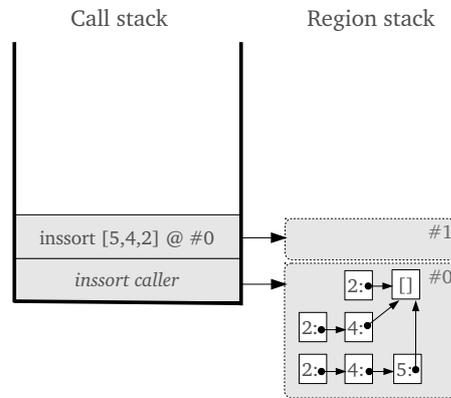
(a) Base case finishes



(b) Call to *inssort [2]* finishes



(c) Call to *inssort [4,2]* finishes



(d) Call to *inssort [5,4,2]* finishes

Figure 2.7: Memory consumption of the *inssort* function.

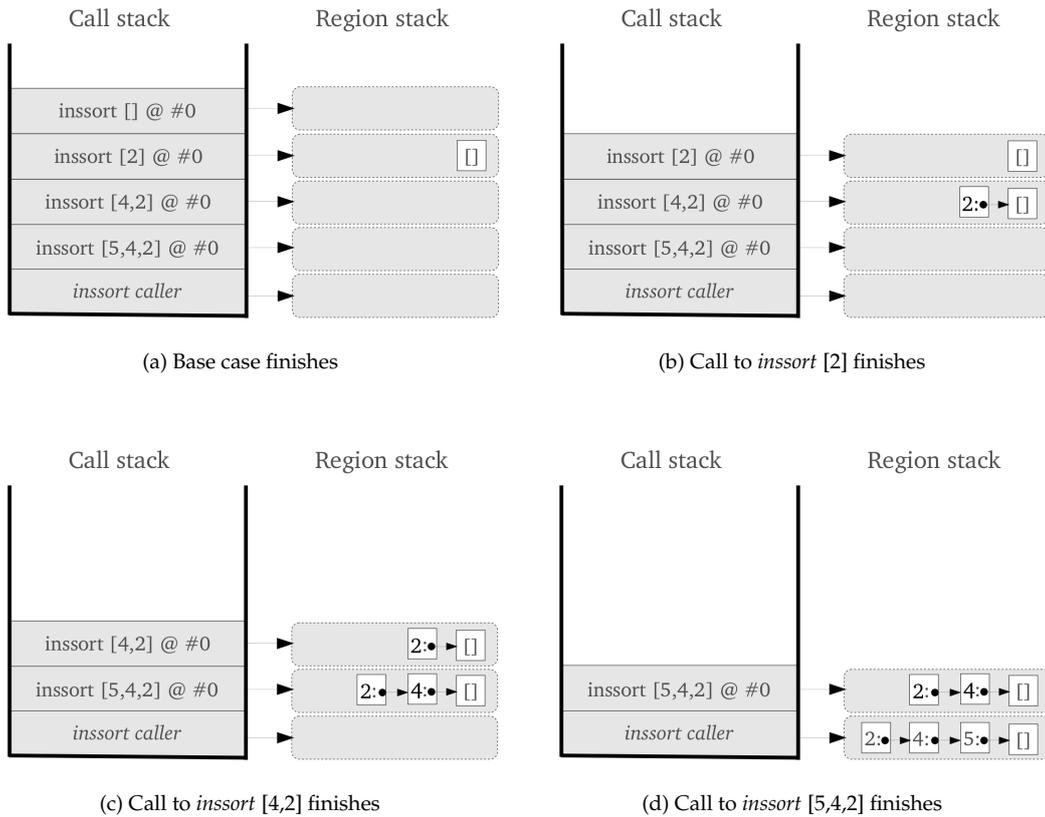


Figure 2.8: When doing a copy of the input list before inserting, *inssort* becomes  $\mathcal{O}(n)$  space.

which does not necessarily live in the same region as the input list. This will allow the function *inssort* to build the intermediate lists in the working region *self*:

$$\begin{aligned} \textit{inssort} \ [] \quad @ r &= [] @ r \\ \textit{inssort} \ (x : xs) \ @ r &= \textit{insert} \ x \ (\textit{inssort} \ xs \ @ \textit{self}) \ @ r \end{aligned}$$

This results in an algorithm of  $\mathcal{O}(n)$  space complexity (see Figure 2.8). However, it could not be trivial for the programmer to discover that doing a copy may result in avoiding such memory leaks. A more direct approach is to consider a destructive version of *insert*:

$$\begin{aligned} \textit{insertD} \ x \ []! &= [x] \\ \textit{insertD} \ x \ (y : ys)! & \\ & \quad | x \leq y = x : (y : ys) \\ & \quad | x > y = y : \textit{insertD} \ x \ ys \end{aligned}$$

This version only needs an additional cell in memory to build the new list node. Assuming that we replace the call to *insert* by *insertD* in the *inssort* function, Figure 2.9 shows the state of the output region when returning from every recursive call. We can even develop a destructive version of *inssort* that also

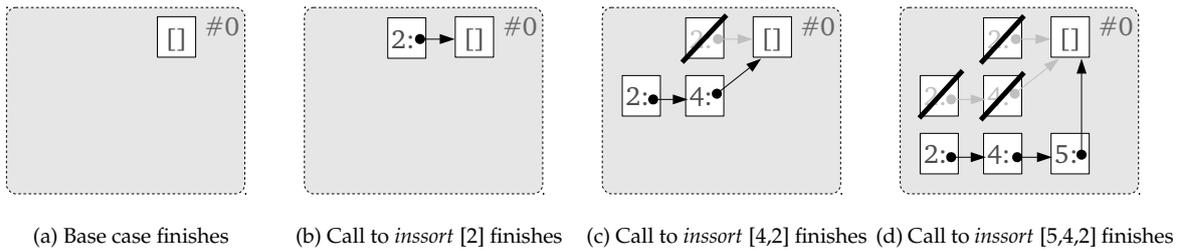


Figure 2.9: Insertion sort destroying the intermediate results in each call to *insert*

```

void PushRegion ()      -- creates a top empty region
void PopRegion ()      -- removes the topmost region
cell ReserveCell ()    -- returns a fresh cell
void InsertCell (p, j) -- inserts cell p into region j
void ReleaseCell (p)   -- releases cell p

```

Figure 2.10: The interface of the *Safe Memory Management System*.

disposes of the input list.

```

insertD []! @ r = [] @ r
insertD (x : xs)! @ r = insertD x (insertD xs @ r) @ r

```

The cell of the input list being destroyed in the pattern matching can be reused by the *insertD* function. Thus *insertD* needs no additional heap space for building the result. □

### 2.2.3 Runtime system implementation

As we said above, the heap is implemented as a stack of regions. Each region is pushed initially empty, this action being associated to a *Safe* function invocation. During function execution new cells can be added to (or removed from) any active region as a consequence of constructor applications and destructive pattern matching. Upon function termination the whole topmost region is deallocated. In Figure 2.10 we show, in a *Java*-like syntax, the main interface between a running *Safe* program and the *Memory Management System* (MMS). The MMS maintains a pool of fresh cells, so that ‘allocating’ and ‘deallocating’ a cell respectively mean removing it from, or adding it to the pool.

Notice that access to an arbitrary region is needed in *InsertCell*, whereas *ReleaseCell* is only provided with the cell pointer as an argument. We have implemented all the methods running in constant time by representing the regions and the pool as circular doubly-linked lists. Figure 2.11 shows a picture of the heap. Removing a region amounts to joining two circular lists, which can obviously be done in constant time. The region stack is represented by a static array of dynamic lists, so that constant time access to each region is provided.

### 2.2.4 Full-Safe vs Core-Safe

The functions presented in previous sections were written in *Full-Safe*, which is the language in which the programmer writes his programs. However, *Full-Safe* results cumbersome when designing program analyses, since the number of language syntactic constructs to consider becomes overwhelming.

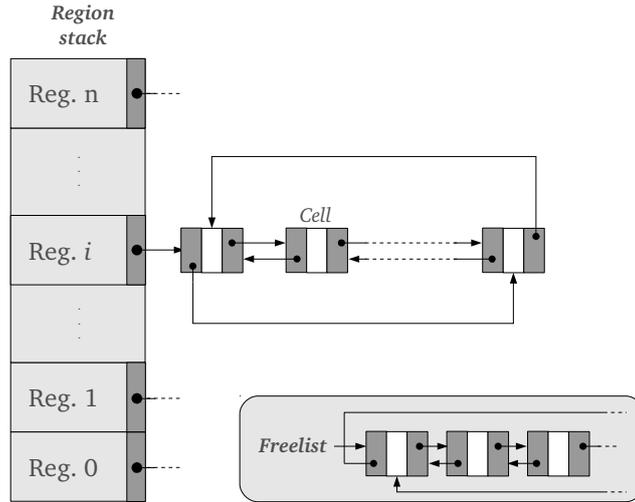


Figure 2.11: A picture of the *Safe* Virtual Machine heap and fresh cells pool

For this reason, we have a simplified variant of *Full-Safe* (which is called *Core-Safe*), with a fewer number of syntactic expressions. This is similar to the translation of Haskell programs into a Core language, as done in the GHC compiler [60]. The details of the translation phase from *Full-Safe* to *Core-Safe* are beyond the scope of this thesis (see [31] for details), but this process follows these general guidelines:

- Each function is represented by a single equation.
- Pattern matching is translated into **case** expressions, whose syntax are similar to that of Haskell.
- Destructive pattern matching is translated into **case!** expressions, which will be explained in Section 2.3.
- Region variables are made explicit in *Core-Safe*.
- Only atomic expressions (constants and variables) are allowed in function and constructor applications. Non-atomic expressions occurring inside function and constructor arguments must be introduced via **let** bindings, in the style of A-normal form [44]. For instance,  $f(2 + 4)$  is transformed into **let**  $z = 2 + 4$  **in**  $f z$ .

**Example 2.12.** The translation phase applied to the *append* and *appendD* functions defined previously yields the following result:

$$\begin{aligned}
 \mathit{append} \ xs \ ys \ @ \ r &= \mathbf{case} \ xs \ \mathbf{of} \\
 &[] \rightarrow ys \\
 (x : xx) &\rightarrow \mathbf{let} \ x_1 = \mathit{append} \ xx \ ys \ @ \ r \ \mathbf{in} \ (x : x_1)@r
 \end{aligned}$$

$$\begin{aligned}
 \mathit{appendD} \ xs \ ys \ @ \ r &= \mathbf{case!} \ xs \ \mathbf{of} \\
 &[] \rightarrow ys \\
 (x : xx) &\rightarrow \mathbf{let} \ x_1 = \mathit{appendD} \ xx \ ys \ @ \ r \ \mathbf{in} \ (x : x_1)@r
 \end{aligned}$$

□

<b>Prog</b> $\ni$	<i>prog</i>	$\rightarrow$	$\overline{data}_i; \overline{def}_i; e$	
<b>DecData</b> $\ni$	<i>data</i>	$\rightarrow$	<b>data</b> $T \overline{a}_i @ \overline{\rho}_j = \overline{C}_i \overline{s}_{ij}^{n_i} @ \rho$	
<b>DecFun</b> $\ni$	<i>def</i>	$\rightarrow$	$f \overline{x}_i @ \overline{r}_j = e$	
<b>Exp</b> $\ni$	<i>e</i>	$\rightarrow$	$a$	{atom: literal $c$ or variable $x$ }
			$  x @ r$	{copy}
			$  a_1 \oplus a_2$	{basic operator application}
			$  C \overline{a}_i @ r$	{constructor application}
			$  f \overline{a}_i @ \overline{r}_j$	{function application}
			$  \mathbf{let} x_1 = e_1 \mathbf{in} e_2$	{ <b>let</b> declaration: nonrecursive, monomorphic}
			$  \mathbf{case} x \mathbf{of} \overline{C}_i \overline{x}_{ij}^{n_i} \rightarrow e_i$	{read-only pattern matching}
			$  \mathbf{case!} x \mathbf{of} \overline{C}_i \overline{x}_{ij}^{n_i} \rightarrow e_i$	{destructive pattern matching}

Figure 2.12: *Core-Safe* language definition.

Since every analysis and inference algorithm in this thesis works at the *Core-Safe* level, this language will be described in detail in next section. However, and for the sake of clarity, we will use *Full-Safe* for most medium- and large-sized examples. We shall even use region-annotated *Full-Safe* code, when regions are relevant.

## 2.3 Syntax of *Core-Safe*

In Figure 2.12 we show the syntax of *Core-Safe* programs and expressions. A program *prog* is a sequence of **data** declarations, followed by a sequence of function definitions  $\overline{def}_i$  and a main expression  $e$ , whose result is the result of the program. The **data** declarations section follows a syntax similar to that of Haskell, with the addition of region type variables  $\rho$ . The role of region type variables, and the syntax of the sequence of types  $\overline{s}_{ij}$  are not relevant at this point, and we shall defer their explanation until Chapter 3.

A function definition is a function name  $f$ , followed by a list of formal parameters  $\overline{x}_i$  (which are variables), a list of formal *region* parameters  $\overline{r}_j$  (which are called *region variables*) and the *body* of the function  $e$ . The sets of function symbols, variables and region variables are respectively denoted by **Fun**, **Var** and **RegVar**.

We denote by **Exp** the set of *Core-Safe* expressions. Basic expressions include: atomic expressions (literals or variables), copy expressions (Section 2.2.1), function and constructor applications, and a special kind of function applications that we consider to be built-in: basic operator applications. The set of basic operators  $\oplus$  is left unspecified. We only demand that applications of these operators require no additional heap space and only two stack words for the arguments. In this work we will only make distinction between basic operator applications and the rest of function applications when this distinction is relevant: when translating *Core-Safe* to machine language, and when considering resource consumption. In the rest of cases basic operator applications are treated as plain function applications.

We assume the existence of a set **Cons** of constructor names, and that, for every constructor  $C$ , the set of its recursive positions (denoted by  $RecPos(C)$ ) is known at runtime. For instance,

$$RecPos([]) = \emptyset \quad RecPos(:) = \{2\} \quad RecPos(Empty) = \emptyset \quad RecPos(Node) = \{1, 3\}$$

We also assume that there is no mutual recursion between functions. This is done for the sake of sim-

plicity. Every analysis and the algorithm described in the next chapters can be adapted with relative ease in order to support mutual recursion.

The **let** construct allows having non-recursive, monomorphic intermediate declarations. Throughout this thesis we use the terms *auxiliary* and *main* expression to refer to the  $e_1$  and  $e_2$  components, respectively.

*Core-Safe* supports two kinds of pattern matching: read-only (**case**) and destructive (**case!**). In the latter, the cell against which the patterns are matched is also disposed of, so its space can be reused by the runtime system.

By the notation  $fv(e)$  we denote the set of variables occurring free in  $e$ . These do *not* include region variables and function names. When  $e$  is in the context of a function definition,  $scope(e)$  denotes the set of variables in scope in the given expression (again, excluding region variables and function names).

## 2.4 Semantics of *Safe*

In the following subsections we will define a big-step and a small-step operational semantics for *Safe*. All the correctness results in this thesis are proven with respect to the big-step operational semantics. However, the small-step operational semantics is useful as an intermediate step for deriving the *Safe* virtual machines of Section 2.5.

Both semantics describe how a *Core-Safe* expression  $e$  is reduced to a *value* (normal form). We use  $v, v_i, \dots$  metavariables to denote *values*, which are defined by the following grammar:

$$\begin{array}{ll} \mathbf{Val} \ni v & ::= p \in \mathbf{Loc} & \{ \text{heap pointer} \} \\ & | c \in \mathbf{Int} \cup \mathbf{Bool} & \{ \text{literal: integer or boolean} \} \end{array}$$

Since one of the aims of the language is the inference of safety properties regarding memory pointers, our semantics needs a model of the heap. A *heap*  $h$  is defined as a finite mapping from heap pointers to construction cells. *Heap pointers* specify memory locations. We assume the existence of a denumerable set of pointers  $\mathbf{Loc}$  and use  $p, p_i, q, \dots$  to denote elements from this set. A construction cell  $w$  is an element of the form  $(j, C \bar{v}_i^n)$ , where  $j$  is a natural number,  $C \in \mathbf{Cons}$  a constructor symbol of arity  $n$ , and  $\bar{v}_i^n$  is the list of values to which  $C$  is applied. The number  $j$  stands for the region of the heap in which the cell is located. With this heap model the region number may be considered as a property of a cell. This implies, on the one hand, that every cell belongs to a region and, on the other hand, that every cell belongs to a *single* region (in other words, regions are *disjoint*). For example, the following mapping  $h_0$  models the heap shown in Figure 2.1b.

$$h_0 = \left[ \begin{array}{l} p_1 \mapsto (2, 5 : p_2) \\ p_2 \mapsto (2, 7 : p_3) \\ p_3 \mapsto (2, []) \end{array} \right]$$

The notation  $region(w)$  represents the region where  $w$  lives (that is, the first component of the pair  $(j, C \bar{v}_i^n)$ ), whereas  $fresh_h(p)$  denotes that the pointer  $p$  is fresh in  $h$ , that is, it does not occur neither in its domain nor in their cells.

## 2.4.1 Big-step semantics

In Figure 2.14 we show the big-step operational semantics of *Core-Safe* expressions. A judgement of the form  $E \vdash h, k, e \Downarrow h', k, v$  means that expression  $e$  is successfully reduced to a normal form  $v$  under a runtime environment  $E$  and a heap  $h$  with  $k + 1$  regions (ranging from 0 to  $k$ ) and that a final heap  $h'$  with the same number  $k$  of regions<sup>1</sup> is produced as a side effect. A runtime environment  $E$  (also denoted *value environment*) is a pair of partial functions in  $(\mathbf{Var} \rightarrow \mathbf{Val}) \times (\mathbf{RegVar} \rightarrow \mathbb{N})$  which map, respectively, program variables  $x$  to values and region variables  $r$  to actual regions (i.e. natural numbers) in the heap. By abuse of notation, and since we use different metavariables to distinguish program variables from region variables (respectively  $x$  and  $r$ ), we will consider both mappings as a single one. Whenever we use  $E(x)$  or  $E(r)$  we will know which mapping we are referring to. We adopt the convention that, for every value environment  $E$ , if  $c$  is a literal,  $E(c) = c$ .

We assume that, during the evaluation of an expression, a program signature  $\Sigma$  is propagated through the  $\Downarrow$  judgements. This signature maps function names to program definitions. Since the function name itself occurs in its definition, we shall use the notation  $(f \bar{x}_i^n @ \bar{r}_j^m = e_f) \in \Sigma$  for denoting the result of  $\Sigma(f)$ . Sometimes a signature will be made explicit in the semantic judgements when it is attached to the arrow symbol ( $\Downarrow_\Sigma$ ).

The semantics of a program  $prog \equiv \overline{data}_i; \overline{def}_i; e$  is the result of evaluating its main expression  $e$  in an environment  $\Sigma$  containing all the function declarations  $\overline{def}_i$ , under an empty heap with a single region and a value environment which maps the *self* identifier to that region:

$$[self \mapsto 0] \vdash [], 0, e \Downarrow h', k', v \quad (2.1)$$

Now we explain in detail the semantic rules. Rules *[Lit]* and *[Var]* just say that literals and heap pointers are normal forms. Rule *[Copy]* executes a copy expression by copying the data structure pointed to by  $p$  and living in a region  $j'$  into a (possibly different) region  $j$ . The runtime system function *copy* follows the pointers in recursive positions of the structure starting at  $p$  and creates in region  $j$  a copy of all recursive cells. Some restricted type information is assumed to be available in our runtime system (namely, the recursive positions of each constructor) so that this function can be implemented.

**Definition 2.13.** The *copy* function is defined as follows:

$$\begin{aligned} copy(h_0[p \mapsto (k, C \bar{v}_i^n)], p, j) &= (h_n \uplus [p' \mapsto (j, C \bar{v}_i^n)], p') \\ \text{where } fresh_{h_n}(p') & \\ \forall i \in \{1..n\}. (h_i, v'_i) &= \begin{cases} (h_{i-1}, v_i) & \text{if } i \notin RecPos(C) \\ copy(h_{i-1}, v_i, j) & \text{otherwise} \end{cases} \end{aligned}$$

Should *copy* find a dangling pointer during the traversal, then the whole rule would fail. If there is no failure, the normal form becomes a fresh pointer  $p'$  pointing to the copy. The pointers in non recursive positions of all the copied cells are kept identical in the new cells. This implies that both data structures (the original and the copy), may share some subparts. For instance, if the original DS is a list of lists, the structure created by *copy* is a copy of the outermost list, while the innermost lists become shared between the old and the new list (see Figure 2.13).

Rule *[App]* shows when a new region is allocated. Notice that the body of the function is executed in a heap with  $k + 2$  regions (from 0 to  $k + 1$ ). The formal identifier *self* is bound to the newly created

<sup>1</sup>Actually, the latter  $k$  is redundant, as the final heap always has the same number of regions as the initial one. However, in this thesis we shall make the  $k$  of the final configuration explicit.

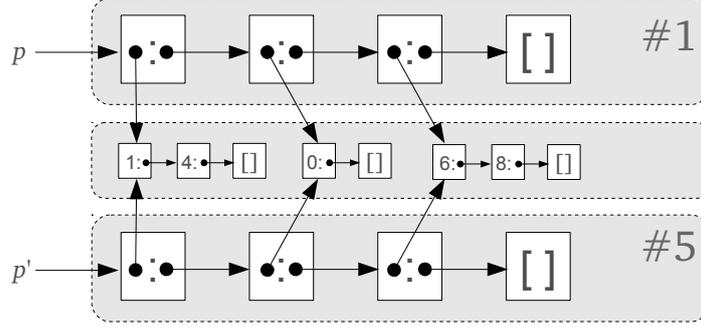


Figure 2.13: Result (pointed to by  $p'$ ) of copying the list of lists pointed to by  $p$  into the region number 5. Only the outermost list spine is copied, whereas the innermost lists are shared between original and copy.

region  $k + 1$  so that the function body may create DSs in this region or pass this region as a parameter to other function calls. Before returning from the function, all cells created in region  $k' + 1$  are deleted. This action is a source of possible dangling pointers. By the notation  $h \downarrow_k$  we denote the heap obtained by deleting from  $h$  those bindings living in regions greater than  $k$ :

$$h \downarrow_k \stackrel{\text{def}}{=} h \upharpoonright_{P(k,h)} \quad \text{where } P(k,h) = \{p \in \text{dom } h \mid \text{region}(h(p)) \leq k\}$$

Rule [Cons] generates a fresh location  $p$  pointing to the newly constructed cell. The parameters of the corresponding constructor are looked up into the value environment  $E$ .

Rule [Let] shows the eagerness of the language: first, the auxiliary expression  $e_1$  is reduced to normal form and then the main expression  $e_2$  is evaluated. In the latter evaluation the environment is extended by binding the program variable  $x_1$  to the normal form to which  $e_1$  is reduced.

The [Case] rule is the usual one for an eager language, whereas the [Case!] rule expresses what happens in a destructive pattern matching: the binding of the discriminant variable disappears from the heap. This action is another source of possible dangling pointers.

Now we use these semantics to prove some general properties of the language. In principle we are interested only in those value environments that map the *self* identifier to the highest possible region number, and the remaining region variables to numbers strictly lower than the one bound to *self*. When a value environment meets these requirements, it is said to be *admissible*.

**Definition 2.14.** A value environment  $E$  is admissible with respect to  $k$  iff  $E(\text{self}) = k$  and for every other region variable  $r \in \text{dom } E$  it holds that  $E(r) < k$ .

The next Proposition shows that admissibility is preserved by the evaluation of an expression.

**Proposition 2.15.** Let us consider a judgement  $E \vdash h, k, e \Downarrow h', k, v$  in which  $E$  is admissible w.r.t.  $k$ . Then any value environment occurring in the derivation of this judgement is also admissible w.r.t. its corresponding  $k$ .

*Proof.* The property is true at the initial judgement and is preserved in every inductive rule. The only relevant case is the [App] rule.  $\square$

It is easy to show that the initial value environment in the execution of a *Core-Safe* program (2.1) is admissible and, hence, that all the value environments taking place in the execution of the program are admissible. This allows us to leave out the conditions  $j \leq k$  in [Copy] and  $E(r) \leq k$  in [Cons], since they are guaranteed to hold when their corresponding judgements are admissible.

$$\begin{array}{c}
\frac{}{E \vdash h, k, c \Downarrow h, k, c} \text{ [Lit]} \\
\\
\frac{}{E[x \mapsto v] \vdash h, k, x \Downarrow h, k, v} \text{ [Var]} \\
\\
\frac{}{E \vdash h, k, a_1 \oplus a_2 \Downarrow h, k, E(a_1) \oplus E(a_2)} \text{ [PrimOp]} \\
\\
\frac{j \leq k \quad (h', p') = \text{copy}(h, p, j)}{E[x \mapsto p, r \mapsto j] \vdash h, k, x @ r \Downarrow h', k, p'} \text{ [Copy]} \\
\\
\frac{E(r) \leq k \quad \text{fresh}_h(p)}{E \vdash h, k, C \bar{a}_i^n @ r \Downarrow h \uplus [p \mapsto (E(r), C \bar{E}(a_i)^n)], k, p} \text{ [Cons]} \\
\\
\frac{(g \bar{y}_i^n @ \bar{r}_j^m = e_g) \in \Sigma \quad \overline{[y_i \mapsto E(a_i)^n, r_j \mapsto E(r_j)^m, \text{self} \mapsto k+1]} \vdash h, k+1, e_g \Downarrow h', k+1, v}{E \vdash h, k, g \bar{a}_i^n @ \bar{r}_j^m \Downarrow h' |_{k, k, v}} \text{ [App]} \\
\\
\frac{E \vdash h, k, e_1 \Downarrow h_1, k, v_1 \quad E \uplus [x_1 \mapsto v_1] \vdash h_1, k, e_2 \Downarrow h', k, v}{E \vdash h, k, \mathbf{let } x_1 = e_1 \mathbf{ in } e_2 \Downarrow h', k, v} \text{ [Let]} \\
\\
\frac{E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, e_r \Downarrow h', k, v}{E[x \mapsto p] \vdash h[p \mapsto (j, C_r \bar{v}_i^{nr}), k, \mathbf{case } x \mathbf{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i^n \Downarrow h', k, v]} \text{ [Case]} \\
\\
\frac{E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, e_r \Downarrow h', k, v}{E[x \mapsto p] \vdash h \uplus [p \mapsto (j, C_r \bar{v}_i^{nr}), k, \mathbf{case! } x \mathbf{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i^n \Downarrow h', k, v]} \text{ [Case!]}
\end{array}$$

Figure 2.14: Big-step operational semantics of Core-Safe expressions.

We are interested in the following question: given a value environment  $E$ , an initial heap  $h$  with  $k$  regions, and a *Core-Safe* expression (assuming  $E$  admissible w.r.t.  $k$ ), is the judgement  $E \vdash h, k, e \Downarrow h', k, v$  derivable for some  $h'$  and  $v$ ? The answer would be negative if the derivation of  $E \vdash h, k, e \Downarrow h', k, v$  requires proving a judgement in which none of the rules of Figure 2.14 is applicable. This happens in the following situations:

1. Access to a variable or region variable which does not belong to the domain of the value environment  $E$  (rules  $[Var]$ ,  $[Copy]$ ,  $[App]$ ,  $[Cons]$ ,  $[Case]$  and  $[Case!]$ ).
2. Call to a function not contained in the signature  $\Sigma$  (rule  $[App]$ ).
3. Unsatisfiability of the disjointness conditions imposed by the  $\uplus$  operator in value environments. For example, when we obtain  $E \uplus [x_1 \mapsto v_1]$  and the  $x_1$  variable is already in  $\text{dom } E$ . (rules  $[Let]$ ,  $[Case]$  and  $[Case!]$ ).
4. Impossibility of obtaining a finite derivation for a given judgement.
5. Pattern matching-related errors: the data constructor of the cell pointed to by the discriminant of a **case**(!) does not match any of the provided alternatives (rules  $[Case]$  and  $[Case!]$ ).
6. A literal is found in  $E$ , when a pointer was expected (rules  $[Copy]$ ,  $[Case]$ ,  $[Case!]$ ).
7. Access to a *dangling pointer*, that is, a location  $p$  not belonging to the domain of  $h$  (rules  $[Copy]$ ,  $[Case]$  and  $[Case!]$ ).

The first two points are easy to address. Most compilers provide a contextual constraints checker for ensuring that every variable only occurs in the scope of its definition. *Safe*'s compiler is not an exception, so we will always assume that, during the evaluation of  $e$  under a value environment  $E$ , it holds that  $fv(e) \subseteq \text{dom } E$ . Moreover, we assume that every function being called (direct or indirectly) from  $e$  is defined in the signature  $\Sigma$  under which  $e$  is evaluated. With respect to the third point, an intermediate renaming phase in the *Safe* compiler prevents us from having two variables with the same name and non-disjoint scopes. Therefore we can safely assume that the disjoint union  $E \uplus [x_1 \mapsto v_1]$  in the  $[Let]$  rule, and those occurring in the value environments of the  $[Case]$  and  $[Case!]$  rules are well-defined.

The fourth point (non-finite derivations) arises when dealing with non-terminating expressions. In general, semantic definitions written in a big-step style can only specify finite computations, since a finite proof of the judgement  $E \vdash h, k, e \Downarrow h', k, v$  implicitly assumes that the evaluation of  $e$  terminates under the environment  $E$  and initial heap  $h$ . Whenever we prove a semantic property of a program via the rules of Figure 2.14, the finiteness of the corresponding derivation is implicitly assumed. To approximate whether a program terminates or not under a given input is a currently active research field, and there are numerous techniques for proving termination of programs. This topic is closely related to resource analysis, since a terminating program always consume a finite amount of resources. In Chapter 3.2 this connection will become more apparent.

The fifth and sixth points are addressed with a Hindley-Milner type system [34] which guarantees the absence of failed pattern matchings, since every **case**(!) expression must have a branch for every constructor of the type of its discriminant.

The last point is one of the aims of this thesis. In Chapter 3 we will introduce a type system guaranteeing that dangling pointers are never accessed by a program. First we need to focus on how dangling pointers are created: they arise as a result of removing a binding from the heap. If we inspect the rules in Figure 2.14, it turns out there are two possible sources of dangling pointers:

- The deallocation of the topmost region  $h \upharpoonright_k$  when the evaluation of a function application finishes ([*App*] rule).
- The removal of a pointer mapped by a **case!** discriminant ([*Case!*] rule).

**Example 2.16.** Consider the following region-annotated function definitions:

$$\mathit{copyToSelf} \ xs = xs @ \mathit{self}$$

$$f \ xs \ ys @ r = \mathbf{let} \ zs = \mathit{appendD} \ xs \ ys @ r \ \mathbf{in} \ \mathit{length} \ xs + \mathit{length} \ zs$$

The first one builds a copy of the input list in the working region and returns a reference to it. However, and since the working region disappears when the function finishes, *copyToSelf* actually returns a dangling pointer. In the second function, the *length* function access the *xs* list, which has been previously destroyed by *appendD*. The execution of *length* fails because *xs* refers to a dangling pointer. Both functions are rejected by the type system of Chapter 3.  $\square$

## 2.4.2 Small-step semantics

In Figure 2.15 we show the small-step semantic rules. There are two kinds of judgements:

- The first kind,  $E, h, k_0, k, e \longrightarrow h', k_0, v$ , is applied when an expression  $e$  is evaluated to a value  $v$  in a single step. These correspond to literals, variables, copy expressions and constructors.
- The other kind,  $E, h, k_0, k, e \longrightarrow E', h', k_0, k', e'$ , covers the remaining cases: function application, **let**, **case** and **case!** expressions.

In the configurations,  $k$  denotes the highest region available in  $h$ , as in the big step semantics. The meaning of  $k_0$  will be explained below. Notice that, in the rules, **let** expressions are marked with a natural number  $\delta$  and an environment  $E$ . In rule [*App-s*], the number of available regions is incremented by one, as a new local region is allocated and assigned number  $k + 1$ . Additionally, this rule discards the environment  $E$ , as in the function body only the arguments and the *self* region are in scope. However, and due to **let** expressions, a continuation is possible after function application. Thus we need to recover the discarded environment and the original value of  $k$ . The environment and the number  $\delta$  are attached to the **let**-binding in order to remember the newly created regions during the evaluation of the auxiliary expression, so that the original  $k$  can be later recovered. The initial values of  $\delta$  and  $E$  are respectively 0 and  $\perp$ , which are assumed to be annotated in the text of the program being executed. Rule [*Let2-s*] saves the environment for the first time, whereas the [*Let4-s*] rule updates the information as necessary during the evaluation of the auxiliary expression. In case this evaluation is successful, rules [*Let1-s*] or [*Let3-s*] are applied to proceed with the evaluation of the main expression  $e_2$ .

New regions being created during the evaluation of the bound expression  $e_1$  of a **let** cannot contain the result of the evaluation because these regions are removed from the heap when  $e_1$  is reduced to a normal form. Region  $k_0$  denotes the highest region that must be available when the machine stops reducing the expression. It can be considered a lower watermark indicating which was the topmost region when the expression under evaluation began to be evaluated. When a normal form is reached, we remove all the regions  $\{k, k - 1, \dots, k_0 + 1\}$  from the heap. Initially  $k = k_0 = 0$ . Rule [*App-s*] increments  $k$  while rules [*Lit-s*], [*Var-s*], [*Copy-s*] and [*Cons-s*] discard all the local regions greater than  $k_0$ .

$$\begin{array}{c}
\frac{k \geq k_0}{E, h, k_0, k, c \longrightarrow h \mid_{k_0, k_0, c}} \text{ [Lit-s]} \\
\\
\frac{k \geq k_0}{E[x \mapsto v], h, k_0, k, x \longrightarrow h \mid_{k_0, k_0, v}} \text{ [Var-s]} \\
\\
\frac{k \geq k_0 \quad k \geq j \quad (h', p') = \text{copy}(h, p, j)}{E[x \mapsto p, r \mapsto j], h, k_0, k, x @ r \longrightarrow h' \mid_{k_0, k_0, p'}} \text{ [Copy-s]} \\
\\
\frac{E(r) \leq k \quad \text{fresh}_h(p)}{E, h, k_0, k, C \bar{a}_i^n @ r \longrightarrow (h \uplus [p \mapsto (E(r), C \bar{E}(a_i)^n)]) \mid_{k_0, k_0, p}} \text{ [Cons-s]} \\
\\
\frac{(g \bar{y}_i^n @ \bar{r}_j^m = e_g) \in \Sigma}{E, h, k_0, k, g \bar{a}_i^n @ \bar{r}_j^m \longrightarrow [\bar{y}_i \mapsto \bar{E}(a_i)^n, \bar{r}_j \mapsto \bar{E}(r_j)^m, \text{self} \mapsto k+1], h, k_0, k+1, e_g} \text{ [App-s]} \\
\\
\frac{E, h, k, k, e_1 \longrightarrow h', k, v_1}{E, h, k_0, k, \mathbf{let} x_1 =_{\perp}^{\perp} e_1 \mathbf{in} e_2 \longrightarrow E \uplus [x_1 \mapsto v_1], h', k_0, k, e_2} \text{ [Let1-s]} \\
\\
\frac{E, h, k, k, e_1 \longrightarrow E', h', k, k + \eta, e'_1}{E, h, k_0, k, \mathbf{let} x_1 =_{\perp}^{\perp} e_1 \mathbf{in} e_2 \longrightarrow E', h', k_0, k + \eta, \mathbf{let} x_1 =_{\eta}^E e'_1 \mathbf{in} e_2} \text{ [Let2-s]} \\
\\
\frac{E' \neq \perp \quad E, h, k, k + \delta, e_1 \longrightarrow h', k, v_1}{E, h, k_0, k + \delta, \mathbf{let} x_1 =_{\delta}^{E'} e_1 \mathbf{in} e_2 \longrightarrow E' \uplus [x_1 \mapsto v_1], h', k_0, k, e_2} \text{ [Let3-s]} \\
\\
\frac{E' \neq \perp \quad E, h, k, k + \delta, e_1 \longrightarrow E'', h', k, k + \eta, e'_1}{E, h, k_0, k + \delta, \mathbf{let} x_1 =_{\delta}^{E'} e_1 \mathbf{in} e_2 \longrightarrow E'', h', k_0, k + \eta, \mathbf{let} x_1 =_{\eta}^{E'} e'_1 \mathbf{in} e_2} \text{ [Let4-s]} \\
\\
\frac{E(x) = p}{E, h[p \mapsto (j, C_r \bar{v}_i^{nr})], k_0, k, \mathbf{case} x \mathbf{of} C_i \bar{x}_{ij}^{ni} \rightarrow e_i^m \longrightarrow E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}], h, k_0, k, e_r} \text{ [Case-s]} \\
\\
\frac{E(x) = p}{E, h \uplus [p \mapsto (j, C_r \bar{b}_i^{nr})], k_0, k, \mathbf{case!} x \mathbf{of} C_i \bar{x}_{ij}^{ni} \rightarrow e_i^m \longrightarrow E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}], h, k_0, k, e_r} \text{ [Case!-s]}
\end{array}$$

Figure 2.15: Small-step operational semantics of Core-Safe expressions.

This small-step semantics is equivalent to the big-step semantics described in Section 2.4.1: for any  $k$  and  $k_0 \leq k$ ,  $E \vdash h, k, e \Downarrow h', k, v$  if and only if  $E, h, k_0, k, e \longrightarrow^* h' \mid_{k_0}, k_0, v$ . However, we leave out the proof of this fact, since these semantics are only of interest for deriving the SAFE-M2 and SVM machines described below. We shall directly prove the equivalence between big-step semantics and the translation to SVM code in Section 2.8.

## 2.5 Virtual machines

In this section we formally describe two abstract machines for executing *Safe* expressions: SAFE-M2 and SVM (*Safe Virtual Machine*). The former is functional and it serves as an intermediate step between the SVM and the small-step semantics defined previously. The SVM is an imperative abstract machine based on the execution of a sequence of instructions.

### 2.5.1 SAFE-M2 abstract machine

Our next refinement is to introduce an abstract machine, called SAFE-M2 because there was a previous one called SAFE-M1 now abandoned. Both are named after Sestoft's *Mark-1* and *Mark-2* abstract machines [104]. A configuration of the machine is a 7-tuple  $(h, k_0, k, e, E, S, \Sigma)$ , where  $h$  is the heap,  $k_0, k$  are region numbers as in the small-step semantics,  $e$  is the expression being executed,  $E$  is the runtime environment,  $S$  is a stack, and  $\Sigma$  is a function giving the code of every defined *Core-Safe* function. In Figure 2.16 we show the transitions of the abstract machine SAFE-M2. The only new element w.r.t. the small-step semantics is the stack  $S$ , which is intended to replace the annotations in **let** expressions. A stack is a sequence of *continuation frames* of the form  $(k_0, x_1, e_2, E)$ . More precisely,

$$S ::= (k_0, x_1, e_2, E) : S \quad \{ \text{where } k_0 \in \mathbb{N}, x_1 \in \mathbf{Var}, e_2 \in \mathbf{Exp}, E \in (\mathbf{Var} \rightarrow \mathbf{Val}) \times (\mathbf{RegVar} \rightarrow \mathbb{N}) \}$$

$$\mid [] \quad \{ \text{empty stack} \}$$

The  $e_2$  corresponds to the pending main expression of a **let** whose auxiliary expression  $e_1$  is under evaluation. Region  $k_0$  is the topmost region where the normal form of  $e_2$  will be returned; higher regions will be disposed of after that normal form is reached. Variable  $x_1$  is the **let**-bound variable free in  $e_2$ , and  $E$  is the environment which has been kept before the evaluation of  $e_1$ , and that will be restored before the evaluation of  $e_2$ . Corresponding to the inductive small-step semantic rules ( $[Let1-s]$  to  $[Let4-s]$ ), the abstract machine rule  $[Let-M2]$  pushes a continuation to the stack and proceeds with the evaluation of the auxiliary expression  $e_1$ . When the normal form of  $e_1$  is reached in rules  $[Lit2-M2]$  and  $[Var2-M2]$ , the continuation is popped and the machine proceeds with the evaluation of the main expression. If there are no continuations in the stack, the machine stops. This situation would happen if (and only if) the normal form of the main expression is reached.

Notice that the current environment is discarded in rules  $[Lit2-M2]$  and  $[Var2-M2]$  when a continuation is popped from the stack. Also, it is discarded in rule  $[App-M2]$  when a function body is entered and the formal arguments become the only variables in scope. In Section 2.6 this will have the important consequence that tail recursion is translated so that only a constant stack space is needed. Notice also in rule  $[Let-M2]$  that the current environment is saved in the stack but it is not discarded from the control. One important aspect of the translation given in Section 2.6 is that it manages to avoid this implicit duplication of environments.

The current environment is extended with new bindings in the following situations:

Initial/final configuration	Condition	Label
$(h, k_0, k, c, E, S, \Sigma)$ $\rightarrow (h _{k_0}, k_0, k_0, c, E, S, \Sigma)$	$k > k_0$	[Lit1-M2]
$(h, k, k, c, E', (k_0, x_1, e, E) : S, \Sigma)$ $\rightarrow (h, k_0, k, e, E \uplus [x_1 \mapsto c], S, \Sigma)$		[Lit2-M2]
$(h[p \mapsto (j, C \bar{v}_i^n)], k_0, k, x, E[x \mapsto p], S, \Sigma)$ $\rightarrow (h _{k_0}, k_0, k_0, x, E, S, \Sigma)$	$k > k_0$	[Var1-M2]
$(h[p \mapsto (j, C \bar{v}_i^n)], k, k, x, E'[x \mapsto p], (k_0, x_1, e, E) : S, \Sigma)$ $\rightarrow (h, k_0, k, e, E \uplus [x_1 \mapsto p], S, \Sigma)$		[Var2-M2]
$(h, k_0, k, a_1 \oplus a_2, E, S, \Sigma)$ $\rightarrow (h', k_0, k, y, E \uplus [y \mapsto E(a_1) \oplus E(a_2)], S, \Sigma)$	$fresh(y)$	[PrimOp-M2]
$(h[p \mapsto (l, C \bar{v}_i^n)], k_0, k, x @ r, E[x \mapsto p, r \mapsto j], S, \Sigma)$ $\rightarrow (h', k_0, k, y, E \uplus [y \mapsto p'], S, \Sigma)$	$(h', p') = copy(h, p, j)$ $j \leq k, fresh(y)$	[Copy-M2]
$(h, k_0, k, C \bar{a}_i^n @ r, E, S, \Sigma)$ $\rightarrow (h \uplus [p \mapsto (E(r), C \bar{E}(a_i)^n)], k_0, k, y, E \uplus [y \mapsto p], S, \Sigma)$	$E(r) \leq k$ $fresh_h(p), fresh(y)$	[Cons-M2]
$(h, k_0, k, g \bar{a}_i^n @ \bar{r}_j^m, E, S, \Sigma)$ $\rightarrow (h, k_0, k+1, e_g, [\bar{y}_i \mapsto E(a_i)^n, \bar{r}'_j \mapsto E(r_j)^m, self \mapsto k+1], S, \Sigma)$	$(g \bar{y}_i^n @ \bar{r}'_j^m = e_g) \in \Sigma$	[App-M2]
$(h, k_0, k, \mathbf{let} x_1 = e_1 \mathbf{in} e_2, E, S, \Sigma)$ $\rightarrow (h, k, k, e_1, E, (k_0, x_1, e_2, E) : S, \Sigma)$		[Let-M2]
$(h[p \mapsto (j, C_r \bar{v}_i^{nr})], k_0, k, \mathbf{case} x \mathbf{of} \overline{C_i \bar{x}_{ij}^{ni}} \rightarrow e_i, E[x \mapsto p], S, \Sigma)$ $\rightarrow (h, k_0, k, e_r, E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}], S, \Sigma)$		[Case-M2]
$(h \uplus [p \mapsto (j, C_r \bar{v}_i^{nr})], k_0, k, \mathbf{case!} x \mathbf{of} \overline{C_i \bar{x}_{ij}^{ni}} \rightarrow e_i, E[x \mapsto p], S, \Sigma)$ $\rightarrow (h, k_0, k, e_r, E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}], S, \Sigma)$		[Case!-M2]

Figure 2.16: Transition rules of the SAFE-M2 abstract machine.

Initial/final configuration	Condition
$(\text{DECREGION} : is, h, k_0, k, S, cs)$ $\rightarrow (is, h _{k_0}, k_0, k_0, S, cs)$	$k \geq k_0$
$([\text{POPCONT}], h, k, k, b : (k_0, \mathbf{p}) : S, cs[\mathbf{p} \mapsto is])$ $\rightarrow (is, h, k_0, k, b : S, cs)$	
$(\text{PUSHCONT } \mathbf{p} : is, h, k_0, k, S, cs[\mathbf{p} \mapsto is'])$ $\rightarrow (is, h, k, k, (k_0, \mathbf{p}) : S, cs)$	
$(\text{COPY} : is, h, k_0, k, p : j : S, cs)$ $\rightarrow (is, h', k_0, k, p' : S, cs)$	$(h', p') = \text{copy}(h, p, j)$ $j \leq k$
$([\text{CALL } \mathbf{p}], h, k_0, k, S, cs[\mathbf{p} \mapsto is])$ $\rightarrow (is, h, k_0, k+1, S, cs)$	
$(\text{PRIMOP } \oplus : is, h, k_0, k, c_1 : c_2 : S, cs)$ $\rightarrow (is, h, k_0, k, c : S, cs)$	$c = c_1 \oplus c_2$
$([\text{MATCH } l \overline{\mathbf{p}}_j^m], h[S!l \mapsto (j, C_r^m \overline{v}_i^n)], k_0, k, S, cs[\overline{\mathbf{p}}_j \mapsto \overline{is}_j^m])$ $\rightarrow (is_r, h, k_0, k, \overline{v}_i^n : S, cs)$	
$([\text{MATCH! } l \overline{\mathbf{p}}_j^m], h \uplus [S!l \mapsto (j, C_r^m \overline{v}_i^n)], k_0, k, S, cs[\overline{\mathbf{p}}_j \mapsto \overline{is}_j^m])$ $\rightarrow (is_r, h, k_0, k, \overline{v}_i^n : S, cs)$	
$(\text{BUILDENV } \overline{K}_i^n : is, h, k_0, k, S, cs)$ $\rightarrow (is, h, k_0, k, \overline{\text{Item}}_k(K_i)^n : S, cs)$	(1)
$(\text{BUILDCLS } C_r^m \overline{K}_i^n K : is, h, k_0, k, S, cs)$ $\rightarrow (is, h \uplus [b \mapsto (\text{Item}_k(K), C_r^m \overline{\text{Item}}_k(K_i)^n)], k_0, k, b : S, cs)$	$\text{Item}_k(K) \leq k, \text{fresh}(b)$ (1)
$(\text{SLIDE } m n : is, h, k_0, k, \overline{b}_i^m : \overline{b}_i^n : S, cs)$ $\rightarrow (is, h, k_0, k, \overline{b}_i^m : S, cs)$	
(1) $\text{Item}_k(K) \stackrel{\text{def}}{=} \begin{cases} S!j & \text{if } K = \text{Pos } j \in \mathbb{N} \\ c & \text{if } K = \text{Lit } c \\ k & \text{if } K = \text{self} \end{cases}$	

Figure 2.17: Transition rules of the SVM abstract machine.

- In rules  $[\text{Case-M2}]$  and  $[\text{Case!-M2}]$ , as soon as let-bound or case-bound variables become free variables in scope in the continuation expression.
- In rules  $[\text{Copy-M2}]$  and  $[\text{Cons-M2}]$ , the environment is extended with a fresh program variable  $y$ . This is merely an artifact due to the fact that a fresh data structure must be referenced in the control expression.
- In rules  $[\text{Lit2-M2}]$  and  $[\text{Var2-M2}]$ , the environment  $E$  saved in the continuation must be extended with the new binding introduced by **let**.

## 2.5.2 SVM imperative abstract machine

The main differences of the SVM with respect to SAFE-M2 are two:

- Expressions are replaced by a sequence of imperative instructions.

- There are no runtime environments: values and continuations live in the stack.

Instructions and instruction sequences of the SVM are given by the following grammar:

$$\begin{aligned}
\iota & ::= \text{DECREGION} & | & \text{COPY} & | & \text{MATCH } l \bar{\mathbf{p}}_i & | & \text{BUILDCLS } C \bar{K}_i K \\
& & | & \text{POPCONT} & | & \text{CALL } \mathbf{p} & | & \text{MATCH! } l \bar{\mathbf{p}}_i & | & \text{SLIDE } m n \\
& & | & \text{PUSHCONT } \mathbf{p} & | & \text{PRIMOP } \oplus & | & \text{BUILDENV } \bar{K}_i \\
is & ::= [] \\
& & | & \iota : is
\end{aligned}$$

In this section we introduce only the semantics of these instructions in the context of the SVM machine. Their specific role will become clearer when dealing with the translation from *Core-Safe* to SVM code. During the execution a *code store* (resulting from the compilation of program fragments) is kept. A code store maps *code pointers* to instruction sequences. We use the metavariables  $\mathbf{p}, \mathbf{p}_i, \dots$  for denoting code pointers. We assume the existence of a denumerable set **PCode** containing them. The  $l$  variable occurring in the MATCH and MATCH! instructions stands for a natural number representing a stack position (counting from the topmost element). Finally, the elements  $K$  occurring in BUILDENV and BUILDCLS, which will be called *keys*, are generated by the following grammar:

$$\begin{aligned}
K & ::= \text{Pos } j & \{ j \in \mathbb{N}, \text{ stack position } \} \\
& & | & \text{Lit } c & \{ \text{literal} \} \\
& & | & \text{self} & \{ \text{working region identifier} \}
\end{aligned}$$

A SVM configuration  $c$  consists of six components

$$c \equiv (is, h, k_0, k, S, cs)$$

where  $is$  is the instruction sequence currently being executed,  $h$  is a heap with  $k$  regions, and  $k_0$  plays a similar role as in the small-step semantics and the SAFE-M2 abstract machine: it denotes the identifier of the highest region available after a normal form is computed. The  $cs$  variable denotes a code store containing all the program fragments, whereas  $S$  denotes a stack which may contain values, region numbers and continuations of the form  $(k_0, \mathbf{p})$ . We use the metavariable  $b$  for denoting stack values:

$$\begin{aligned}
b & ::= j & \{ \text{where } j \in \mathbb{N} \} \\
& & | & v & \{ \text{where } v \in \mathbf{Val} \} \\
& & | & (k_0, \mathbf{p}) & \{ \text{continuation: } k_0 \in \mathbb{N}, \mathbf{p} \in \mathbf{PCode} \}
\end{aligned}$$

In Figure 2.17 we show the semantics of SVM instructions in terms of transitions between configurations. By  $C_r^m$  we denote the data constructor which is the  $r$ -th in its **data** definition out of a total of  $m$  data constructors. By  $S!j$  we denote the  $j$ -th element of the stack  $S$  counting from the top and starting at 0 (i.e.  $S!0$  is the topmost element). Notice that continuations take up two words in the stack, so, if  $n \geq 2$  we get  $((k_0, \mathbf{p}) : S)!n = S!(n - 2)$ . The notation  $\bar{b}_i^n : S$  abbreviates the stack  $b_1 : \dots : b_n : S$ .

Instruction DECREGION deletes from the heap all the regions, if any, between the current region  $k$  and region  $k_0$ , excluding the latter. It will be used when a normal form is reached.

Instruction POPCONT pops a continuation from the stack or stops the execution if there is none. It is always assumed to be the last element of the current instruction sequence (the notation  $[\text{POPCONT}]$  stands for  $\text{POPCONT} : []$ ). Notice that  $b$  (which is usually a value) is left in the stack so that it can be accessed by the continuation. Instruction PUSHCONT pushes a continuation in the stack. It is used in the

translation of a **let**.

The **COPY** instruction just mimics its counterpart [*Copy-M2*] in the **SAFE-M2** abstract machine.

Instruction **CALL** jumps to a new instruction sequence and creates a new region. It will be used in the compilation of a function application.

Instruction **PRIMOP** operates two basic values located in the stack and replaces them by the result of the operation.

Instruction **MATCH** performs a vectored jump depending on the constructor of the matched closure. The vector of sequences pointed to by the  $\mathbf{p}_j$  corresponds to the compilation of a set of **case** alternatives. The **MATCH!** instruction additionally destroys the matched cell.

The **BUILDENV** instruction receives a list of keys  $K_i$  and creates a portion of environment on top of the stack: If a key  $K$  is a natural number  $j$ , the item  $S!j$  is copied and pushed on the stack; if it is a basic constant  $c$ , it is directly pushed on the stack; if it is the identifier *self*, then the current region number  $k$  is pushed on the stack.

Instruction **BUILDCLS** allocates fresh memory and constructs a new cell in the style of the [*Cons-M2*] rule in **SAFE-M2**. Similarly to **BUILDENV**, it receives a list of keys and uses the same conventions. It also receives the constructor  $C_r^m$  of the cell being created.

Finally, instruction **SLIDE** removes some parts of the stack. It will be used to remove environments when they are no longer needed.

## 2.6 Translating *Core-Safe* into SVM code

A major difference of SVM with respect to M2 is the lack of a value environment  $E$  in the SVM. The values of  $E$  are assumed to live in the stack  $S$ . In order to perform the translation from *Core-Safe* into SVM instructions, we need to set up a correspondence between program variables (which are no longer present during the execution of a SVM program) and the positions of the stack which contain the values of these variables.

The main idea of the translation is to keep a compile-time environment  $\rho$  mapping program variables to stack positions. As the stack grows dynamically, a first idea is to assign numbers to the variables from the bottom of the environment to the top. In this way, if the current environment occupies the top  $m$  positions of the stack and  $\rho(x) = 1$ , then  $S!(m - 1)$  will contain the runtime value corresponding to  $x$  (see Figure 2.18).

A second idea is to reuse the current environment when pushing a continuation into the stack. Let us recall the [*Let-M2*] rule of Figure 2.16. The environment  $E$  being pushed into the stack is the same as the environment in which the auxiliary expression  $e_1$  is evaluated. In a naive implementation, the current environment  $E$  would be duplicated, and its copy pushed into the stack. Our aim is to *share* the environment instead of duplicating it, and to push only the following two elements of the continuation:  $k_0$  and  $e$  (more precisely, the code pointer containing the SVM code of  $e$ ). The variable  $x_1$  is not needed, since the compilation process will ensure that a pointer to its value will be on top of the stack when the continuation is popped.

In order to carry out this sharing between runtime environments, we split the whole compile-time environment  $\rho$  into a list of smaller environments  $[\rho_1, \dots, \rho_n]$ , each one topped with a continuation, except the topmost one ( $\rho_1$ ). Each individual block  $\rho_i$  consists of a triple  $(\Delta_i, l_i, n_i)$  with the actual environment  $\Delta_i$  mapping variables to numbers in the range  $(1 \dots m_i)$ , its length  $l_i = m_i + n_i$ , and an indicator  $n_i$  whose value is 2 for all the blocks except for the first one, whose value is  $n_1 = 0$ . (see Figure 2.19). This value stands for the length of the continuation  $(k_0, \mathbf{p})$ : we assume that a continuation needs

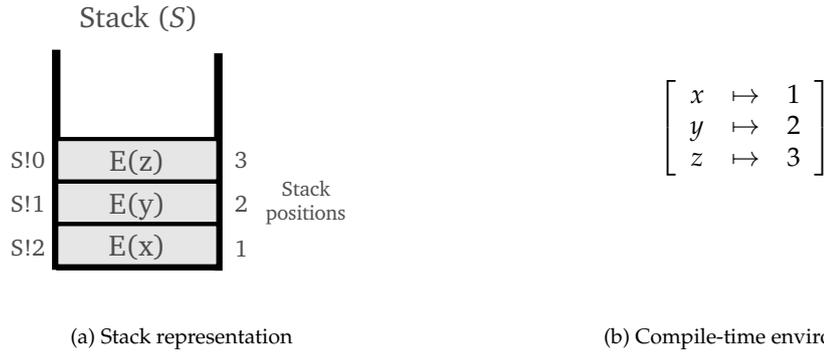


Figure 2.18: A first approach to mapping program variables to stack offsets.

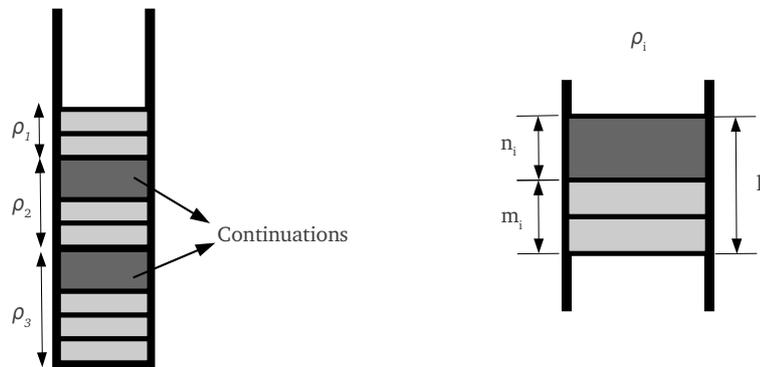


Figure 2.19: Division of  $\rho$  into a list of smaller environments.

two words in the stack and that the remaining items need one word. The positions given by the smaller environments  $\Delta_i$  are relative to the bottommost position of their block in the stack.

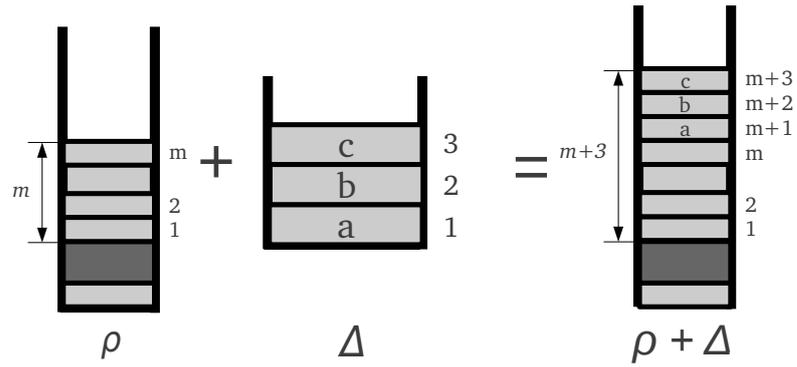
Given this definition of a compile-time environment, we can compute the offset w.r.t. the top of the stack of a given variable  $x$ , defined in the  $k$ -th block. We use the notation  $\rho(x)$  to denote this offset.

$$\rho(x) \stackrel{\text{def}}{=} \sum_{i=1}^k l_i - \Delta_k(x)$$

We assume by convention that  $\rho(\text{self}) = \text{self}$ . As soon we introduce new variables in scope, we need to update the environment  $\rho$  accordingly. Only the topmost environment can be extended with new bindings. We define the following operations and functions on compile-time environments:

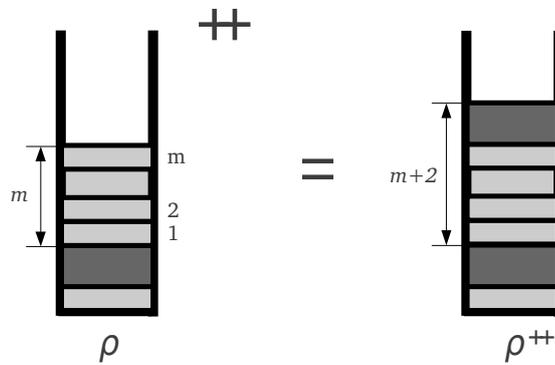
1. **Operator  $+$ .** It adds new bindings to the compile-time environment:

$$((\Delta, m, 0) : \rho) + \overline{[x_i \mapsto j_i^n]} \stackrel{\text{def}}{=} (\Delta \uplus \overline{[x_i \mapsto m + j_i^n]}, m + n, 0) : \rho$$



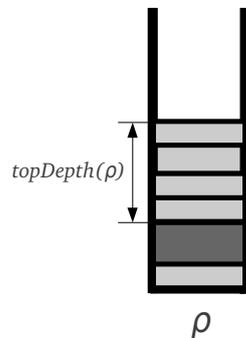
2. **Operator**  $^{++}$ . It finishes the topmost block, and starts a new one above it:

$$((\Delta, m, 0) : \rho)^{++} \stackrel{\text{def}}{=} ([], 0, 0) : (\Delta, m + 2, 2) : \rho$$



3. **Function** *topDepth*. It returns the length of the topmost block. Its result is undefined when applied to empty environments.

$$\text{topDepth}((\Delta, m, 0) : \rho) \stackrel{\text{def}}{=} m$$



Given these conventions, in Figure 2.20 we show the translation function *trE* which receives a *Core-Safe* expression with a compile-time environment  $\rho$ , and returns a list of SVM instructions and

$$\begin{aligned}
trE\ c\ \rho &= \text{BUILDENV } [c]; \\
&\quad \text{NormalForm } \rho \\
trE\ x\ \rho &= \text{BUILDENV } [\rho(x)]; \\
&\quad \text{NormalForm } \rho \\
trE\ (x\ @\ r)\ \rho &= \text{BUILDENV } [\rho(x), \rho(r)]; \\
&\quad \text{COPY}; \\
&\quad \text{NormalForm } \rho \\
trE\ (a_1\ \oplus\ a_2)\ \rho &= \text{BUILDENV } [\rho(a_1), \rho(a_2)]; \\
&\quad \text{PRIMOP } \oplus; \\
&\quad \text{NormalForm } \rho \\
trE\ (g\ \bar{a}_i^n\ @\ \bar{r}_j^m)\ \rho &= \text{BUILDENV } [\overline{\rho(a_i)^n}, \overline{\rho(r_j)^m}]; \\
&\quad \text{SLIDE } (n + m)\ (\text{topDepth}(\rho)); \\
&\quad \text{CALL } \mathbf{p} \\
\text{where } (g\ \bar{y}_i^n\ @\ \bar{r}_j^m = e_g) \in \Sigma & \\
\quad [\mathbf{p} \mapsto trE\ e_g\ [([\bar{r}_j^m \mapsto m - j + 1^m, \bar{y}_i^n \mapsto n - i + m + 1^n], n + m, 0)]] \in cs & \\
trE\ (C_l^m\ \bar{a}_i^n\ @\ r)\ \rho &= \text{BUILDCLS } C_l^m\ [\overline{\rho(a_i)^n}]\ (\rho(r)); \\
&\quad \text{NormalForm } \rho \\
trE\ (\text{let } x_1 = e_1\ \text{in } e_2)\ \rho &= \text{PUSHCONT } \mathbf{p}; \quad \& cs \uplus [\mathbf{p} \mapsto trE\ e_2\ (\rho + [x_1 \mapsto 1])] \\
&\quad trE\ e_1\ \rho^{++} \\
trE\ (\text{case } x\ \text{of } \bar{alt}_i^n)\ \rho &= \text{MATCH } (\rho(x))\ \bar{\mathbf{p}}_i^n \quad \& cs \uplus [\bar{\mathbf{p}}_i \mapsto trA\ \bar{alt}_i^n\ \rho^n] \\
trE\ (\text{case! } x\ \text{of } \bar{alt}_i^n)\ \rho &= \text{MATCH! } (\rho(x))\ \bar{\mathbf{p}}_i^n \quad \& cs \uplus [\bar{\mathbf{p}}_i \mapsto trA\ \bar{alt}_i^n\ \rho^n] \\
trA\ (C\ \bar{x}_i^n \rightarrow e_i)\ \rho &= trE\ e_i\ (\rho + [x_i \mapsto n - i + 1^n])
\end{aligned}$$

Figure 2.20: Translation schemes from *Core-Safe* expressions to SVM instructions

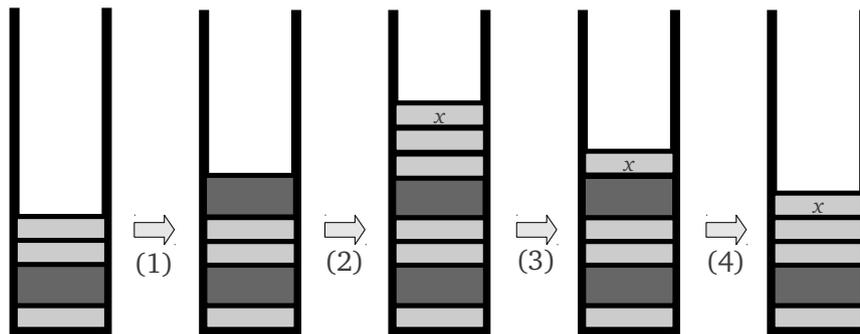


Figure 2.21: Evaluation of a **let** expression in the SVM. The white areas separate the different blocks, First, a continuation is inserted (1). Then, the evaluation of the auxiliary expression takes place until a normal form is reached and it is placed at the top of the stack (2). The previous environment is discarded (3) and the topmost continuation is removed (4), so the evaluation of the main expression begins, with the value of the bound variable at the top of the stack.

a code store. The latter is handled as an accumulator parameter of  $trE$  that is implicitly passed to every subexpressions being compiled. The notation  $\& cs \uplus [...]$  should be understood as the addition of the specified binding to the code store accumulated so far. The expression  $NormalForm \rho$  is a compilation macro defined as follows:

$$NormalForm \rho \stackrel{\text{def}}{=} \begin{array}{l} \text{SLIDE 1 } (topDepth(\rho)) \\ \text{DECREGION} \\ \text{POPCONT} \end{array}$$

This macro is used when the expression being translated gives place to a normal form (i.e. no additional evaluations are needed). This is the case of literal, variables, copy and constructor expressions, and also the case of primitive operators.

When evaluating a **let** expression (see Figure 2.21), a continuation is inserted before evaluating the auxiliary expression  $e_1$ . During translation, every binding added to the compile-time environment  $\rho$  will be added to a separate block, which is now the topmost one, until the translation of  $e_1$  finishes. This corresponds to the removal of topmost block and the continuation at runtime, so that the evaluation of the main expression  $e_2$  begins.

A interesting case is that of function application (Figure 2.22). Firstly the actual parameters are inserted into the stack. Since the execution of the function being called occur in a different context, the previous environment can be discarded before the evaluation of the body of the function. The removal of the previous environment allows us to obtain constant stack space for tail-recursive functions, as the following example shows.

**Example 2.17.** Let us consider the following function definitions:

$$\begin{aligned} sum \ xs \quad &= \mathbf{case} \ xs \ \mathbf{of} \\ & \quad [] \rightarrow 0 \\ & \quad (x : xx) \rightarrow \mathbf{let} \ x_1 = sum \ xx \ \mathbf{in} \ x + x_1 \end{aligned}$$

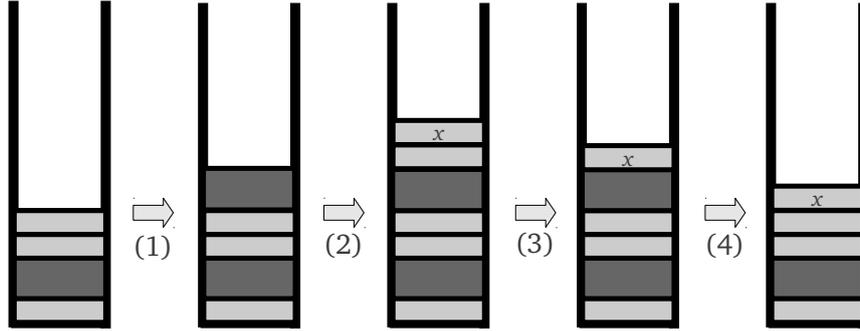


Figure 2.22: Evaluation of a function application in the SVM. The arguments of the function (dark-gray elements) are inserted at the top of the stack (1). Then, the previous environment, until the topmost continuation, is discarded in (2), so that, the only stack elements available during the evaluation of the body of the function are the arguments (corresponding to the formal parameters of the function, which are the only variables in scope before its evaluation).

```

sumAc xs ac = case xs of
  [] → ac
  (x : xx) → let x1 = x + ac in sumAc xx x1

```

Function *sum* adds the elements of a list. Function *sumAc* is its tail-recursive version with an accumulator parameter *ac*. The translation function *trE* applied to *sum* returns the following SVM code:

```

1 sum : MATCH 0 [p1, p2]           8 SLIDE 1 0
2 p1 : BUILDENV [Lit 0]           9 CALL sum
3 SLIDE 1 1                          10 p3 : BUILDENV [Pos 0, Pos 1]
4 DECREGION                          11 PRIMOP +
5 POPCONT                             12 SLIDE 1 4
6 p2 : PUSHCONT p3              13 DECREGION
7 BUILDENV [Pos 3]                    14 POPCONT

```

Let us assume we execute *sum* with the list [5,7] given as input. In other words, we execute the SVM code starting from the *sum* label, with an initial heap  $h = [p_1 \mapsto (1, (:) 5 p_2), p_2 \mapsto (1, (:) 7 p_3), p_3 \mapsto (1, [])]$ , and assuming  $p_1$  at the top of the evaluation stack. Figure 2.23 shows how the stack evolves during the execution of *sum*. The arrows between stacks contain the number of the SVM instruction being executed in each step (as labelled in the code above).

With respect to *sumAc*, the translation generates the following SVM code:

```

1 sumAc : MATCH 0 [p1, p2]       8 PRIMOP +
2 p1 : BUILDENV [Pos 1]          9 SLIDE 1 0
3 SLIDE 1 2                          10 DECREGION
4 DECREGION                          11 POPCONT
5 POPCONT                             12 p3 : BUILDENV [Pos 2, Pos 0]
6 p2 : PUSHCONT p3              13 SLIDE 2 5
7 BUILDENV [Pos 2, Pos 5]            14 CALL sumAc

```

Under the same conditions as above, Figure 2.24 shows how the stack changes as the execution of *sumAc*



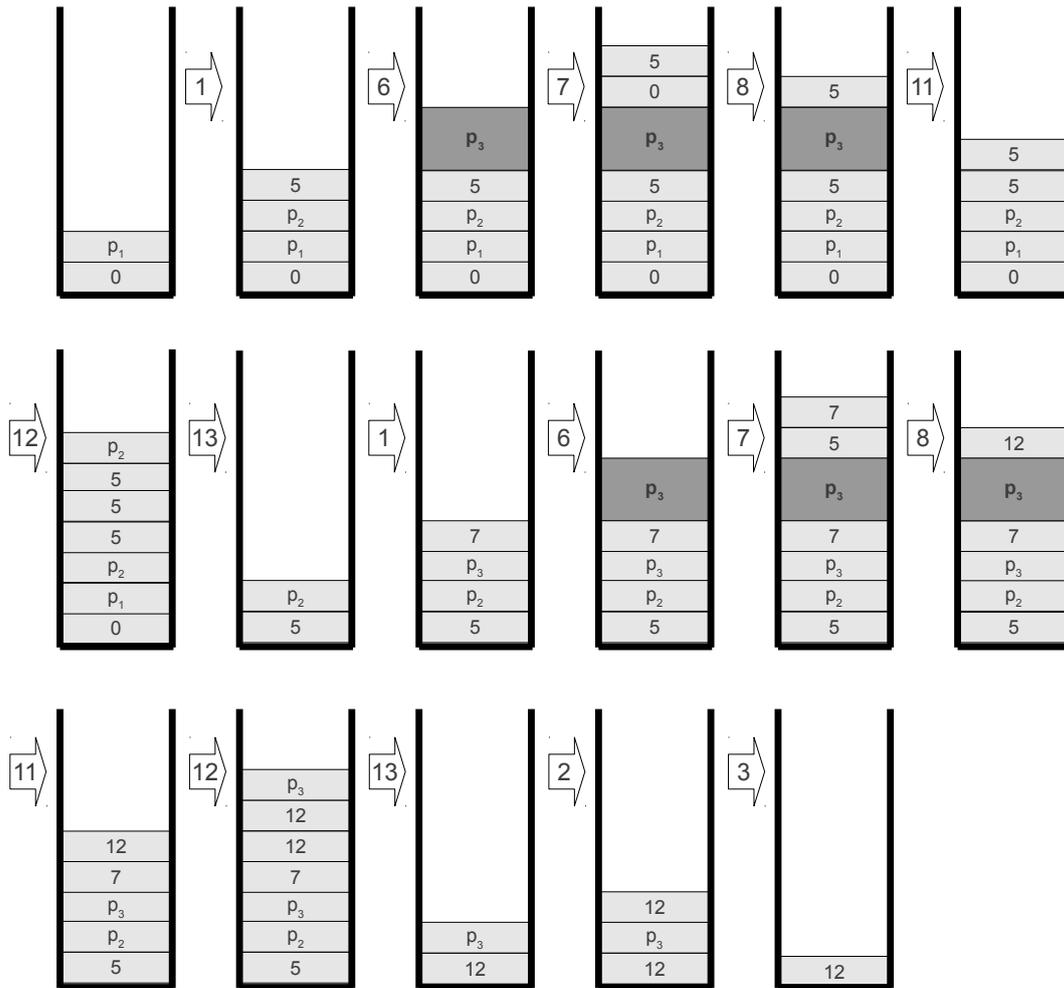


Figure 2.24: Contents of the stack during the execution of *sumAc*.

progresses. In tail-recursive functions there are no continuations left in the stack when doing a recursive call. This means we can discard the whole environment in the current function call before proceeding to the next recursive call. As a result, *sumAc* runs in constant stack space (6 words), while *sum* needs linear stack space ( $5n + 1$  words, where  $n$  is the number of elements in the input list).  $\square$

Although the bytecode of the SVM is not very close to that of the Java Virtual Machine (JVM), the compiler provides a further translation phase from SVM to JVM bytecode (Figure 2.25). This is described in [38]. We have two translation phases: from *Core-Safe* to the SVM (shown in this section), and from SVM to the JVM. It is worth mentioning that both phases have been formally certified [38, 39]. The proof of the latter phase consists in establishing a bijection between the heap of the SVM and the (more complex) heap of the JVM, which is preserved along the execution of both machines. This does not only certify the correctness of the translation itself, but also certifies that the memory consumption properties and the absence of dangling pointers are preserved in the JVM.

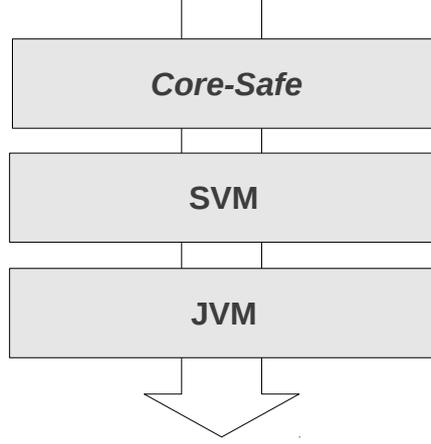


Figure 2.25: The SVM serves as an intermediate step between the *Core-Safe* code and the Java Virtual Machine (JVM). Both translations have been certified, and preserve memory costs. The translation from *Core-Safe* to SVM is dealt with in this chapter. The translation from SVM to JVM is described in [38].

$$\begin{array}{c}
\frac{}{E \vdash h, k, td, c \Downarrow h, k, c, ([ ]_k, 0, 1)} \text{ [Lit]} \\
\\
\frac{}{E[x \mapsto v] \vdash h, k, td, x \Downarrow h, k, v, ([ ]_k, 0, 1)} \text{ [Var]} \\
\\
\frac{}{E \vdash h, k, td, a_1 \oplus a_2 \Downarrow h, k, E(a_1) \oplus E(a_2), ([ ]_k, 0, 2)} \text{ [PrimOp]} \\
\\
\frac{j \leq k \quad (h', p') = \text{copy}(h, p, j) \quad m = \text{size}(h, p)}{E[x \mapsto p, r \mapsto j] \vdash h, k, td, x @ r \Downarrow h', k, p', ([j \mapsto m]_k, m, 2)} \text{ [Copy]} \\
\\
\frac{(g \bar{y}_i^n @ \bar{r}_j^l = e_g) \in \Sigma}{\frac{[y_i \mapsto E(a_i)^n, r_j^l \mapsto E(r_j)^l, \text{self} \mapsto k+1] \vdash h, k+1, n+l, e \Downarrow h', k+1, v, (\delta, m, s)}{E \vdash h, k, td, g \bar{a}_i^n @ \bar{r}_j^l \Downarrow h' |_k, k, v, (\delta |_k, m, \max\{n+l, s+n+l-td\})} \text{ [App]} \\
\\
\frac{E(r) = j \quad j \leq k \quad \text{fresh}_h(p)}{E \vdash h, k, td, C \bar{a}_i^n @ r \Downarrow h \uplus [p \mapsto (j, C E(a_i)^n)], k, v, ([j \mapsto 1]_k, 1, 1)} \text{ [Cons]} \\
\\
\frac{E \vdash h, k, 0, e_1 \Downarrow h', k, v_1, (\delta_1, m_1, s_1) \quad E \cup [x_1 \mapsto v_1] \vdash h', k, td+1, e_2 \Downarrow h'', k, v, (\delta_2, m_2, s_2)}{E \vdash h, k, td, \text{let } x_1 = e_1 \text{ in } e_2 \Downarrow h'', k, v, (\delta_1 + \delta_2, \max\{m_1, |\delta_1| + m_2\}, \max\{2 + s_1, 1 + s_2\})} \text{ [Let]} \\
\\
\frac{C = C_r \quad E(x) = p \quad E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, td + n_r, e_r \Downarrow h', k, v, (\delta, m, s)}{E \vdash h[p \mapsto (j, C \bar{v}_i^{nr}), k, td, \text{case } x \text{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i^n \Downarrow h', k, v, (\delta, m, s + n_r)} \text{ [Case]} \\
\\
\frac{C = C_r \quad E(x) = p \quad E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, td + n_r, e_r \Downarrow h', k, v, (\delta, m, s)}{E \vdash h \uplus [p \mapsto (j, C \bar{v}_i^{nr}), k, td, \text{case! } x \text{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i^n \Downarrow h', k, v, (\delta + [j \mapsto -1]_k, \max\{0, m-1\}, s + n_r)} \text{ [Case!]}
\end{array}$$

Figure 2.26: Resource-aware operational semantics of *Core-Safe* expressions.

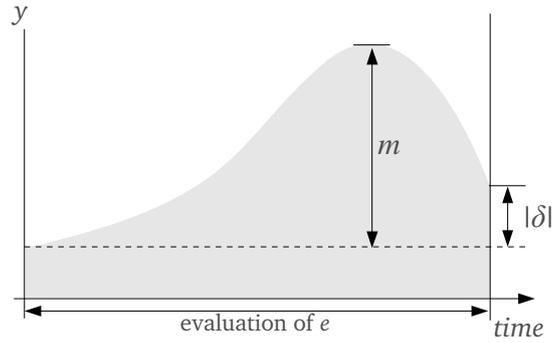


Figure 2.27: Intuitive meaning of  $\delta$  and  $m$  components in the resource vector. The  $y$  coordinate represents the number of cells in the heap.

## 2.7 Resource-aware semantics

Once the resource consumption of each instruction of the SVM is known, we enrich the big-step semantics given in Section 2.4.1 with a resource vector  $(\delta, m, s)$ , which can be conceived as a side effect of evaluating an expression  $e$ .

- The first component is a partial function  $\delta : \mathbb{N} \rightarrow \mathbb{Z}$  giving, for each region  $k$ , the signed difference between the number of cells after and before evaluating the expression. A positive difference means that more cells have been created than destroyed. A negative one describes the opposite situation.
- The component  $m$  is a natural number describing the *minimum* number of fresh cells in the heap needed to successfully evaluate  $e$ . This number corresponds to the maximum amount of cells existing simultaneously in memory during the evaluation of this expression.
- The component  $s$  is a natural number whose meaning is analogous to that of the  $m$  component. The  $s$  component describes the minimum size of stack (in words) needed for the evaluation of the expression.

Figure 2.27 gives an intuition on the meaning of the first two components. Assume the evaluation of an expression  $e$ . The figure represents the global amount of cells in memory as the evaluation of  $e$  proceeds. In this case, the evaluation of  $e$  reclaims memory until some time point, from which memory is disposed of. The  $m$  value represents the maximum amount of memory taken during the evaluation of  $e$ , whereas  $\delta$  represents the difference of memory amount between the initial and final heaps. Notice, however, that the  $\delta$  contains this difference *for every region* in the heap. What is represented in Figure 2.27 is the global balance  $|\delta|$  of heap cells between the final and initial heaps, which will be formally defined below. Also notice that both values  $m$  and  $\delta$  are relative to the memory consumption level at the beginning of the evaluation of  $e$  (dashed line in Figure 2.27).

If we represented the stack consumption in the style of Figure 2.27, the  $s$  component of the resource vector would take the role of the  $m$  component in the heap consumption. There is no component for denoting the difference in stack words between the final and initial heaps, because the final stack contains the same elements of the initial stack with an additional element at the top, which is the result of evaluating the expression, so this difference is always 1.

The domain of  $\delta$  is the set  $\{0..k\}$ , where  $k$  is the number of regions in the heap to which the  $\delta$  refers. The notation  $[ ]_k$  stands for the function  $[i \mapsto 0 \mid i \in \{0..k\}]$ , whereas  $[i \mapsto n]_k$  abbreviates the function

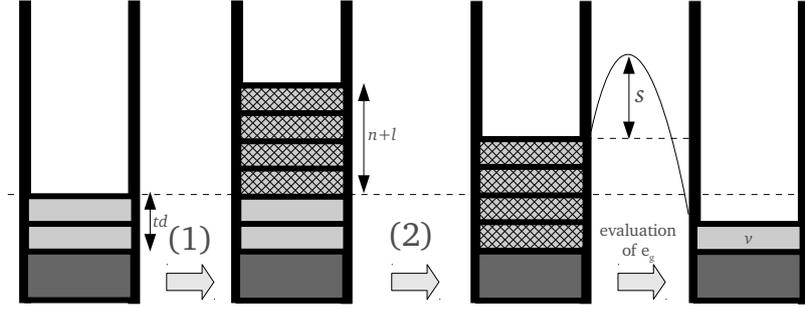


Figure 2.28: Stack consumption while evaluating a function application: we have to take the maximum between the number of arguments pushed onto the stack (1) and the maximum stack level reached during evaluation of the function's body, assuming that the arguments are already in the stack and the previous environment has been discarded (2).

$[i \mapsto n] \uplus [j \mapsto 0 \mid j \in \{0..k\} \setminus \{i\}]$ . The *total balance* of cells (denoted by  $|\delta|$ ) is the sum of the balances obtained in each region:

$$|\delta| \stackrel{\text{def}}{=} \sum_{i \in \text{dom } \delta} \delta(i)$$

The notation  $\delta_1 + \delta_2$  represents the component-wise addition of  $\delta_1$  and  $\delta_2$ , provided these have the same domain:

$$\delta_1 + \delta_2 \stackrel{\text{def}}{=} [\delta_1(i) + \delta_2(i) \mid i \in \text{dom } \delta_1 \cap \text{dom } \delta_2]$$

The enriched semantic rules are shown in Figure 2.26. In addition to the resource vector  $(\delta, m, s)$ , we need a new component  $td$  in order to simulate the *topDepth* function of compile time environments. This component represents the number of stack words inserted after the topmost continuation, and it influences the  $s$ , since an amount  $td$  of words are removed from the stack before function calls.

The evaluation of atom (rules  $[Lit]$  and  $[Var]$ ) does not require memory consumption, and it requires a stack word space to push the result into. The evaluation of a primitive operator application requires two stack words, since the operands have to be pushed into the stack before computing the result. The evaluation of a copy expression  $[Copy]$  requires as many heap cells as the size of the recursive spine of the structure being copied. The *size* function defines this notion of size:

$$\text{size}(h[p \mapsto (j, C \bar{v}_i^n)], p) \stackrel{\text{def}}{=} 1 + \sum_{i \in \text{RecPos}(C)} \text{size}(h, v_i)$$

In rule  $[App]$ , by  $\delta|_k$  we mean a function like  $\delta$  but restricted to the domain  $\{0..k\}$ . The computation  $\max\{n + l, s + n + l - td\}$  of fresh stack words takes into account that the first  $n + l$  words are needed to store the actual arguments, then the current environment of length  $td$  is discarded, and then the function body is evaluated (Figure 2.28).

In rule  $[Let]$ , a continuation (2 words) is stacked before evaluating  $e_1$ , and this leaves a value in the stack before evaluating  $e_2$ . That is why we obtain  $\max\{2 + s_1, 1 + s_2\}$  as stack consumption (see Figure 2.29). With regard to the heap consumption, we take the maximum between the memory needs of  $e_1$  and those of  $e_2$ , but taking the balance  $|\delta_1|$  into account (Figure 2.30).

**Example 2.18.** In Figure 2.31 we show the resource vector corresponding to the execution of most of the examples in this chapter. The concrete input DSs are shown above the table. We assume that these DSs have been created before evaluating each function call, so their building costs are not reflected in

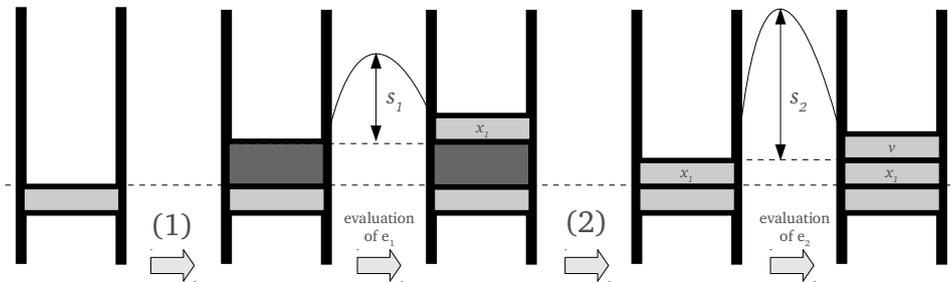


Figure 2.29: Stack consumption while evaluating a **let** expression: we have to take the maximum between the execution of  $e_1$  assuming that a continuation has been pushed before into the stack (1), and the execution of  $e_2$  assuming that the value of  $x_1$  has been pushed after discarding the continuation (2).

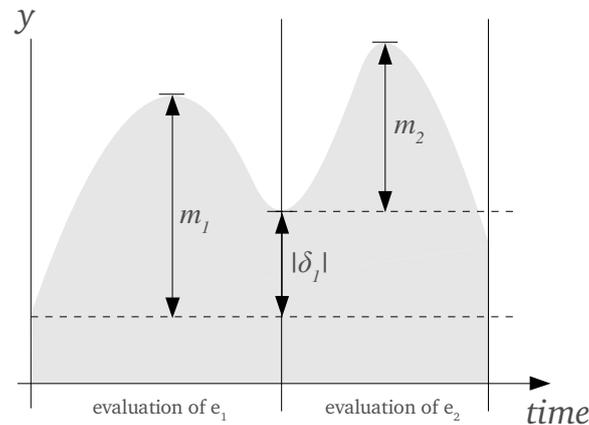


Figure 2.30: Heap consumption of two expressions executed in sequence (**let**). The execution of the first expression leaves  $|\delta_1|$  cells in the heap. We take this point as the reference level on which we measure the memory needs of the second expression  $m_2$ . Hence, the memory needs of the whole sequence is given by  $\sqcup\{m_1, |\delta_1| + m_2\}$ .

$xs = [1, 2, 3]$  (4 cells)     $ys = [4, 5]$  (3 cells)     $zs = [5, 4, 3, 2, 1]$  (6 cells)

$t = \text{Node} (\text{Node Empty } 2 \text{ Empty}) 4 (\text{Node Empty } 7 \text{ Empty})$  (7 cells)

Expression	$\delta$	$m$	$s$
$\text{append } xs \ ys \ @ \ r$	$[E(r) \mapsto 3]_k$	3	23
$\text{appendC } xs \ ys \ @ \ r$	$[E(r) \mapsto 6]_k$	6	24
$\text{appendD } xs \ ys \ @ \ r$	$[E(r) \mapsto -1]_k$	0	23
$\text{insert } 10 \ xs \ @ \ r$	$[E(r) \mapsto 5]_k$	5	32
$\text{insertD } 10 \ xs \ @ \ r$	$[E(r) \mapsto 1]_k$	1	32
$\text{inssort } zs \ @ \ r$	$[E(r) \mapsto 21]_k$	21	41
$\text{inssortD } zs \ @ \ r$	$[E(r) \mapsto 0]_k$	0	41
$\text{mkTree } zs \ @ \ r$	$[E(r) \mapsto 11]_k$	11	50
$\text{inorder } t \ @ \ r$	$[E(r) \mapsto 4]_k$	5	29
$\text{treesort } zs \ @ \ r$	$[E(r) \mapsto 6]_k$	18	62
$\text{sumAc } xs \ 0$	$[\ ]_k$	0	6
$\text{sumAc } zs \ 0$	$[\ ]_k$	0	6

Figure 2.31: Memory consumption results. The table shows, for each expression, the resource vector  $(\delta, m, s)$  resulting from its evaluation.

the table.

Function *append* creates as many cells in the output region as the number of cons cells in the list passed as first parameter. In *appendC* we need three additional cells for copying the list passed as second parameter. Function *appendD* destroys four cells of the input list and builds three in the output region. Its execution needs no additional heap space. The heap cost of *insert* and *inssort* is proportional to the length of the input list, whereas in *insertD* and *inssortD* this cost is constant. In the case of *insertD* we need an additional cell for storing the new element. Function *treesort* leaves in the output region as many cells as the input list. However, more cells are needed in order to build the intermediate tree. Finally, the calls to *sumAc* produce no heap costs and a constant stack cost.  $\square$

## 2.8 Correctness of the translation into SVM

Now we show that the pair translation-abstract machine is sound and complete with respect the semantics defined in last section. The correctness of the translation involves the proof of the following facts:

1. Both big-step semantics and SVM lead to the same result and final heap, whenever their initial configurations are *equivalent*.
2. The  $(\delta, m, s)$  vector of the semantics models the actual memory needs of an expression when being evaluated in the SVM.

The first fact proves that the translation from *Core-Safe* to SVM preserves the *semantics* of a given program, whereas the second one shows the preservation of its *memory needs*. In order to prove these results, we must, on the one hand, make precise the idea of a semantic configuration being equivalent to a SVM configuration. On the other hand, we have to determine the memory costs of a program when being run in the SVM. First, we note that both the semantics and the SVM machine rules are syntax driven, and that their computations are deterministic (up to fresh names generation).

**Lemma 2.19.** *Given a value environment  $E$ , an initial heap  $h$  with  $k$  regions, a natural number  $td$ , and a Core-Safe expression  $e$ , if  $E \vdash h, k, td, e \Downarrow h, k, v, (\delta, m, s)$  and  $E \vdash h, k, td, e \Downarrow h', k, v', (\delta', m', s)$ , then  $h = h'$  (up to pointer renaming),  $v = v'$ , and  $(\delta, m, s) = (\delta', m', s')$*

*Proof.* By induction on the  $\Downarrow$ -derivation. All cases are straightforward.  $\square$

**Lemma 2.20.** *Given an initial configuration  $c_{init}$  such that  $c_{init} \rightarrow c_1$  and  $c_{init} \rightarrow c'_1$ . Then  $c_1 = c'_1$  (modulo pointer renaming).*

*Proof.* By case distinction on the rule being applied. All cases are straightforward.  $\square$

The main difference between the big-step operational semantics of Section 2.4.1 and the SVM machine is the way in which value environments are represented. In the big-step semantics we have a mapping  $E$  from variables to values, whereas in the SVM we have a stack. The correspondence between variables and positions of the stack is given by the  $\rho$  environment used in the translation (see Section 2.6).

Obviously, we cannot ensure the equivalence of big-step semantics and SVM if their starting point (i.e. mapping  $E$  in big-step semantics, stack  $S$  in the SVM) denote different value environments. This is the motivation for the following:

**Definition 2.21.** We say that the environment  $E$  and the pair  $(\rho, S)$  are equivalent, denoted  $E \equiv (\rho, S)$ , if  $\text{dom } E = \text{dom } \rho$ , and  $\forall x \in \text{dom } \rho \setminus \{\text{self}\}. E(x) = S!(\rho(x))$ .

A stack  $S'$  is said to be a *suffix* of  $S$  if there exists a number  $n \geq 0$  of stack elements  $b_i$  such that  $S = \bar{b}_i^n : S'$ . Given a SVM configuration  $c_0 = (is, h, k_0, k, S, cs)$  and  $S'$  a suffix of  $S$ , we denote by  $c_0 \rightarrow_{S'}^* c_n$  a derivation in which all the stacks in intermediate configurations have  $S'$  as a suffix. Should the topmost instruction of a configuration create a stack smaller than  $S'$ , then the machine would stop at that configuration.

Now we follow with some definitions regarding the maximum memory consumption produced by a SVM program.

**Definition 2.22.** The function  $sizeST$ , which returns the size (in words) of a stack, is defined as follows:

$$\begin{aligned} sizeST([]) &= 0 \\ sizeST(v : S) &= 1 + sizeST(S) \text{ if } v \text{ is not a continuation.} \\ sizeST((k_0, \mathbf{p}) : S) &= 2 + sizeST(S) \end{aligned}$$

Analogously, the size (in cells) of a heap is given by the function  $sizeH$ :

$$sizeH(h) = |\text{dom } h|$$

The reason for having  $2 + sizeST(S)$  in the case where a continuation is at the top of the stack is that any sensible implementation of the SVM would need two words for storing the pair  $(k_0, \mathbf{p})$ .

We shall extend the notation of these functions to configurations, so  $sizeST(c)$  (resp.  $sizeH(c)$ ) will be used to denote the size of the stack (resp. heap) of the given configuration  $c$ .

Given  $c_0 = (is, h, k_0, k, S, cs)$  and  $c_0 \rightarrow_{S'} \dots \rightarrow_{S'} c_n$  a SVM derivation, the *maximum number of fresh cells* of the derivation (denoted  $maxFreshCells(c_0 \rightarrow_{S'}^* c_n)$ ) is the highest difference in cells between the

heaps of the configurations  $c_0, \dots, c_n$  and the initial heap  $h$ . Likewise, we define the *maximum number of fresh words* created in the stack  $S$ , denoted  $\text{maxFreshWords}(c_0 \rightarrow_{S'}^* c_n)$ . Finally, by  $\text{diff}(k, h, h')$  we denote a function giving for each region in  $\{0, \dots, k\}$  the signed difference in cells between  $h'$  and  $h$ .

**Definition 2.23.** Given a derivation  $c_0 \rightarrow_{S'} \dots c_i \dots \rightarrow_{S'} c_n$ , we define:

$$\begin{aligned} \text{maxFreshCells}(c_0 \rightarrow_{S'}^* c_n) &= \max \{ \text{sizeH}(c_i) - \text{sizeH}(c_0) \mid 0 \leq i \leq n \} \\ \text{maxFreshWords}(c_0 \rightarrow_{S'}^* c_n) &= \max \{ \text{sizeST}(c_i) - \text{sizeST}(c_0) \mid 0 \leq i \leq n \} \end{aligned}$$

From these definitions the following properties can be easily obtained:

$$\begin{aligned} \text{maxFreshCells}(c_0 \rightarrow_{S'}^* c_i \rightarrow_{S'}^* c_n) &= \max \{ \text{maxFreshCells}(c_0 \rightarrow_{S'}^* c_i), \\ &\quad \text{maxFreshCells}(c_i \rightarrow_{S'}^* c_n) \\ &\quad + \text{sizeH}(c_i) - \text{sizeH}(c_0) \} \end{aligned}$$

$$\begin{aligned} \text{maxFreshWords}(c_0 \rightarrow_{S'}^* c_i \rightarrow_{S'}^* c_n) &= \max \{ \text{maxFreshWords}(c_0 \rightarrow_{S'}^* c_i), \\ &\quad \text{maxFreshWords}(c_i \rightarrow_{S'}^* c_n) \\ &\quad + \text{sizeST}(c_i) - \text{sizeST}(c_0) \} \end{aligned}$$

**Definition 2.24.** Given two heaps  $h$  and  $h'$  and a number  $k$  of regions, we denote by  $\text{diff}(k, h, h')$  a function  $\delta$  such that  $\text{dom } \delta = \{0..k\}$  and:

$$\forall i \in \{0..k\} . \delta(i) = |\{p \in \text{dom } h' \mid \text{region}(h'(p)) = i\}| - |\{p \in \text{dom } h \mid \text{region}(h(p)) = i\}|$$

The definitions of  $\text{diff}$  and  $\text{sizeH}$  are related by the following property, which is easy to establish from the corresponding definitions. Let  $h$  and  $h'$  be two heaps with  $k$  regions. Then:

$$\text{sizeH}(h') - \text{sizeH}(h) = \sum_{i=0}^k \text{diff}(i, h, h')$$

The next Lemma shows some properties of the *copy* function. In particular, that the size of the copy and the DS being copied are the same.

**Lemma 2.25.** If  $(h', p') = \text{copy}(h, p, j)$  then:

1.  $h \subseteq h'$
2.  $\text{size}(h', p') = \text{size}(h, p)$
3.  $\text{sizeH}(h') - \text{sizeH}(h) = \text{size}(h', p')$
4.  $\forall k \geq j, \text{diff}(k, h, h') = [j \mapsto \text{size}(h', p')]_k$

*Proof.* Straightforward induction on the size of the DS pointed to by  $p$ . □

Now we prove that, if we translate an expression  $e$  into a sequence  $is$  of SVM instructions and the SVM executes this sequence, then the number of regions in the heap of the final configuration is equal to the  $k_0$  component of the initial configuration.

**Lemma 2.26.** For all  $S, S', h, h', td, k_0, k, e, v, \rho, cs, cs'$  of their respective types, such that

$$\begin{aligned} td &= \text{topDepth}(\rho) & (is, cs) &= \text{trE } e \rho \\ S' &= \text{drop } td \ S & cs' &\supseteq cs \\ k_0 &\leq k \end{aligned}$$

If  $\underbrace{(is, h, k_0, k, S, cs')}_{c_{init}} \rightarrow_{S'}^* \underbrace{([\text{POPCONT}], h', k', k', v : S', cs')}_{c_{final}}$  then  $k' = k_0$ .

*Proof.* This can be trivially proved by induction on the length of  $\rightarrow_{S'}^*$ , and by cases over the expression  $e$  whose translation is executed.  $\square$

The following theorem establishes the correctness of the translation, showing that the computed value and the resource consumption of a given expression  $e$  in the big-step semantics are the same as those obtained by executing in the SVM the translation of  $e$  (see Appendix C).

We denote by  $\text{drop } n \ S$  the stack resulting from removing the  $n$  topmost elements of  $S$ . That is,  $\text{drop } n \ (\bar{b}_i^n : S) = S$  and undefined in case the number of elements in the input stack is less than  $n$ .

**Theorem 2.27.** For all  $S, S', E, h, h', td, k_0, k, e, v, \delta, m, s, \rho, cs, cs'$  of their respective types, if

$$\begin{array}{llll} E \equiv (\rho, S) & (is, cs) = \text{tr}E \ e \ \rho & td = \text{topDepth}(\rho) & E \text{ admissible w.r.t. } k \\ S' = \text{drop } td \ S & cs' \supseteq cs & k_0 \leq k & \end{array}$$

then  $E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$  if and only if

1.  $\underbrace{(is, h, k_0, k, S, cs')}_{c_{init}} \rightarrow_{S'}^* \underbrace{([\text{POPCONT}], h' |_{k_0}, k_0, k_0, v : S', cs')}_{c_{final}}$
2.  $\delta = \text{diff}(k, h, h')$
3.  $m = \text{maxFreshCells}(c_{init} \rightarrow_{S'}^* c_{final})$
4.  $s = \text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_{final})$

*Proof.* The  $(\Rightarrow)$  direction is shown by induction on the  $\Downarrow$ -derivation. The  $(\Leftarrow)$  direction of the theorem can be proved by induction on the length of the  $\rightarrow_{S'}^*$  derivation. In both proofs we distinguish cases depending on the expression being translated.

- **Case [Lit]:**  $e \equiv c$

Let  $(is, cs) = \text{tr}E \ e \ \rho$  be the SVM code generated, where  $cs = []$  and:

$$is = \text{BUILDENV } [c] : \overbrace{\text{SLIDE } 1 \ (\text{topDepth}(\rho)) : \text{DECREGION} : [\text{POPCONT}]}^{is_1}$$

$$\underbrace{\hspace{15em}}_{is_2}$$

$$\underbrace{\hspace{10em}}_{is_3}$$

In addition, by the rule [Lit] we have  $E \vdash h, k, td, c \Downarrow h, k, c, ([ ]_k, 0, 1)$ . For all  $cs'$  we get the following derivation:

$$\begin{array}{l} \overbrace{(\text{BUILDENV } [c] : is_1, h, k_0, k, S, cs')}^{c_{init}} \\ \rightarrow_{S'} \\ (\text{SLIDE } 1 \ (\text{topDepth}(\rho)) : is_2, h, k_0, k, c : S, cs') \\ \rightarrow_{S'} \quad \{\text{since } td = \text{topDepth}(\rho)\} \\ (\text{DECREGION} : is_3, h, k_0, k, c : \text{drop } td \ S, cs') \\ \rightarrow_{S'} \end{array}$$

$$\underbrace{([\text{POPCONT}], h|_{k_0}, k_0, k_0, c : \text{drop td } S, cs')}_{c_{\text{final}}}$$

Because of the assumption  $S' = \text{drop td } S$  the property (1) holds. Moreover, we get  $\text{diff}(k, h, h) = [i \mapsto 0 \mid 0 \leq i \leq k] = \delta$ , so property (2) holds. Finally, both properties (3) and (4) hold because  $\text{maxFreshCells}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 0$  and  $\text{maxFreshWords}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 1$ .

- **Case [Var]:**  $e \equiv x$

The proof is similar to the case  $e \equiv c$ . Now  $E[x \mapsto v] \vdash h, k, \text{td}, x \Downarrow h, k, v, ([ ]_k, 0, 1)$  holds and the code generated is as follows:

$$is = \text{BUILDENV } [\rho(x)] : \overbrace{\text{SLIDE 1 } (topDepth(\rho)) : \text{DECREGION : } [\text{POPCONT}]}^{is_1}$$

$$\underbrace{\hspace{15em}}_{is_2}$$

$$\underbrace{\hspace{10em}}_{is_3}$$

From which the following derivation is obtained:

$$\begin{array}{l} \overbrace{(\text{BUILDENV } [\rho(x)] : is_1, h, k_0, k, S, cs')}^{c_{\text{init}}} \\ \rightarrow_{S'} \quad \{\text{since } S!(\rho(x)) = E(x) = v\} \\ (\text{SLIDE 1 } (topDepth(\rho)) : is_2, h, k_0, k, v : S, cs') \\ \rightarrow_{S'} \quad \{\text{since } \text{td} = topDepth(\rho)\} \\ (\text{DECREGION : } is_3, h, k_0, k, v : \text{drop td } S, cs') \\ \rightarrow_{S'} \\ \underbrace{([\text{POPCONT}], h|_{k_0}, k_0, k_0, v : \text{drop td } S, cs')}_{c_{\text{final}}} \end{array}$$

Hence property (1) holds. The proof of (2), (3) and (4) is analogous to the case  $e \equiv c$ .

- **Case [Copy]:**  $e \equiv x @ r$

On the one hand, by using the [Copy] rule we know that:

$$E \vdash h, k, \text{td}, x @ r \Downarrow h', k, p', ([j \mapsto \text{size}(h, p)]_k, \text{size}(h, p), 2)$$

where  $E(x) = p, E(r) = j \leq k$  and  $(h', p') = \text{copy}(h, p, j)$ . On the other hand, we obtain  $(is, [ ]) = \text{tr}E e \rho$  where:

$$is = \text{BUILDENV } [\rho(x), \rho(r)] : \overbrace{\text{COPY : } \overbrace{\text{SLIDE 1 } (topDepth(\rho)) : \text{DECREGION : } [\text{POPCONT}]}^{is_2}}^{is_1}}$$

$$\underbrace{\hspace{15em}}_{is_3}$$

$$\underbrace{\hspace{10em}}_{is_4}$$

For any  $cs'$ , the first step of the  $\rightarrow_{S'}^*$  derivation is as follows:

$$\begin{array}{l}
\overbrace{(\text{BUILDENV } [\rho(x), \rho(r)] : is_1, h, k_0, k, S, cs')}^{c_{init}} \\
\rightarrow_{S'} \quad \{ \text{since } E(x) = p = S!(\rho(x)) \} \\
(\text{COPY} : is_2, h, k_0, k, p : \text{Item}_k(\rho(r)) : S, cs')
\end{array}$$

Let us proceed by case distinction: on one hand, if  $r \neq self$  then  $\text{Item}_k(\rho(r)) = S!(\rho(r)) = E(r) = j$ . On the other hand, if  $r = self$  then we have  $\text{Item}_k(\rho(r)) = \text{Item}_k(self) = k$ , but since  $E(self) = k$  (by Proposition 2.15) and  $E(self) = j$  (by rule [Copy]), we have that  $\text{Item}_k(\rho(r)) = j$ . Hence the current machine configuration can be rewritten as  $(\text{COPY} : is_2, h, k_0, k, p : j : S, cs')$ .

We shall now resume the  $\rightarrow_{S'}^*$  derivation:

$$\begin{array}{l}
(\text{COPY} : is_2, h, k_0, k, p : j : S, cs') \\
\rightarrow_{S'} \quad \{ \text{since } (h', p') = \text{copy}(h, p, j) \text{ and } j \leq k \} \\
(\text{SLIDE 1 } (topDepth(\rho)) : is_3, h', k_0, k, p' : S, cs') \\
\rightarrow_{S'} \quad \{ \text{since } td = topDepth(\rho) \} \\
(\text{DECREGION} : is_4, h', k_0, k, p' : drop td S, cs') \\
\rightarrow_{S'} \\
\underbrace{([\text{POPCONT}], h'|_{k_0}, k_0, k_0, p' : drop td S, cs')}_{c_{final}}
\end{array}$$

Therefore (1) holds, because of the assumption  $S' = drop td S$ . Properties (2) and (3) follow from Lemma 2.25. With respect to (4), from the resulting  $\rightarrow_{S'}^*$  derivation it can be easily shown that  $\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_{final}) = 2$ .

- **Case [PrimOp]:**  $e \equiv a_1 \oplus a_2$

The translation yields the following instruction sequence:

$$is = \overbrace{\text{BUILDENV } [\rho(a_1), \rho(a_2)] : \text{PRIMOP } \oplus : \text{SLIDE 1 } (topDepth(\rho)) : \text{DECREGION} : \underbrace{[\text{POPCONT}]}_{is_4}}^{is_1} : \underbrace{\hspace{10em}}_{is_3}$$

By executing it, we obtain the following derivation:

$$\begin{array}{l}
\overbrace{(\text{BUILDENV } [\rho(a_1), \rho(a_2)] : is_1, h, k_0, k, S, cs')}^{c_{init}} \\
\rightarrow_{S'} \quad \{ \text{since } S!(\rho(a_i)) = E(a_i) \text{ for every } i \in \{1..2\}, \text{ see below } \} \\
(\text{PRIMOP } \oplus : is_2, h, k_0, k, E(a_1) : E(a_2) : S, cs') \\
\rightarrow_{S'} \\
(\text{SLIDE 1 } (topDepth(\rho)) : is_3, h, k_0, k, (E(a_1) \oplus E(a_2)) : S, cs') \\
\rightarrow_{S'} \\
(\text{DECREGION} : is_4, h, k_0, k, (E(a_1) \oplus E(a_2)) : drop td S, cs') \\
\rightarrow_{S'}
\end{array}$$

$$([\text{POPCONT}], h|_{k_0}, k_0, k_0, (E(a_1) \oplus E(a_2)) : \text{drop td } S, cs') \equiv c_{\text{final}}$$

Hence, property 1 holds. From the initial and final configurations we obtain  $\text{diff}(k, h, h) = [i \mapsto 0 \mid 0 \leq i \leq k] = \delta$ , so property (2) holds. Properties (3) and (4) hold because  $\text{maxFreshCells}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 0$  and  $\text{maxFreshWords}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 2$ .

- **Case [Cons]:**  $e \equiv C \bar{a}_i^n @ r$

In the big-step semantics we get  $E \vdash h, k, e \Downarrow h \uplus [p \mapsto (E(r), C \overline{E(a_i)^n})], k, p, ([E(r) \mapsto 1]_k, 1, 1)$  where  $p$  is a fresh pointer. The translation yields the following instruction sequence:

$$is = \text{BUILDCLS } C \overline{[\rho(a_i)^n]} \rho(r) : \overbrace{\text{SLIDE 1 } (\text{topDepth}(\rho)) : \text{DECREGION} : [\text{POPCONT}]}^{is_1} \underbrace{\hspace{10em}}_{is_2} \underbrace{\hspace{2em}}_{is_3}$$

By executing this, we obtain the following derivation:

$$\begin{aligned} & \overbrace{(\text{BUILDCLS } C \overline{[\rho(a_i)^n]} \rho(r) : is_1, h, k_0, k, S, cs')}^{c_{\text{init}}} \\ \rightarrow_{S'} & \quad \{\text{since } S!(\rho(a_i)) = E(a_i) \text{ for every } i \in \{1..n\}, \text{ see below } \} \\ & (\text{SLIDE 1 } (\text{topDepth}(\rho)) : is_2, h \uplus [p \mapsto (E(r), C \overline{E(a_i)^n})], k_0, k, p : S, cs') \\ \rightarrow_{S'} & \quad \{\text{since } td = \text{topDepth}(\rho) \} \\ & (\text{DECREGION} : is_3, h \uplus [p \mapsto (E(r), C \overline{E(a_i)^n})], k_0, k, p : \text{drop td } S, cs') \\ \rightarrow_{S'} & \quad \overbrace{([\text{POPCONT}], (h \uplus [p \mapsto (E(r), C \overline{E(a_i)^n})])|_{k_0}, k_0, k_0, p : \text{drop td } S, cs')}^{c_{\text{final}}} \end{aligned}$$

In the first step we have assumed that  $\text{Item}_k(\rho(r)) = E(r)$ . The proof of this is similar to that seen in the [Copy] rule. Thus (1) holds. From the initial and final configurations we obtain:

$$\text{diff}(k, h, h \uplus [p \mapsto (E(r), C \overline{E(a_i)^n})]) = [E(r) \mapsto 1]_k$$

$$\text{maxFreshCells}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 1$$

$$\text{maxStackWords}(c_{\text{init}} \rightarrow_{S'}^* c_{\text{final}}) = 1$$

from which (2), (3) and (4) follow trivially.

- **Case [App]:**  $e \equiv g \bar{a}_i^n @ \bar{r}_j^l$

We assume, by rule [App] that  $E_g \vdash h, k+1, n+l, e_g \Downarrow h', k+1, v, (\delta, m, s)$ , where  $E_g$  is defined as follows:

$$E_g = \overline{[y_i \mapsto E(a_i)^n, r'_j \mapsto E(r_j)^l, self \mapsto k+1]}$$

and  $(g \bar{y}_i^n @ \bar{r}_j^l = e_g) \in \Sigma$ . Furthermore, let  $(is_g, cs_g)$  be the result of  $\text{trE } e_g \rho_g$ , where  $e_g$  is the body of the function to be called and  $\rho_g$  is defined as follows:

$$\rho_g = \overline{[[r'_j \mapsto (l-j)+1, y_i \mapsto (n-i)+(l+1)^n], n+l, 0]}$$

Assume that  $cs(\mathbf{p}) = is_g$ . We obtain  $(is, cs) = trE e \rho$ , where:

$$is = \text{BUILDENV } [\overline{\rho(a_i)^n}, \overline{\rho(r_j)^l}] : \overbrace{\text{SLIDE } (n+l) \text{ (topDepth}(\rho))}^{is_1} : \underbrace{[\text{CALL } \mathbf{p}]}_{is_2}$$

For any  $cs'$  such that  $cs' \supseteq cs$ , the code generated leads to the following SVM derivation:

$$\begin{aligned} & \overbrace{(\text{BUILDENV } [\overline{\rho(a_i)^n}, \overline{\rho(r_j)^l}]) : is_1, h, k_0, k, S, cs'}^{c_{init}} \\ \rightarrow_{S'} & (\text{SLIDE } (n+l) \text{ (topDepth}(\rho)) : is_2, h, k_0, k, \overline{S!(\rho(a_i))^n} : \overline{S!(\rho(r_j))^l} : S, cs') \\ \rightarrow_{S'} & \{ \text{since } td = \text{topDepth}(\rho) \} \\ & ([\text{CALL } \mathbf{p}], h, k_0, k, \overline{S!(\rho a_i)^n} : \overline{S!(\rho r'_j)^l} : \text{drop } td \ S, cs') \\ \rightarrow_{S'} & \{ \text{since } cs'(\mathbf{p}) = cs(\mathbf{p}) = is_g \} \\ & (is_g, h, k_0, k+1, \overline{S!(\rho a_i)^n} : \overline{S!(\rho r'_j)^l} : \text{drop } td \ S, cs') \equiv c_f \\ \rightarrow_{S'}^* & \{ \text{by i.h. (see below) and } k_0 \leq k \} \\ & \underbrace{([\text{POPCONT}], h'|_{k_0}, k_0, k_0, v : \text{drop } td \ S, cs')}_{c_{final}} \end{aligned}$$

where  $S' = \text{drop } td \ S$ . In order to apply the induction hypothesis in the last step, we prove that every assumption of the theorem holds for the  $\Downarrow$  derivation corresponding to  $e_f$ . The only nontrivial assumption to prove is that  $E_g \equiv (\rho_g, S_g)$ , denoting by  $S_g$  the result from pushing the actual parameters  $\overline{S!(\rho a_i)^n}$  and  $\overline{S!(\rho r'_j)^l}$  into  $S$ . Firstly, let  $x \in \text{dom } \rho_g$ . We prove that  $S_g!(\rho_g(x)) = E_g(x)$ . On the one hand, if  $x = y_i$  for some  $i \in \{1..n\}$ , we get  $\rho_g(x) = n+l - [(n-i) + (l+1)] = i-1$ . Hence:

$$S_g!(\rho_g(x)) = S_g!(i-1) = S!(\rho(a_i)) = E(a_i) = E_g(y_i)$$

On the other hand, if  $x = r'_j$  for some  $j \in \{1..l\}$  then we get  $\rho_g(x) = n+l - [(l-j) + 1] = n+j-1$  and:

$$S_g!(\rho_g(x)) = S_g!(n+j-1) = S!(\rho(r_j)) = E(r_j) = E_g(r'_j)$$

So  $E_g \equiv (\rho_g, S_g)$ , and the induction hypothesis can be applied on the derivation of  $e_g$ , which proves property (1), as we have shown above. The proof for (2) can be easily established from the fact that  $\text{diff}(k, h, h') = \text{diff}(k+1, h, h')|_k$  and that  $\delta = \text{diff}(k+1, h, h')$  by induction hypothesis. Property (3) also follows trivially from the induction hypothesis. With respect to (4), let us denote:

$$\begin{aligned} S_1 &= \overline{S!(\rho(a_i))^n} : \overline{S!(\rho(r_j))^l} : S \\ S_2 &= \overline{S!(\rho(a_i))^n} : \overline{S!(\rho(r_j))^l} : \text{drop } td \ S \\ S_3 &= v : \text{drop } td \ S \end{aligned}$$

$$\begin{aligned} \text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_f) &= \max\{\text{sizeST}(S_1) - \text{sizeST}(S), \\ & \quad \text{sizeST}(S_2) - \text{sizeST}(S)\} \\ &= \{n+l, n+l-td\} \end{aligned}$$

$$= n + l$$

On the other hand,  $\text{maxFreshWords}(c_f \rightarrow_{S'}^* c_{final}) = s$  by induction hypothesis. Therefore:

$$\begin{aligned} \text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_f \rightarrow_{S'}^* c_{final}) &= \max\{\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_f), \\ &\quad \text{maxFreshWords}(c_f \rightarrow_{S'}^* c_{final}) \\ &\quad + \text{sizeST}(S_2) - \text{sizeST}(S)\} \\ &= \max\{n + l, s + n + l - td\} \end{aligned}$$

- **Case [Let]:**  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$

By rule [Let] we get:

$$\begin{aligned} E \vdash h, k, 0, e_1 \Downarrow h', k, v_1, (\delta_1, m_1, s_1) \\ E_1 \vdash h', k, td + 1, e_2 \Downarrow h'', k, v, (\delta_2, m_2, s_2) \end{aligned}$$

Where we define  $E_1 = E \uplus [x_1 \mapsto v_1]$ . In addition, we assume  $(is_1, cs_1) = \text{tr}E e_1 \rho^{++}$  and  $(is_2, cs_2) = \text{tr}E e_2 \rho_1$  where  $\rho_1 = \rho + [x_1 \mapsto 1]$ . On the other hand, we get  $cs \supseteq cs_1$ ,  $cs \supseteq cs_2$  and  $is = \text{PUSHCONT } \mathbf{p} : is_1$ , in which  $cs_1(\mathbf{p}) = is_2$ . Given  $cs' \supseteq cs$ , the trace corresponding to the execution of  $e$  starts as follows:

$$\begin{aligned} &\overbrace{(\text{PUSHCONT } \mathbf{p} : is_1, h, k_0, k, S, cs')}^{c_{init}} \\ \rightarrow_{S'} & (is_1, h, k, k, (k_0, \mathbf{p}) : S, cs') \equiv c_1 \end{aligned}$$

In order to apply induction hypothesis on  $e_1$  we have to check that  $E \equiv (\rho^{++}, (k_0, \mathbf{p}) : S)$ , which trivially holds from the hypothesis  $E \equiv (\rho, S)$  and because  $\text{dom } \rho = \text{dom } \rho^{++}$ . The rest of the assumptions hold trivially.

In this case  $td = 0$ , so by applying i.h. we obtain:

$$\begin{aligned} &c_1 \\ \rightarrow_{(k_0, \mathbf{p}) : S}^* & ([\text{POPCONT}], h'|_k, k, k, v_1 : \text{drop } 0 ((k_0, \mathbf{p}) : S), cs'[\mathbf{p} \mapsto is_2]) \equiv c_2 \end{aligned}$$

which trivially implies  $c_1 \rightarrow_{S'}^* c_2$ .

Consequently:

$$\begin{aligned} &\overbrace{(\text{PUSHCONT } \mathbf{p} : is_1, h, k_0, k, S, cs')}^{c_{init}} \\ \rightarrow_{S'} & (is_1, h, k, k, (k_0, \mathbf{p}) : S, cs') \\ \rightarrow_{S'}^* & ([\text{POPCONT}], h'|_k, k, k, v_1 : \text{drop } 0 ((k_0, \mathbf{p}) : S), cs'[\mathbf{p} \mapsto is_2]) \equiv c_2 \\ \rightarrow_{S'} & (is_2, h'|_k, k_0, k, v_1 : S, cs') \equiv c_3 \end{aligned}$$

$$\begin{aligned}
& \rightarrow_{S'}^* \quad \{ \text{by i.h. (see below)} \} \\
& \quad ([\text{POPCONT}], h''|_{k_0}, k_0, k_0, v : \text{drop}(td+1)(v_1 : S), cs') \\
& \equiv \\
& \quad \underbrace{([\text{POPCONT}], h''|_{k_0}, k_0, k_0, v : S', cs')}_{c_{final}}
\end{aligned}$$

In order to apply the induction hypothesis on  $e_2$  in the last step we have to check that  $E_1 \equiv (\rho_1, v_1 : S)$ . Let  $x \in \text{dom}(E_1)$ : if  $x = x_1$  then  $(v_1 : S)!(\rho_1(x)) = (v_1 : S)!0 = v_1 = E_1(x)$ . If  $x \neq x_1$  then  $(v_1 : S)!(\rho_1(x)) = (v_1 : S)!(\rho(x) + 1) = S!(\rho(x)) = E(x) = E_1(x)$ . Additionally  $\text{drop}(td+1)(v_1 : S) = \text{drop } td \ S = S'$ . The rest of the assumptions hold trivially.

Hence (1) holds. With respect to (2), let  $i \in \{1..k\}$

$$\begin{aligned}
(\text{diff}(k, h, h''))(i) &= |\{p \in h'' \mid \text{region}(h''(p)) = i\}| - |\{p \in h \mid \text{region}(h(p)) = i\}| \\
&= |\{p \in h'' \mid \text{region}(h''(p)) = i\}| - |\{p \in h' \mid \text{region}(h'(p)) = i\}| \\
&\quad + |\{p \in h' \mid \text{region}(h'(p)) = i\}| - |\{p \in h \mid \text{region}(h(p)) = i\}| \\
&= (\text{diff}(k, h, h'))(i) + (\text{diff}(k, h', h''))(i) \\
&= \delta_1(i) + \delta_2(i)
\end{aligned}$$

Therefore  $\text{diff}(k, h, h'') = \delta_1 + \delta_2$ . Properties (3) and (4) are proven as follows:

$$\begin{aligned}
\text{maxFreshCells}(c_{init} \rightarrow_{S'}^* c_2) &= m_1 \\
\text{maxFreshCells}(c_2 \rightarrow_{S'}^* c_{final}) &= m_2 \\
\text{maxFreshCells}(c_{init} \rightarrow_{S'}^* c_{final}) &= \max \{m_1, m_2 + \text{sizeH}(c_2) - \text{sizeH}(c_{init})\} \\
&= \max \{m_1, m_2 + \sum_{i=0}^k (\text{diff}(k, h, h'))(i)\} \\
&= \max \{m_1, m_2 + |\delta_1|\}
\end{aligned}$$

$$\begin{aligned}
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_1) &= 2 \\
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_2) &= \max \{2, 2 + s_1\} = 2 + s_1 \\
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_2 \rightarrow_{S'}^* c_3) &= \max \{2 + s_1, 1\} = 2 + s_1 \\
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_3 \rightarrow_{S'}^* c_{final}) &= \max \{2 + s_1, 1 + s_2\}
\end{aligned}$$

- **Case [Case]:**  $e \equiv \text{case } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i}^n$

We assume that the  $r$ -th **case** alternative is executed under an environment  $E_r \stackrel{\text{def}}{=} E \uplus [\overline{x_{ri} \mapsto \overline{v_i}^{n_r}}]$ , where the values  $v_i$  are the parameters of the data construction pointed to by  $E(x)$ .

$$E_r \vdash h, k, td + n_r, e_r \Downarrow h', k, v, (\delta, m, s)$$

In addition, let us denote  $\rho_r \stackrel{\text{def}}{=} \rho + [\overline{x_{ri} \mapsto n_r - i - 1}^{n_r}]$ . If  $(is_r, cs_r) = \text{trE } e_r \rho_r$  then it holds that  $cs \supseteq cs_r$ . Moreover, we get  $is = \text{MATCH}(\rho(x)) \overline{\mathbf{p}_j}^m$  with  $cs(\mathbf{p}_r) = is_r$  for each  $r \in \{1..n\}$ . The SVM derivation corresponding to  $is$  results as follows:

$$\begin{aligned}
& \overbrace{(\text{MATCH } (\rho(x)) \overline{\mathbf{p}}_j^{m_i}, h[p \mapsto (j, C_r \overline{v}_i^{n_r})], k_0, k, S, cs'[\mathbf{p}_r \mapsto is_r])}^{c_{init}} \\
\rightarrow_{S'} & \quad \{ \text{since } S!(\rho(x)) = E(x) = p \} \\
& \quad (is_r, h, k_0, k, \overline{v}_i^{n_r} : S, cs') \equiv c_1 \\
\rightarrow_{S'}^* & \quad \{ \text{by i.h. (see below)} \} \\
& \quad ([\text{POPCONT}], h|_{k_0}, k_0, k_0, v : \text{drop } (td + n_r) (\overline{v}_i^{n_r} : S), cs') \\
\equiv & \\
& \quad \underbrace{([\text{POPCONT}], h|_{k_0}, k_0, k_0, v : \text{drop } td S, cs')}_{c_{final}}
\end{aligned}$$

In the same way as in the previous cases, we have to ensure that  $E_r \equiv (\rho_r, \overline{v}_i^{n_r} : S)$ . Let  $z \in \text{dom } \rho_r$ . We make the following case distinction: on one hand, if  $z = x_{ri}$  for some  $i \in \{1..n_r\}$  then  $\rho_r(z) = i - 1$  and hence:

$$(\overline{v}_i^{n_r} : S)!(\rho_r(z)) = (\overline{v}_i^{n_r} : S)!(i - 1) = v_i = E_r(x_{ri}) = E_r(z)$$

On the other hand, if  $z \neq x_{ri}$  for all  $i \in \{1..n_r\}$  then  $\rho_r(z) = \rho(x) + n_r$ . Therefore:

$$(\overline{v}_i^{n_r} : S)!(\rho_r(z)) = (\overline{v}_i^{n_r} : S)!(\rho(x) + n_r) = S!(\rho(x)) = E(z) = E_r(z)$$

Hence we get  $E_r \equiv (\rho_r, \overline{v}_i^{n_r} : S)$  and (1) holds by applying i.h. since  $\text{drop } (td + n_r) S_g = \text{drop } td S$ . Properties (2) and (3) follow trivially from the induction hypothesis. With respect to (4) we get:

$$\begin{aligned}
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_1) &= n_r \\
\text{maxFreshWords}(c_1 \rightarrow_{S'}^* c_{final}) &= s \\
\text{maxFreshWords}(c_{init} \rightarrow_{S'}^* c_{final}) &= \max \{n_r, s + n_r\} = s + n_r
\end{aligned}$$

- **Case [Case!]:**  $e \equiv \text{case! } x \text{ of } \overline{C_i \overline{x}_{ij}^{n_i} \rightarrow e_i}^n$

The proofs of (1) and (4) are similar to those seen for the nondestructive **case**. Property (2) follows trivially from the induction hypothesis and the definition of *diff*. With respect to (3), let  $c_1$  denote the SVM state prior to the execution of the branch  $e_r$ . Then:

$$\begin{aligned}
\text{maxFreshCells}(c_{init} \rightarrow_{S'}^* c_1) &= 0 \\
\text{maxFreshCells}(c_1 \rightarrow_{S'}^* c_{final}) &= m \\
\text{maxFreshCells}(c_{init} \rightarrow_{S'}^* c_{final}) &= \max \{0, m - 1\}
\end{aligned}$$

which proves the desired result.

The ( $\Leftarrow$ ) direction of the theorem can be proved by induction on the length of the  $\rightarrow_{S'}^*$  derivation. Since this proof is mostly standard, we only describe it briefly: since it is known that the SVM is deterministic, we prove that, given a *Safe* program, if the SVM eventually halts then a corresponding  $\Downarrow$ -derivation can be built. We distinguish cases depending on the expression  $e$  being evaluated. The base cases  $e \equiv c$ ,  $e \equiv x$ ,  $e \equiv x @ r$ ,  $e \equiv a_1 \oplus a_2$ , and  $e \equiv C \overline{a}_i^{n_i} @ r$  are straightforward: if the SVM

machine halts with the corresponding result on the top of stack, this result can be used with rules [Lit], [Var], [Copy] and [Cons].

As an example, let us consider  $e \equiv x$ . Let  $(is, cs) = trE e \rho$  be the SVM code generated, where  $cs = []$  and:

$$is = \text{BUILDENV } [\rho(x)] : \overbrace{\text{SLIDE 1 } (topDepth(\rho)) : \text{DECREGION} : [\text{POPCONT}]}^{is_1}$$

$$\underbrace{\hspace{15em}}_{is_2}$$

$$\underbrace{\hspace{10em}}_{is_3}$$

Given any  $cs'$ , the only possible derivation is the following one:

$$\begin{aligned} & \overbrace{(\text{BUILDENV } [\rho(x)] : is_1, h, k_0, k, S, cs')}^{c_{init}} \\ \rightarrow_{S'} & \quad \{\text{since } S!(\rho(x)) = E(x) = v\} \\ & (\text{SLIDE 1 } (topDepth(\rho)) : is_2, h, k_0, k, v : S, cs') \\ \rightarrow_{S'} & \quad \{\text{since } td = topDepth(\rho)\} \\ & (\text{DECREGION} : is_3, h, k_0, k, v : drop td S, cs') \\ \rightarrow_{S'} & \\ & \underbrace{([\text{POPCONT}], h|_{k_0}, k_0, k_0, v : drop td S, cs')}_{c_{final}} \end{aligned}$$

Trivially, by rule [Var],  $E[x \mapsto v] \vdash h, k, td, x \Downarrow h, k, v, ([ ]_k, 0, 1)$ . Property (2) holds because  $diff(k, h, h) = [i \mapsto 0 \mid 0 \leq i \leq k] = \delta$ . Property (3) and (4) hold because  $maxFreshCells(c_{init} \rightarrow_{S'}^* c_{final}) = 0$  and  $maxFreshWords(c_{init} \rightarrow_{S'}^* c_{final}) = 1$ .

With respect to the remaining cases, the  $\rightarrow_{S'}^*$  execution sequence is made up of a set of preliminary actions (building the variable environment in the case of function application, pushing a continuation in the case of **let** expression or executing a MATCH/MATCH! at the beginning of a **case/case!**) followed by the evaluation of the corresponding subexpressions (i.e, either the function being called or the main/auxiliary expressions of a **let** or the **case/case!** branch). The induction hypothesis can be applied to these ( $\rightarrow^*$ )-subderivations in order to get the required assumptions of the corresponding  $\Downarrow$  rule.

As an example, consider  $e \equiv \mathbf{let } x_1 = e_1 \mathbf{ in } e_2$ . Assume  $(is_1, cs_1) = trE e_1 \rho^{++}$  and  $(is_2, cs_2) = trE e_2 \rho_1$  where  $\rho_1 = \rho + [x_1 \mapsto 1]$ . Then  $is = \text{PUSHCONT } \mathbf{p} : is_1$ , in which  $cs_1(\mathbf{p}) = is_2$ . Given  $cs' \supseteq cs$ , the trace corresponding to the execution of  $e$  must be as follows:

$$\begin{aligned} & \overbrace{(\text{PUSHCONT } \mathbf{p} : is_1, h, k_0, k, S, cs')}^{c_{init}} \\ \rightarrow_{S'} & \\ & (is_1, h, k, k, (k_0, \mathbf{p}) : S, cs') \equiv c_1 \\ \rightarrow_{S'}^* & \{\text{by Lemma 2.26}\} \\ & ([\text{POPCONT}], h', k, k, v_1 : (k_0, \mathbf{p}) : S), cs'[\mathbf{p} \mapsto is_2]) \equiv c_2 \\ \rightarrow_{S'} & \\ & (is_2, h', k_0, k, v_1 : S, cs') \equiv c_3 \\ \rightarrow_{S'}^* & \\ & \underbrace{([\text{POPCONT}], h'', k_0, k_0, v : S', cs')}_{c_{final}} \end{aligned}$$

Notice that  $c_1$  contains a continuation  $(k_0, \mathbf{p})$  on top of the stack. The only machine instruction which removes such continuation from the stack is POPCONT. So, in order to reach  $c_{final}$  an intermediate configuration  $c_2 \equiv ([\text{POPCONT}], h', k', k', v_1 : (k_0, \mathbf{p}) : S), cs'[\mathbf{p} \mapsto is_2])$  must be reached so that the execution can proceed. By Lemma 2.26,  $k' = k$ .

In fact, it holds that  $c_1 \rightarrow_{(k_0, \mathbf{p}) : S}^* c_2$ , so we can apply i.h. to obtain  $E \vdash h, k, 0, e_1 \Downarrow h', k, v_1, (\delta_1, m_1, s_1)$ , as  $(k_0, \mathbf{p}) : S = \text{drop } 0 ((k_0, \mathbf{p}) : S)$ .

Notice also that  $E \uplus [x_1 \mapsto v_1] \equiv \rho + [x_1 \mapsto 1]$  and that  $\text{drop } (td + 1) (v_1 : S) = \text{drop } td S = S'$ , so we can also apply i.h. to  $c_3 \rightarrow_S^* c_{final}$  to obtain  $E \uplus [x_1 \mapsto v_1] \vdash h', k, td + 1, e_2 \Downarrow h'', k, v, (\delta_2, m_2, s_2)$ . The reasoning about resources consumption is the same as the opposite implication.  $\square$

## 2.9 A global overview of the *Safe* certifying compiler

The *Safe* compiler is mostly implemented in Haskell, and it runs in several phases. These are shown in Figure 2.32. First, the input is scanned and parsed in order to obtain an abstract syntax tree (AST), which is represented as a Haskell term. For the implementation of this phase we have used standard tools [42, 77]. Then, the compiler's front-end performs the following phases:

- **Renamer/Contextual constraints checker [31]:** The analyses implemented in the subsequent phases assume that all bound variables have different names. Thus all variables are renamed in this phase. Additionally, the contextual constraints of the language are checked, e.g. that every occurrence of a variable is in scope, functions and data constructors are called with the right number of arguments, etc.
- **Hindley-Milner type and region inference [85]:** This phase decorates every expression, data constructor and function definition with its corresponding type, after inferring it. The AST is also decorated with information about the regions in which each data structure lives. In particular, it determines whether a given data structure is local or not to the current function call. Chapter 4 deals extensively with this phase.
- **Core-Safe transformation [31]:** The original *Safe* program is translated into a desugared and semantically equivalent *Core-Safe* version. The previously inferred type decorations are preserved.
- **Sharing analysis [98]:** Given two variables belonging to the same function definition, it computes whether the respective data structures pointed to by them may share memory locations at run-time. The program is decorated with the sharing information that is needed by the following phase.
- **Destruction analysis [84, 81]:** It infers a typing for the source program w.r.t. the type system given in Chapter 3, guaranteeing that dangling pointers are not generated as a consequence of destructive pattern matching. The design of this phase is dealt with in Chapter 5.
- **Memory Consumption Analysis [83]:** It computes an upper-bound to the memory requirements of a *Safe* program. Chapter 7 explains the way in which these bounds are obtained.

The compiler can produce several kinds of outputs. So far we have implemented the following back-ends:

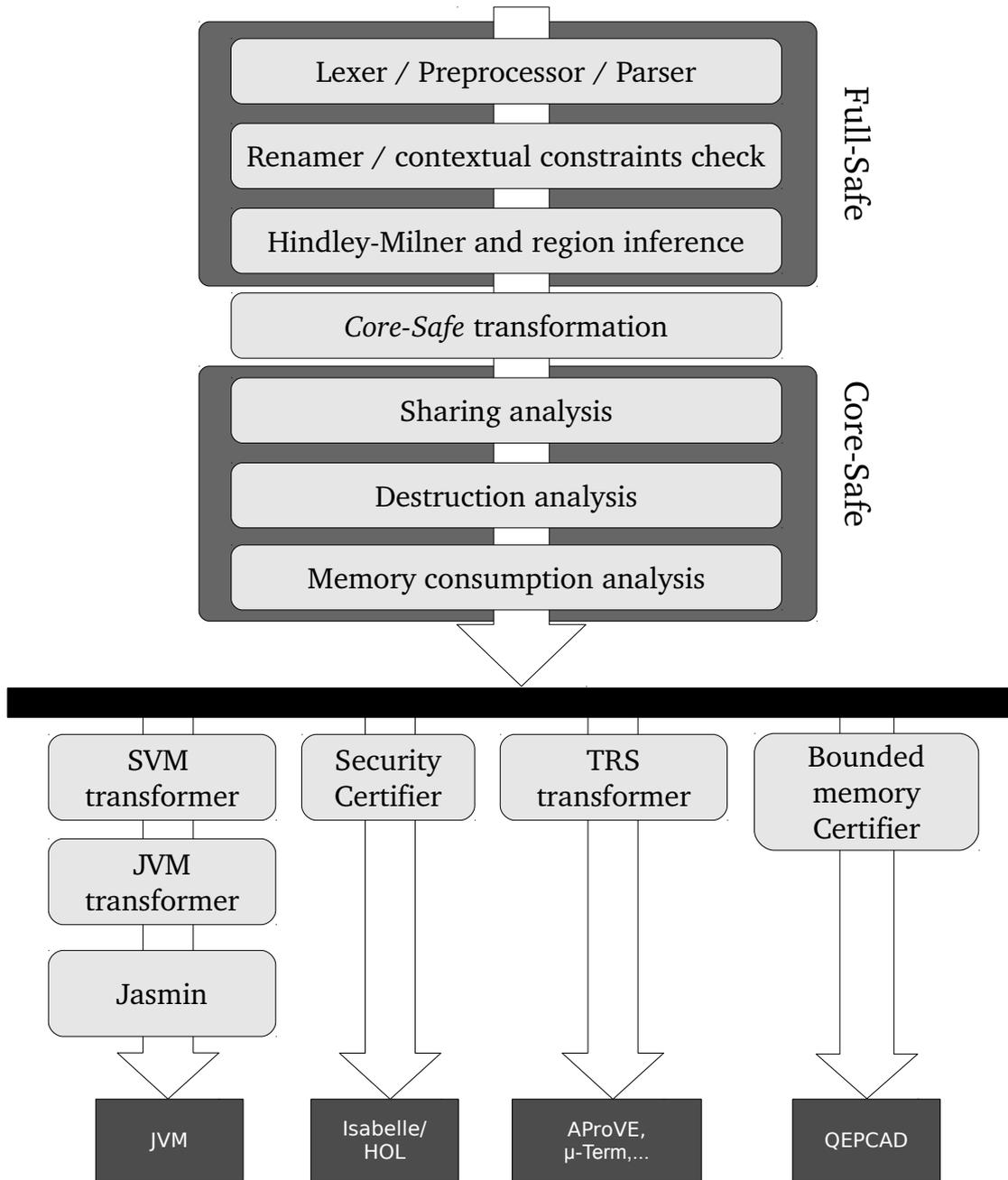


Figure 2.32: Front-end and back-ends of *Safe*'s compiler.

- **Java bytecode generation** [39, 38]: The SVM translation process explained in Section 2.6 is used as an intermediate step for obtaining Java bytecode, which can be run under the Java Virtual Machine (JVM).
- **Certification of security properties** [36]: By using the information given by the destruction analysis phase, this back-end generates an Isabelle/HOL script proving the absence of dangling pointers in the compiled program. In a PCC scenario, the code consumer would receive this formal proof attached to the Java bytecode, in order to verify these safety properties. Chapter 6 describes the process of generating the certificates.
- **Certification of memory bounds** [40]: Given the results of the memory consumption analysis, this back-end certifies that the computed memory bounds are correct for the program being analysed. This phase generates first-order formulas in Tarski’s theory of closed fields. These formulas are proven valid with the QEPCAD system [23].
- **TRS transformer** [73]: It translates the *Core-Safe* definitions into Term Rewriting Systems (TRS), so that their termination can be checked by existing tools, like AProVE [47] or MU-TERM [72].
- **Other back-ends**: For instance, translation of *Safe* programs into C.

This work concerns mainly the design and correctness properties of some of these phases, but hardly any implementation detail. These can be found in [80], or in [79] (the latter in Spanish). The *Safe* compiler is available for download at <http://dalila.sip.ucm.es/safe/>. There also exists a web-based interface at <http://dalila.sip.ucm.es/~safe/>.

## 2.10 Conclusions and related work

In this chapter we have applied a semi-systematic method for refining operational semantics and abstract machines in order to find the way from an abstract view of the language to an efficient implementation. Other contributions include a semantics enriched with memory costs, and the proof of correctness of these costs when translating *Safe* to imperative code. This resource-aware semantics is the basis for proving correct the memory consumption static analysis presented in Chapter 7, and for certifying these bounds, done in [40].

The use of regions in functional languages to avoid garbage collection is not new, as it was explained in Section 1.1.1. Tofte and Talpin [116] introduced in MLKit (a variant of ML) the use of nested regions by means of a **letregion** construct. A lot of work has been done on this system [2, 19, 113]. Their main contribution is a *region inference* algorithm adding region annotations at the intermediate language level. A small difference with these approaches is that, in *Safe*, region allocation and deallocation are synchronized with function calls instead of being introduced by a special language construct. This simplifies the process of inferring regions, as we shall explain in Chapter 4. However, this comes at the cost of granularity for determining the region scopes: MLKit allows several lexically-scoped regions in the same function. A more relevant difference is that *Safe* has an additional mechanism allowing the programmer to selectively destroy data structures inside a region.

A difficulty with the original Tofte and Talpin’s system is the fact that regions have nested lifetimes. There exist a few programs (such as the *inssort* function shown in Example 2.11) that may result in memory leaks due to this restriction. In [45] this problem is alleviated by defining a variant of  $\lambda$ -calculus with type-safe primitives for creating, accessing and destroying regions, which are not restricted to

have nested lifetimes. Programs are written in a C-like language called *Cyclone* having explicit memory management primitives, then they are translated into this variant of  $\lambda$ -calculus, and then type checked. So, the price of this flexibility is explicit region control. In our language *Safe*, regions also suffer from the nested lifetimes constraint, since both region allocation and deallocation are bound to function calls, which are necessarily nested. However, the destructive pattern matching facility compensates for this, since it is possible to dispose of a data structure without deallocating the whole region in which it resides. Allocation and destruction of distinct data structures are not necessarily nested, and the type system presented here protects the programmer against missuses of this feature. Again, the price of this flexibility is explicit deallocation of cells. Allocation is implicit in constructions and the target region of the allocation is inferred by the compiler. It is arguable whether it is better to explicitly manage regions or cells. In [98] a more detailed comparison with all these works can be found.

The destructive pattern matching features of the language have been inspired by Hofmann and Jost's match construct [58], whose operational behaviour is similar to that of *Safe's case!*. The main difference is that they lack a compile time analysis guaranteeing the safe use of this dangerous feature, since it is beyond the scope of their work.

There have been other successful derivations of abstract machines starting from high level descriptions of the semantics. For instance, in [52] and [1] a number of such derivations are done. Well known abstract machines for the  $\lambda$ -calculus such as SECD, Krivine's, CLS and CAM are derived and proved correct. These papers propose general schemes for achieving this kind of derivations. The differences with the present work are the following:

- They concentrate on the pure  $\lambda$ -calculus and they consider neither sharing nor heaps. Algebraic types, **case** and **let** expressions are not considered either.
- In the second paper, the starting point is a denotational meaning of the source language, while here we start from an operational semantics.
- In order to refine their machines they use predefined correct transformations such as closure conversion, transformation into continuation passing style, defunctionalization and inlining.
- They ignore the compilation issues from the source language to machine instructions, and also resource consumption.

In [65] a broad survey of both abstract and virtual machines for the  $\lambda$ -calculus and for practical functional languages is done. The author presents in detail some well-known and other less known abstract machines. When the machines execute compiled code, also the translation schemes are provided. The aim of the book is to serve as a text for a graduate course and no attempt is done to provide proofs of correctness either of the machines or of the compilation schemes.

For the first abstract machine M2 we have found inspiration in Sestoft's derivation of abstract machines for a lazy  $\lambda$ -calculus [104]. For the rest of the derivation, Peña (supervisor of this thesis) has reported some previous experience in [41], but, in that occasion, the destination machine was known in advance. The present work represents a 'real' derivation in the sense that the destination machine has been invented from scratch.

Compared to other eager machines such as Landin's SECD machine [70], it is an added value of our abstract machine that the standard translation yields constant stack space for tail recursion, as we have shown in Example 2.17. For instance, in the G-machine the compiler needs to explicitly identify tail recursion and to do a special translation in this case, i.e. it is considered as an optimization of the code

generation phase. The same happens in other compiled virtual machines such as  $\pi$ -RED<sup>+</sup> [46]. Additionally, our SVM machine does not need a garbage collector and all memory allocation/deallocation actions have been implemented in constant time.

For the semantics enriched with a resource vector, we have found inspiration in [9]. Some other resource-enriched semantics have been proposed. See for example [58, 56] for a big-step semantics of a simple functional language with some information about the number of cells needed by an expression, while [64] extends the same idea to a higher-order language. In [51] Hammond and Michaelson define a model for computing heap and stack costs for a subset of the Hume language [50].

The translation shown in this chapter has been formally verified by de Dios and Peña with the help of the Isabelle/HOL proof assistant. They address the translation from *Core-Safe* to SVM in [39], and the translation from SVM instructions to Java bytecode in [38].



## Chapter 3

# Type system

### 3.1 Introduction

The destruction features provided by *Safe* allow a memory efficient implementation of several data structures and algorithms. However, there exists a risk at runtime: the execution may access some parts of the memory that have been disposed of previously. A program is said to be *pointer-safe* if this eventuality never arises at runtime for any input given to the program.

In this chapter we devise a typed-based approach to prove that a given program is pointer-safe. This problem is, in general, undecidable: the type system developed in this chapter can prove a program pointer-safe, but it cannot prove the absence of this property. As a consequence, the type system may reject some pointer-safe programs. It is desirable to get as few rejected pointer-safe programs as possible. In Section 3.5 we show some examples of successfully typed programs, which should give an idea about the power of the type system.

The main result of this chapter is the proof of the following fact: *the existence of a typing derivation for a program guarantees its pointer-safety*. However, in this chapter we do not explain *how* to get this typing derivation. This question is postponed to the next two chapters. The contents of this chapter are based on the work described in [84]. However, the type system of this chapter contains significant improvements with respect to that of [84]. In particular, it accepts more pointer-safe programs, and admits an inference algorithm (Chapter 5) which is more efficient than that of [84].

### 3.2 Type system concepts

Since the *Safe* language is largely based on Haskell, it inherits many of the ideas in Haskell's type system, which is based, in turn, on Hindley-Milner type system. In particular, it supports polymorphic algebraic data types and inherits the list data type (denoted by  $[\alpha]$ , being  $\alpha$  the type of the elements of the list) with its standard constructors  $[\ ]$  and  $(:)$ , as well as the tuple data type, denoted by  $(\alpha_1, \dots, \alpha_n)$ .

When extending the Hindley-Milner type system to *Core-Safe*, one must assign a type to region variables. For this purpose we define a new category of types: *region type variables*. The type of a region variable is a region type variable (abbreviated as RTV in the following). These RTVs, which will be denoted by  $\rho, \rho_1, \dots$ , act as ordinary polymorphic variables in Haskell. However, only region variables are allowed to have a RTV as its type.

Algebraic data types are annotated with RTVs, which always coincide with the types of the region

*append xs ys*

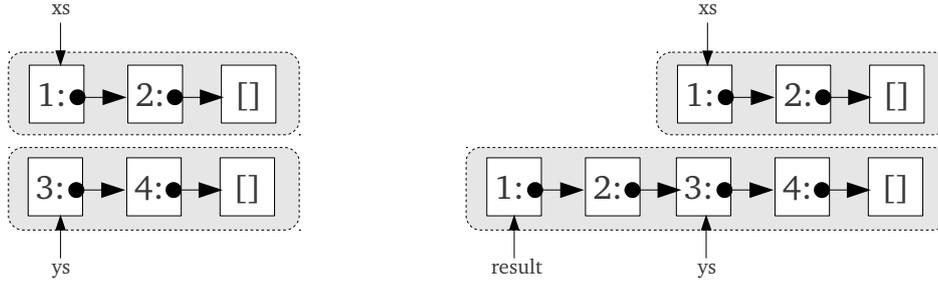


Figure 3.1: Representation of the input lists in the heap before (left-hand side) and after (right-hand side) a call to the *append* function. The input lists can be located in different regions, but the result have to be constructed in the same region as the second list, since the latter is reused in the base case.

variables used in the creation of the corresponding DS. For example, if  $r$  has type  $\rho$ , the expression  $[] @ r$  has type  $[\alpha]@ \rho$ . We have these kind of annotations with the aim of stating a connection between data structures and region variables, and connections among different data structures. For example, if two variables have  $[\alpha]@ \rho$  and  $[\beta]@ \rho$  as their respective types, their corresponding lists must live in the same region at runtime. Moreover, if there exists another region variable  $r'$  with type  $\rho$ , every DS being constructed with this variable at runtime will also live in the same region as these two lists. It is important to have a way to determine these connections at compile time, because it allows us to know, in particular, whether a given data structure resides in the temporary region *self*, which is the only region variable of type  $\rho_{self}$ .

**Example 3.1.** Recall from Example 2.5 the following function for appending two lists:

$$\begin{aligned} \text{append } xs \text{ } ys @ r = & \text{ case } xs \text{ of} \\ & [] \quad \rightarrow ys \\ & (x : xx) \rightarrow \text{let } x_1 = \text{append } xx \text{ } ys @ r \\ & \quad \text{in } (x : x_1) @ r \end{aligned}$$

Let us assume that the  $r$  variable has type  $\rho$ . Without considering region annotations in data types, we would obtain this type for *append*.

$$\text{append} :: \forall \alpha \rho . [\alpha] \rightarrow [\alpha] \rightarrow \rho \rightarrow [\alpha]$$

Now let us determine the RTVs that would be attached to each  $[\alpha]$ . The branch guarded by  $(x : xx)$  returns the result of the construction  $(x : x_1) @ r$ , which has type  $[\alpha]@ \rho$ . Notice, however, that the branch guarded by  $[]$  returns the list passed as second parameter. As we will see below, the typing rules demand that the result of all branches in a **case** expression must have the same type. Therefore, the type of the second parameter must be  $[\alpha]@ \rho$ , which results in the following type for *append*:

$$\text{append} :: \forall \alpha \rho \rho_1 . [\alpha]@ \rho_1 \rightarrow [\alpha]@ \rho \rightarrow \rho \rightarrow [\alpha]@ \rho$$

The  $\rho_1$  annotation in the first parameter means that the region of the corresponding list may be different from that of the second parameter and from the result (Figure 3.1)

Assume the variation of the *append* function given in Example 2.7 that, in the base case, returns a

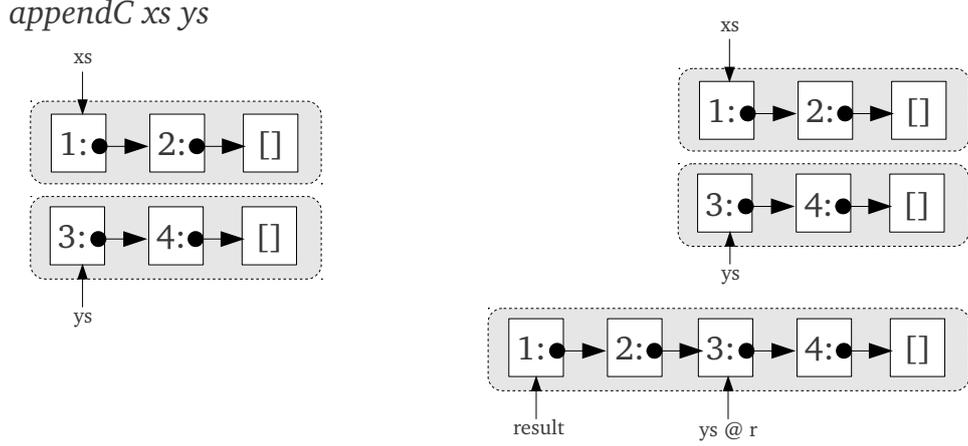


Figure 3.2: Representation of the input lists in the heap before (left-hand side) and after (right-hand side) a call to the `appendC` function. Now the result can be constructed from scratch in a different region.

copy of the second parameter instead of the original:

```

appendC xs ys @ r = case xs of
  []      → ys @ r
  (x : xx) → let x1 = appendC xx ys @ r
              in (x : x1)@r

```

In this case, the result of the expression `ys @ r` has type  $[\alpha]@r$ , whose RTV need not be the same as that of `ys`. Thus we obtain the following type for `appendC`:

$$\text{appendC} :: \forall \alpha \rho \rho_1 \rho_2. [\alpha]@r_1 \rightarrow [\alpha]@r_2 \rightarrow \rho \rightarrow [\alpha]@r$$

The function `appendC` does not require that the list passed as second parameter lives in the same region as the result (Figure 3.2) □

Algebraic data types can be associated with more than one RTV. For example, the following definition:

```

data TBL  $\alpha \beta @ \rho_1 \rho_2 \rho_3 = \text{TBL } [(\alpha, \beta)@r_1]@r_2 @ r_3$ 
```

defines a concrete implementation of the table abstract data type, represented as a list of  $(key, value)$  pairs. Data structures of this type may spread up to three different regions: one for the `TBL` constructor, another for the spine of the list containing the pairs, and another one for the pairs themselves (Figure 3.3).

In general, our algebraic types take the form  $T \bar{s}_i @ \bar{\rho}_j$ , where  $T$  is a *type constructor* (`TBL` in our example), and  $\bar{s}_i, \bar{\rho}_j$  are the type arguments and RTVs that are applied to this constructor. The last RTV of the list  $\bar{\rho}_j$  is the type of the region where the data structures of this type are built. This RTV is called the *outermost region*. In our example, the outermost region of the `TBL` data type is  $\rho_3$ . When the compiler decorates recursive `data` declarations with RTVs (see Section 4.3), the recursive occurrences of the type being defined must have the same RTVs as in the type definition. As a consequence, data types are not polymorphic recursive, unlike function definitions, in which region-polymorphic recursion is allowed.

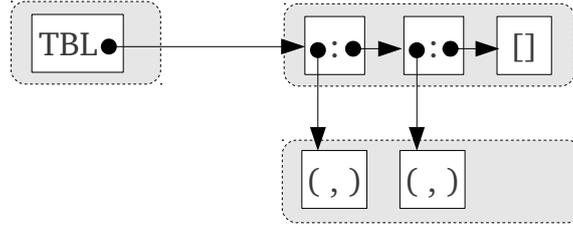


Figure 3.3: The *TBL* datatype can be distributed over three different regions. The outermost region contains the constructor *TBL*.

The following example shows the RTV-annotated definition of a binary search tree data type:

$$\mathbf{data} \text{ BSTree } \alpha @ \rho = \text{Empty } @ \rho \mid \text{Node } (\text{BSTree } \alpha @ \rho) \alpha (\text{BSTree } \alpha @ \rho) @ \rho$$

In the following we will write  $T @ \bar{\rho}_j$  instead of  $T \bar{s}_i @ \bar{\rho}_j$  when the  $\bar{s}_i$  are not relevant. We shall even use the notation  $T @ \rho$  to highlight only the outermost region.

The type of a variable also reflects whether it is destroyed by means of a **case!** expression. This is represented by a (!) symbol in the corresponding types. For example, in the following expression, the variable *xs* gets  $[\alpha]!@ \rho$  as its type:

$$\mathbf{case!} \text{ xs of } [] \rightarrow \dots \\ (x : xx) \rightarrow \dots$$

Types with a (!) mark are called *condemned*. Variables with condemned types can occur in the discriminant of a destructive **case!**, as well as in the destructive arguments of a function call. In our example, the variable *xx* also gets a condemned type  $[\alpha]!@ \rho$ . On the contrary, if a variable is not affected (even indirectly<sup>1</sup>) by a **case!**, we say that it has a *safe* type. We use the term *underlying type* to refer to the Hindley-Milner component of a type with its RTVs, without regard to whether this type is condemned or safe.

In a standard Hindley-Milner type system, a variable always has the same type along its scope. This does not always hold in our type system: the underlying types are preserved along the scope, but some of them may have a condemned mark (!) only in a few of its occurrences. Let us consider the following example:

$$\mathbf{let} \ x_1 = (\mathbf{case} \ \text{xs of } [] \rightarrow \dots) \ \mathbf{in} \ (\mathbf{case!} \ \text{xs of } [] \rightarrow \dots)$$

The variable *xs* has type  $[\alpha]@ \rho$  in the auxiliary expression of the **let**, and type  $[\alpha]!@ \rho$  in the main expression. Each subexpression will be typed under a different typing environment: let us denote by  $\Gamma_1$  the typing environment for the auxiliary expression, and by  $\Gamma_2$  the typing environment for the main expression. The variable *xs* is bound to a safe type in  $\Gamma_1$  and to a condemned type in  $\Gamma_2$ . For typing the whole **let** expression, both environments are merged into a single environment  $\Gamma$  which assigns a condemned type to *xs*, meaning that at some point of the execution of the **let**, the DS pointed to by this variable will be destroyed. However, not every combination is possible. If we swapped the subexpressions of our **let** as follows:

<sup>1</sup>Even if a DS is not directly involved in a **case!**, it may point to a cell being destroyed. We shall explain this in more detail in Section 3.3.

<b>Type</b> $\ni$	$\tau$	$\rightarrow$	$t$	{external}
			$r$	{in-danger}
			$\sigma$	{polymorphic function}
			$\rho$	{region type, being $\rho \in \mathbf{RegType}$ }
			$t$	{safe}
			$d$	{condemned}
<b>SafeType</b> $\ni$	$s$	$\rightarrow$	$T \bar{s} @ \bar{\rho}^m$	{algebraic, with $m \geq 1$ }
			$\alpha$	{variable, being $\alpha \in \mathbf{TypeVar}$ }
			$B$	{basic, being $B \in \{Int, Bool\}$ }
<b>CdmType</b> $\ni$	$d$	$\rightarrow$	$T \bar{s} ! @ \bar{\rho}^m$	{condemned algebraic, with $m \geq 1$ }
<b>DgrType</b> $\ni$	$r$	$\rightarrow$	$T \bar{s} \# @ \bar{\rho}^m$	{in-danger algebraic, with $m \geq 1$ }
<b>FunType</b> $\ni$	$tf$	$\rightarrow$	$\bar{t}_i \rightarrow \bar{\rho} \rightarrow s$	{function}
<b>SafeFunType</b> $\ni$	$sf$	$\rightarrow$	$\bar{s}_i \rightarrow \bar{\rho} \rightarrow s$	{safe function}
			$\sigma$	$\rightarrow \forall \bar{\alpha} \bar{\rho}. tf$ {type scheme}

Figure 3.4: Syntax of type expressions

**let**  $x_1 = (\mathbf{case!} \textit{xs of} [] \rightarrow \dots)$  **in**  $(\mathbf{case} \textit{xs of} [] \rightarrow \dots)$

the outermost **let** would be ill-typed. In Section 3.4 we will describe precisely which combinations of safe and condemned marks are allowed.

**Example 3.2.** Revisiting our *append* example, let us consider its destructive variant (Example 2.10):

$$\begin{aligned} \mathit{appendD} \textit{xs ys} @ r = & \mathbf{case!} \textit{xs of} \\ & [] \quad \rightarrow \textit{ys} \\ & (x : \textit{xx}) \rightarrow \mathbf{let} \textit{x}_1 = \mathit{appendD} \textit{xx ys} @ r \\ & \quad \mathbf{in} (\textit{x} : \textit{x}_1) @ r \end{aligned}$$

The function *appendD* gets the following type:

$$\mathit{appendD} :: \forall \alpha \rho \rho_1. [\alpha]! @ \rho_1 \rightarrow [\alpha] @ \rho \rightarrow \rho \rightarrow [\alpha] @ \rho$$

□

### 3.3 Type expressions and environments

The syntax of type expressions is shown in Figure 3.4. Since our language is first-order, we distinguish between functional types  $\sigma$  and non-functional types, which can be RTVs  $\rho$  or algebraic data types  $t, r$ .

Algebraic types may be *safe* types  $s$ , *in-danger* types  $r$  or *condemned* types  $d$ . In-danger types arise as an intermediate step during typing and are useful to control the side-effects of the destructions. But notice that the types of functions only include either safe or condemned types; that is why there is

a separate syntactical category  $t$  of external types which rules out the possibility of having in-danger types in function signatures.

The intended semantics of each category of algebraic types is the following:

- **Safe types ( $s$ ):** The DSs of this type cannot be directly or indirectly destroyed, so they are kept intact during the evaluation of the expression being typed. Basic types *Int* and *Bool* and type variables are also included in this category.
  - *Capabilities:* DSs of safe types can be read, copied and used to build other DSs.
  - *Prohibitions:* DSs of safe types can neither be used in the discriminant of a **case!** expression, nor point to a cell that is being destroyed by a **case!**. Variables with safe types cannot appear in destructive positions of function applications.
  - *Guarantees:* All the cells pointed to by a DS of a safe type are guaranteed to remain intact during the evaluation of the expression being typed.
- **Condemned types ( $d$ ):** The DSs of this type can receive the action of a destructive **case!**, but this destructive action is limited to the recursive spine of the DS.
  - *Capabilities:* A DS of this type can be read, copied and used to build other DSs before being destroyed. It can be part in the discriminant of a **case!**. In this case, its recursive spine can also be destroyed through its recursive pattern variables.
  - *Prohibitions:* The non-recursive part of a DS of a condemned type cannot point to a cell being destroyed.
  - *Guarantees:* The non-recursive descendants of a condemned DS are ensured to remain intact during the evaluation of the expression being typed, but the recursive spine of a condemned DS may get totally or partially corrupted during this evaluation.
- **In-danger types ( $r$ ):** The DSs of this type can point to a recursive descendant of a condemned DS, so they contain potentially dangling pointers.
  - *Capabilities:* A DS of this type can be read, copied and used to build other DSs before being corrupted. It can also occur in the discriminant of a **case!**. Its recursive spine can be destroyed, as well as its non-recursive descendants.
  - *Prohibitions:* None.
  - *Guarantees:* None. Both recursive and non-recursive descendants of an in-danger DS can be destroyed.

From these informal descriptions, we can easily see that safe types are the “strongest” ones, in the sense that they are the most restrictive (e.g. they cannot point to anything being destroyed), but they offer more guarantees with respect to the integrity of their corresponding DSs. On the other hand, in-danger types are the most permissive, but they cannot ensure nothing about which part of the corresponding DS is kept intact. This motivates the following order on these categories:  $s \leq d \leq r$ . Before presenting this order relation in a formal way, let us introduce some notation: we use the names **SafeType**, **CdmType** and **DgrType** to denote the set of all safe, condemned and in-danger types, respectively. We also define the sets **ExpType** and **UnsafeType** as follows:

$$\begin{aligned} \mathbf{ExpType} &\stackrel{\text{def}}{=} \mathbf{SafeType} \cup \mathbf{CdmType} \cup \mathbf{DgrType} \\ \mathbf{UnsafeType} &\stackrel{\text{def}}{=} \mathbf{CdmType} \cup \mathbf{DgrType} \end{aligned}$$

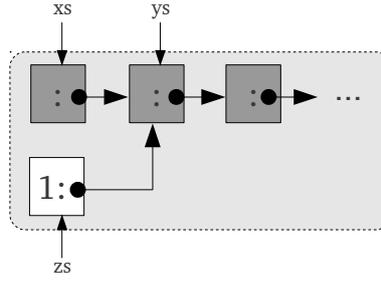
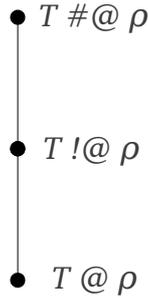


Figure 3.5: Situation described in Example 3.3: the recursive spine of  $xs$  is condemned (gray cells). Since  $zs$  points to it, it gets an in-danger type.

The predicate  $utype?(\tau, \tau')$ , where  $\tau, \tau' \in \mathbf{ExpType}$ , tells whether two types have the same underlying type (i.e. without regard to ! or # marks). By abuse of notation, we extend this predicate to function schemes as follows:  $utype?(\sigma, \sigma') \Leftrightarrow \sigma = \sigma'$ .

Given these notational conventions, the  $\leq$  partial order on types is defined as follows:

$$\tau_1 \leq \tau_2 \stackrel{def}{\Leftrightarrow} utype?(\tau_1, \tau_2) \wedge (\tau_1 = \tau_2 \vee \tau_1 \in \mathbf{SafeType} \vee \tau_2 \in \mathbf{DgrType}) \quad (3.1)$$



**Example 3.3.** In order to illustrate the meanings of these types, let us consider the following *Core-Safe* definition:

$$f \text{ } xs @ r = \mathbf{let} \text{ } ys = (\mathbf{case} \text{ } xs \text{ of } (x : xx) \rightarrow xx) \mathbf{in} \\ \mathbf{let} \text{ } zs = (1 : ys) @ r \mathbf{in} \mathbf{case!} \text{ } xs \text{ of } \dots$$

The variable  $ys$  refers to the tail of the list pointed to by  $xs$ , which is used in the definition of  $zs$  to build another list (Figure 3.5). After that, the first cons cell of the list pointed to by  $xs$  is disposed by means of a **case!**. In order to type the **case!**  $xs \dots$  expression under an environment  $\Gamma$ , the variable  $xs$  must appear in this environment with a condemned type  $[Int]!@ \rho$ . Moreover, since both  $ys$  and  $zs$  are sharing a recursive descendant of  $xs$ , they must occur in  $\Gamma$  with an unsafe type. Notice that the variable  $xs$  may appear in the typing environment of the non-destructive **case** with a safe type  $[Int]@ \rho$ , since it is not being destroyed in that expression.  $\square$

A functional type  $tf$  has the form  $\bar{t}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s$ , where  $\bar{t}_i^n$  and  $\bar{\rho}_j^m$  are the types of the data and region parameters, and  $s$  is the type of the result. Notice that function parameters are restricted by syntax to be safe types, condemned types or RTVs. The motivation for not allowing in-danger types in function signatures is the following: when a parameter is condemned we know clearly that

only the recursive substructure of the corresponding DS may be condemned. When the function is applied to an argument we know that the recursive substructure of such argument may be partially or totally destroyed. However, when a parameter is in-danger, the only thing we know is that some part (recursive or not) of the whole DS may be dangling, but we do not know which part. This is a very imprecise information to put in the type of a function. In spite of this restriction, in-danger types are still useful for controlling and propagating the side effects of explicit destruction through a function's body.

We distinguish a special category of functional types, **SafeFunType**, which contains the types of those functions whose parameters have safe types. We use the notation  $sf$  for denoting an element of this set. Constructor types are a particular instance of these types. They have a single region argument  $\rho$  which coincides with the outermost region variable of the resulting algebraic type  $T \bar{s}_i @ \bar{\rho}_j^m$ . Constructors are given types indicating that the recursive substructure and the structure itself must live in the same region, as we have pointed out before. For example, in the case of lists and trees:

$$\begin{aligned}
[] &:: \forall \alpha \rho . \rho \rightarrow [\alpha]@ \rho \\
(:) &:: \forall \alpha \rho . \alpha \rightarrow [\alpha]@ \rho \rightarrow \rho \rightarrow [\alpha]@ \rho \\
Empty &:: \forall \alpha \rho . \rho \rightarrow BSTree \alpha @ \rho \\
Node &:: \forall \alpha \rho . BSTree \alpha @ \rho \rightarrow \alpha \rightarrow BSTree \alpha @ \rho \rightarrow \rho \rightarrow BSTree \alpha @ \rho
\end{aligned}$$

By abuse of notation, we shall also consider type schemes  $\sigma$  to belong to **FunType** or **SafeFunType**, depending on the functional type after the  $\forall$  quantifiers.

Expressions are typed in the context of *type environments*, which contain the type of variables, and function/constructor names in scope. A type environment  $\Gamma$  contains region type bindings  $[r : \rho]$ , variable type bindings  $[x : t]$ ,  $[x : r]$  and polymorphic scheme assignments to functions  $[f : \sigma]$  and constructor symbols  $[C : \sigma]$ . Formally, a type environment is a triple of mappings in this set:

$$\Gamma : (\mathbf{Var} \rightarrow \mathbf{ExpType}) \times (\mathbf{RegVar} \rightarrow \mathbf{RegType}) \times (\mathbf{Fun} \cup \mathbf{Cons} \rightarrow \mathbf{FunType})$$

However, by abuse of notation, we consider  $\Gamma$  as a single function. It will be clear which specific component is meant, when applying this function to a variable  $x$ , region variable  $r$ , function symbol  $f$ , or constructor symbol  $C$ . We also define several operators (summarized in Table 3.1) for composing two type environments:

- The usual  $+$  operator demands disjoint domains. Its result is the union of the bindings of the environments to which it is applied.

$$\Gamma_1 + \Gamma_2 \text{ defined} \Leftrightarrow \text{dom } \Gamma_1 \cap \text{dom } \Gamma_2 = \emptyset$$

$$\text{dom } (\Gamma_1 + \Gamma_2) = \text{dom } \Gamma_1 \cup \text{dom } \Gamma_2$$

$$(\Gamma_1 + \Gamma_2)(x) = \begin{cases} \Gamma_1(x) & \text{if } x \in \text{dom } \Gamma_1 \\ \Gamma_2(x) & \text{otherwise} \end{cases}$$

- The  $\oplus$  operator allows a variable to occur in several environments, but only if it has the same *safe* type in all of them.

$$\Gamma_1 \oplus \Gamma_2 \text{ defined} \Leftrightarrow \forall x \in \text{dom } \Gamma_1 \cap \text{dom } \Gamma_2 . \Gamma_1(x) = \Gamma_2(x) \in \mathbf{SafeType}$$

•	$\Gamma_1 \bullet \Gamma_2$ defined if:	Result of $\Gamma_1 \bullet \Gamma_2$ :
+	$\Gamma_1$ and $\Gamma_2$ have disjoint domains	Union of $\Gamma_1$ and $\Gamma_2$ bindings
$\oplus$	Common variables in $\Gamma_1$ and $\Gamma_2$ have the same safe type	
$\sqcup$	Common variables in $\Gamma_1$ and $\Gamma_2$ have the same underlying type	Union of $\Gamma_1$ and $\Gamma_2$ bindings (highest types take precedence)

Table 3.1: Informal descriptions of the operators on type environments.

$$\text{dom}(\Gamma_1 \oplus \Gamma_2) = \text{dom} \Gamma_1 \cup \text{dom} \Gamma_2$$

$$(\Gamma_1 \oplus \Gamma_2)(x) = \begin{cases} \Gamma_1(x) & \text{if } x \in \text{dom} \Gamma_1 \\ \Gamma_2(x) & \text{otherwise} \end{cases}$$

- The  $\sqcup$  operator, when applied to the environments  $\Gamma_1$  and  $\Gamma_2$ , allows a variable to occur with a different type in each environment, provided that the underlying type in each of the occurrences is the same. The result of  $\Gamma_1 \sqcup \Gamma_2$  contains all the bindings from  $\Gamma_1$  and  $\Gamma_2$ . If the same variable occurs in both environments with a different type, the higher (weaker) version takes priority over the lower one.

$$\Gamma_1 \sqcup \Gamma_2 \text{ defined} \Leftrightarrow \forall x \in \text{dom} \Gamma_1 \cap \text{dom} \Gamma_2. \text{utype}(\Gamma_1(x), \Gamma_2(x))$$

It is easy to see that, if  $\tau_1$  and  $\tau_2$  have the same underlying type, then either  $\tau_1 \leq \tau_2$  or  $\tau_2 \leq \tau_1$  w.r.t. the order relation given in 3.1. In these cases we use the notation  $\tau_1 \sqcup \tau_2$  to denote the maximal type between both.

$$\text{dom}(\Gamma_1 \sqcup \Gamma_2) = \text{dom} \Gamma_1 \cup \text{dom} \Gamma_2$$

$$(\Gamma_1 \sqcup \Gamma_2)(x) = \begin{cases} \Gamma_1(x) & \text{if } x \notin \text{dom} \Gamma_2 \\ \Gamma_2(x) & \text{if } x \notin \text{dom} \Gamma_1 \\ \Gamma_1(x) \sqcup \Gamma_2(x) & \text{if } x \in \text{dom} \Gamma_1 \cap \text{dom} \Gamma_2 \end{cases}$$

It is easy to see that these three operators are commutative and associative.

**Example 3.4.** Given the following type environments:

$$\begin{aligned} \Gamma_1 &= [x : [\alpha]@_{\rho_2}] \\ \Gamma_2 &= [x : [\alpha]!@_{\rho_2}, y : Int] \\ \Gamma_3 &= [x : [\alpha]!@_{\rho_2}, z : Bool] \\ \Gamma_4 &= [x : [\alpha]#@_{\rho_2}, y : Bool] \end{aligned}$$

- None of the  $\Gamma_1 + \Gamma_2$ ,  $\Gamma_2 + \Gamma_3$ ,  $\Gamma_1 + \Gamma_3$ ,  $\Gamma_1 + \Gamma_4$ ,  $\Gamma_2 + \Gamma_4$ ,  $\Gamma_3 + \Gamma_4$  is defined.
- $\Gamma_2 \oplus \Gamma_3$  is undefined, since  $x$  has an unsafe type in both environments.
- $\Gamma_2 \sqcup \Gamma_4$  is undefined, since the underlying types of  $\Gamma_2(y)$  and  $\Gamma_4(y)$  are not equal. However,  $\Gamma_2 \sqcup \Gamma_3$ ,  $\Gamma_3 \sqcup \Gamma_4$  as well as  $\Gamma_1 \sqcup \Gamma_3 \sqcup \Gamma_4$  are defined as follows:

$$\begin{aligned} \Gamma_2 \sqcup \Gamma_3 &= [x : [\alpha]!@_{\rho_2}, y : Int, z : Bool] \\ \Gamma_3 \sqcup \Gamma_4 &= [x : [\alpha]#@_{\rho_2}, y : Bool, z : Bool] \\ \Gamma_1 \sqcup \Gamma_3 \sqcup \Gamma_4 &= [x : [\alpha]#@_{\rho_2}, y : Bool, z : Bool] \end{aligned}$$

$$\begin{array}{c}
\frac{\Gamma' \vdash e : s \quad \Gamma \supseteq \Gamma'}{\Gamma \vdash e : s} \text{ [EXT]} \quad \frac{\Gamma + [x : \tau_1] \vdash e : s \quad \tau_1 \leq \tau_2}{\Gamma + [x : \tau_2] \vdash e : s} \text{ [WEAK]} \\
\\
\frac{}{\emptyset \vdash c : B} \text{ [LIT]} \quad \frac{}{[x : s] \vdash x : s} \text{ [VAR]} \\
\\
\frac{}{[x : T @ \rho', r : \rho] \vdash x @ r : T @ \rho} \text{ [COPY]} \\
\\
\frac{\Gamma_1 \vdash e_1 : s_1 \quad \Gamma_2 + [x_1 : \tau_1] \vdash e_2 : s \quad \text{utype?}(\tau_1, s_1) \quad \forall x \in \text{dom } \Gamma_1, \Gamma_1(x) \in \mathbf{UnsafeType} \Rightarrow x \notin \text{fv}(e_2)}{\Gamma_1 \sqcup \Gamma_2 \vdash \mathbf{let } x_1 = e_1 \mathbf{ in } e_2 : s} \text{ [LET]} \\
\\
\frac{\begin{array}{l} \bar{t}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s \leq \sigma \quad \Gamma = \Gamma_R + [f : \sigma] + \bigoplus_{i=1}^n [a_i : t_i] + [\bar{r}_j : \bar{\rho}_j^m] \\ \Gamma_R = \bigsqcup_{a_i \in \mathbf{CdmType}} \text{SHR}(a_i, f \bar{a}_i^n @ \bar{r}_j^m) \quad \bigwedge_{t_i \in \mathbf{CdmType}} \text{isTree}(a_i) \end{array}}{\Gamma \vdash f \bar{a}_i^n @ \bar{r}_j^m : s} \text{ [APP]} \\
\\
\frac{\bar{s}_i^n \rightarrow \rho \rightarrow s \leq \sigma \quad \Gamma = [C : \sigma] + \bigoplus_{i=1}^n [a_i : s_i] + [r : \rho]}{\Gamma \vdash C \bar{a}_i^n @ r : s} \text{ [CONS]} \\
\\
\frac{\begin{array}{l} \forall i \in \{1..n\}. \Gamma(C_i) = \sigma_i \quad \text{utype?}(\Gamma(x), T @ \rho) \quad \forall i \in \{1..n\}. \bar{s}_{ij}^{n_i} \rightarrow \rho \rightarrow T @ \rho \leq \sigma_i \\ \forall i \in \{1..n\}. \Gamma + [\bar{x}_{ij} : \bar{\tau}_{ij}^{n_i}] \vdash e_i : s \quad \forall i \in \{1..n\}. \forall j \in \{1..n_i\}. \text{utype?}(s_{ij}, \tau_{ij}) \end{array}}{\Gamma \vdash \mathbf{case } x \mathbf{ of } \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n : s} \text{ [CASE]} \\
\\
\frac{\begin{array}{l} \forall i \in \{1..n\}. \Gamma(C_i) = \sigma_i \quad \forall i \in \{1..n\}. \bar{s}_{ij}^{n_i} \rightarrow \rho \rightarrow T @ \rho \leq \sigma_i \\ \Gamma_R = \text{SHR}(x, \mathbf{case! } x \mathbf{ of } \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n) \quad \forall i \in \{1..n\}. \forall j \in \{1..n_i\}. \text{inh}(t_{ij}, s_{ij}, T ! @ \rho) \\ \forall i \in \{1..n\}. \forall z \in \text{dom } \Gamma_R \cup \{x\}. z \notin \text{fv}(e_i) \quad \forall i \in \{1..n\}. \Gamma + [\bar{x}_{ij} : \bar{\tau}_{ij}^{n_i}] \vdash e_i : s \end{array}}{\Gamma_R \sqcup (\Gamma \setminus x) + [x : T ! @ \rho] \vdash \mathbf{case! } x \mathbf{ of } \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n : s} \text{ [CASE!]}
\end{array}$$

Figure 3.6: Typing rules for expressions. □

The  $\leq$  order relation defined on types can be extended to typing environments as follows:

$$\Gamma_1 \leq \Gamma_2 \stackrel{\text{def}}{\iff} \text{dom } \Gamma_1 \subseteq \text{dom } \Gamma_2 \wedge \forall x \in (\text{dom } \Gamma_1) \cap \mathbf{Var}. \Gamma_1(x) \leq \Gamma_2(x)$$

This relation is motivated by the fact that, if an expression is typeable under an environment  $\Gamma_0$ , it should be typeable under every environment  $\Gamma$  such that  $\Gamma \geq \Gamma_0$  (i.e.  $\Gamma$  is less restrictive than  $\Gamma_0$ ). In the following section we describe in detail the typing relation.

### 3.4 Typing rules

Figure 3.6 shows the rules for typing expressions. These rules define the derivation of judgements of the form  $\Gamma \vdash e : s$ , meaning that the expression  $e$  has a safe type  $s$  under an environment  $\Gamma$ . Some of these rules use the operators on type environments explained previously, and they implicitly assume that the application of each of these operators results in a well-defined environment. If it does not, the corresponding rule cannot be applied.

There are rules for typing literals [LIT] and variables [VAR]. Notice that these expressions are typed

only under the minimal typing environment w.r.t. the  $\leq$  relation on environments. If we want to type these expressions under a weaker environment, we can do it by means of the [EXT] and [WEAK] rules. The first one is useful for extending the environment with fresh bindings, whereas the second one allows us to downgrade the type category of a given variable to a weaker one. That is, safe variables can be demoted to unsafe, whereas condemned variables can be demoted to in-danger. Both [EXT] and [WEAK] rules are useful in the context of **case** and **case!** expressions, since their corresponding typing rules (which we shall explain later) demands all the **case(!)** branches to be typeable under the same environment (excluding pattern variables, which may differ between branches).

**Example 3.5.** Given the following expression:

$$\begin{array}{l} \mathbf{case } b \mathbf{ of} \\ \quad True \rightarrow e_1 \\ \quad False \rightarrow e_2 \end{array}$$

Assume we have derived the following typing judgements for each branch

$$[x : [\alpha]!\rho] \vdash e_1 \quad [x : [\alpha]\rho, z : Int] \vdash e_2$$

We can apply the [EXT] rule to the first judgement, and the [WEAK] rule to the second one, so both expressions are typable under the same environment:

$$\frac{[x : [\alpha]!\rho] \vdash e_1}{[x : [\alpha]!\rho, z : Int] \vdash e_1} \text{ [EXT]} \quad \frac{[x : [\alpha]\rho, z : Int] \vdash e_2}{[x : [\alpha]!\rho, z : Int] \vdash e_2} \text{ [WEAK]}$$

□

The [COPY] rule allows any variable of an algebraic type to be copied. The outermost region variable  $\rho'$  in the type of the DS being copied is replaced by the type  $\rho$  of the destination region.

Rule [LET] allows the bound variable to get a safe, condemned or in-danger type in the main expression  $e_2$ , provided it has the same underlying type as the expression  $e_1$ . The side condition in this rule demands that any variable with an unsafe type in  $e_1$  cannot be mentioned (i.e. occur free) in  $e_2$ , since it may contain dangling references. Lastly, if the DS pointed by a variable is not guaranteed to remain intact during the evaluation of one of the sub-expressions of the **let**, it cannot be guaranteed to have a safe type when typing the whole **let** expression. That is why we take the least upper bound  $\Gamma_1 \sqcup \Gamma_2$  of the environments typing each sub-expression.

Rule [APP] deals with function application. The operator  $\leq$  denotes the instantiation of a type scheme, in which the bound variables of that scheme can be replaced by safe types. The use of the  $\oplus$  operator avoids a variable to be used in two or more different positions unless they are all safe parameters. Otherwise undesired side-effects could happen, such as trying to destroy the same cell twice in the function being called. The notation  $\Gamma_R = SHR(x, e)$  is defined as follows:

$$\Gamma_R = SHR(x, e) \stackrel{\text{def}}{\iff} \begin{array}{l} \text{dom } \Gamma_R = \text{sharerec}(x, e) \setminus \{x\} \\ \wedge \forall y \in \text{dom } \Gamma_R. \Gamma_R(y) \in \mathbf{DgrType} \end{array}$$

The  $\text{sharerec}(x, e)$  notation denotes the set of variables in the scope of  $e$  that may point to a recursive descendant of the DS pointed to by  $x$  at runtime. This set is computed by an abstract interpretation-based sharing analysis, which is described in [98]. This analysis approximates the sharing relations

between the different variables occurring in the body of a function, but it shall not be detailed here. In Section 3.6.3 we state its correctness properties, since they are essential for proving the correctness of the type system.

The *SHR* notation specifies that  $\Gamma_R$  contain every variable in the scope of  $e$  that may point to a recursive descendant of  $x$ , and these variables are mapped by  $\Gamma_R$  to an in-danger type. This notation can be extended to join the sharing information of several variables as follows:

$$\Gamma_R = \bigsqcup_{i \in I} SHR(x_i, e) \stackrel{def}{\iff} \forall i \in I. \exists \Gamma_i. (\Gamma_i = SHR(x_i, e) \wedge \Gamma_i \leq \Gamma_R)$$

In our particular [APP] rule, the  $\Gamma_R$  environment contains all those variables that may point to an argument being passed as a condemned parameter of  $f$ . The disjoint union of  $\Gamma_R$  with the rest of the arguments ensure that there are no arguments with an in-danger type in the context of the function application, in the same way as there are no in-danger parameters in the function definition. It also propagates the information about possibly damaged pointers downwards in the derivation tree.

**Example 3.6.** Given a function  $g$  with the following type:

$$g :: [Int]!@_1 \rightarrow [Int]!@_1 \rightarrow [Int]@_1 \rightarrow [Int]@_1 \rightarrow Int$$

The function call  $g \ xs \ ys \ xs \ zs$  is ill-typed, because the following environment is undefined:

$$[xs : [Int]!@_1] \oplus [ys : [Int]!@_1] \oplus [xs : [Int]@_1] \oplus [zs : [Int]@_1]$$

If we swap the second and third arguments, the result  $g \ xs \ xs \ ys \ zs$  is also ill-typed for the same reasons:

$$[xs : [Int]!@_1] \oplus [xs : [Int]!@_1] \oplus [ys : [Int]@_1] \oplus [zs : [Int]@_1]$$

However, we can provide the same argument twice in two safe positions. For example, in  $g \ xs \ ys \ zs \ zs$  we would obtain the following environment:

$$[xs : [Int]!@_1] \oplus [ys : [Int]!@_1] \oplus [zs : [Int]@_1] \oplus [zs : [Int]@_1]$$

which is well-defined and equivalent to  $[xs : [Int]!@_1, ys : [Int]!@_1, zs : [Int]@_1]$ . Notice that we cannot provide arguments with an in-danger type. For example, in the following code fragment:

```
case  $xs$  of  
  ( $x : xx$ )  $\rightarrow f \ xs \ ys \ xx \ zs$ 
```

the function call cannot be typed, since  $xx$  is a recursive descendant of a variable in a condemned position ( $xs$ ). Therefore, it occurs in the  $\Gamma_R$  environment with an in-danger type, and the following is undefined:

$$\underbrace{[xx : r]}_{\Gamma_R} + ([xs : [Int]!@_1] \oplus [ys : [Int]!@_1] \oplus [xx : [Int]@_1] \oplus [zs : [Int]@_1])$$

□

The *isTree* predicate is given by an auxiliary analysis, which is run jointly with the sharing analysis of [98]. An intuition on this predicate can be given by considering the recursive spine of a DS as a directed

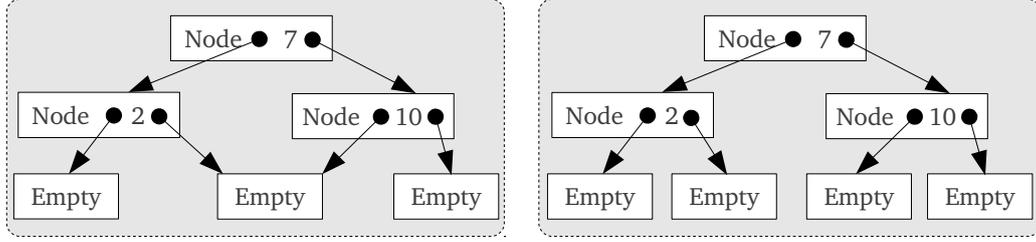


Figure 3.7: Internal sharing in binary search trees (BSTs). The graph of the BST at the left-hand side is not a tree, but a directed acyclic graph. If such a BST is pointed to by a variable, the predicate *isTree* will never hold at compile-time for that variable. On the contrary, the graph of the BST at the right-hand side is a tree.

graph, in which the vertices are the cells of its spine and the edges represent the connections between cells: there is an edge from  $w_1$  to  $w_2$  iff  $w_1$  contains a pointer to  $w_2$ . The predicate  $isTree(x)$  holds if the graph associated to the DS pointed to by  $x$  is guaranteed to be a tree at runtime or, in other words, if we cannot reach a recursive descendant of that DS via two different paths (see Figure 3.7). Otherwise, the graph associated to that DS is not a tree, but a directed acyclic graph (DAG). The  $\bigwedge_{a_i \in \text{CdmType}} isTree(a_i)$  condition in the [APP] rule demands that there should not be internal sharing between the cells pointed to by the variables occurring in condemned positions of the function. Otherwise the function being called might attempt to destroy a child of a condemned DS twice.

**Example 3.7.** Consider the following function for destroying a binary search tree:

$$\begin{aligned}
 &destroyBST :: BSTree \alpha! \rightarrow Int \\
 &destroyBST Empty = 0 \\
 &destroyBST (Node l x r) = \mathbf{let} \ x_1 = destroyBST \ l \ \mathbf{in} \\
 &\quad \mathbf{let} \ x_2 = destroyBST \ r \ \mathbf{in} \ 0
 \end{aligned}$$

If we execute the following expression:

$$\mathbf{let} \ x_1 = Empty \ \mathbf{in} \ destroyBST (Node \ x_1 \ 0 \ x_2)$$

there would be an attempt to destroy the DS pointed to by  $x_1$  twice. □

For the purposes of the type system, a constructor application  $C \bar{a}_i^n @ r$  can be considered as a function application in which all the parameters have a safe type. Hence the [CONS] rule is a particular instance of [APP].

Rule [CASE] allows its discriminant variable to be safe, in-danger, or condemned, since it only reads that variable. We demand the Hindley-Milner types of the discriminant and the pattern variables to be instances of the types of the corresponding constructors, but pattern variables may also occur with a (!) mark or a (#) mark in the environments. This provides more flexibility in comparison with the type system of [84], in which the marks of the pattern variables can be restricting, depending on the mark of the discriminant.

In rule [CASE!] the discriminant is destroyed and consequently the text should not try to reference it in the alternatives. The same happens to those variables sharing a recursive substructure of  $x$ , as they may be corrupted. These substructures can only be accessed through the pattern variables occurring in recursive positions of the constructor. The corrupted variables are added to the environment  $\Gamma_R$  with

$$\begin{aligned} inh(s, s, d) &\Leftrightarrow \neg utype?(s, d) \\ inh(d, s, d) &\Leftrightarrow utype?(s, d) \end{aligned}$$

Figure 3.8: Definitions of inheritance compatibility.

$$\frac{\Gamma + [\overline{x_i : t_i^n}] + [\overline{r_j : \rho_j^l}] + [self : \rho_{self}] + [f : \forall \overline{\rho}. \overline{t_i^n} \rightarrow \overline{\rho_j^l} \rightarrow s] \vdash e : s}{\{\Gamma\} f \overline{x_i^n} @ \overline{r_j^l} = e \quad \{\Gamma + [f : \forall \overline{\alpha} \overline{\rho}. \overline{t_i^n} \rightarrow \overline{\rho_j^l} \rightarrow s]\}} \text{ [FUN]}$$

Figure 3.9: Typing rule for function definitions.

in-danger types. Relation  $inh$ , defined in Figure 3.8, determines the types inherited by pattern variables: recursive ones are condemned while non-recursive ones must be safe. This reflects the intended meaning of safe and condemned types: since the discriminant is condemned, only the recursive part of the DSs pointed to the pattern variables in recursive positions may be destroyed, whereas those variables in non-recursive positions must be left intact.

In Figure 3.9 we show the [FUN] rule for typing function definitions. The informal meaning of a judgement  $\{\Gamma\} def \{\Gamma'\}$  is that, assuming that the type of every function called in the body of  $def$  is contained in  $\Gamma$ , the definition  $def$  is well-typed and  $\Gamma'$  is the extension of  $\Gamma$  with the type of the function being defined. We use  $\forall \overline{\alpha} \overline{\rho}$  to denote the standard generalization of a functional type with respect to all its type variables, and  $\forall \overline{\rho}$  to denote the generalization of the region variables of a functional type. The function's body is typed under an environment which contains the types of the data and region parameters. The  $self$  region is assigned a special type  $\rho_{self}$  which must be distinct from the region types of the  $r_j$ , the region types occurring in the result, and the region types of the input parameters. With this constraint we ensure that the deallocation of the  $self$  region when the function finishes does not cause dangling pointers, since the result of the function does not reside there.

Notice that the  $\forall \overline{\rho}$  notation allows us to have region-polymorphic recursion: inside the body  $e$ , different applications may use different regions. This kind of polymorphism cannot be applied to ordinary type variables, since the type reconstruction problem would be undecidable, as shown by Henglein [53]. However, if we restrict ourselves to recursion over RTVs, this problem becomes decidable. In Chapter 4 we explain the advantages of allowing this kind of recursion.

**Example 3.8.** The following function is ill-typed:

$$f \ xs = \mathbf{case} \ xs \ \mathbf{of} \ (x : xx) \rightarrow xx @ self$$

The type of its body is  $[\alpha] @ \rho_{self}$ , so the condition  $\rho_{self} \notin rtv(s)$  does not hold. Operationally, this function would build a copy of the tail of the input list into the working region and return a reference to the result. However, the result is lost after the function finishes because of the  $self$  deallocation. Hence, this function always returns a dangling pointer.  $\square$

The following property of the type system will be useful in its correctness proof.

**Lemma 3.9** (Substitution lemma). *If  $\Gamma \vdash e : s$  and  $\theta$  is a substitution that maps type variables  $\alpha$  to safe types and region types  $\rho$  to region types, then  $\theta(\Gamma) \vdash e : \theta(s)$ .*

*Proof.* Since  $\theta$  maps type variables only to safe types, and RTVs to RTVs, the application of  $\theta$  to a type  $\tau$  does not change its category. Formally:

$$\tau \in X \Rightarrow \theta(\tau) \in X \quad \text{where } X \in \{\mathbf{SafeType}, \mathbf{DgrType}, \mathbf{CdmType}, \mathbf{RegType}, \mathbf{FunType}\} \quad (3.2)$$

This implies, in particular:

1.  $utype?(\tau, \tau') \Leftrightarrow utype?(\theta(\tau), \theta(\tau'))$  for all  $\tau, \tau' \in \mathbf{ExpType}$ , which can be proven by induction on the structure of  $\tau$ .
2.  $inh(\tau_1, \tau_2, \tau_3) \Leftrightarrow inh(\theta(\tau_1), \theta(\tau_2), \theta(\tau_3))$  for any  $\tau_1, \tau_2, \tau_3 \in \mathbf{ExpType}$ , which follows from (1) and (3.2).
3.  $\Gamma_1 \diamond \Gamma_2$  is defined iff  $\theta(\Gamma_1) \diamond \theta(\Gamma_2)$  is defined, for  $\diamond \in \{+, \oplus, \sqcup\}$  and any  $\Gamma_1, \Gamma_2$ . This follows from (1) and (3.2).
4.  $\theta(\Gamma_1 \diamond \Gamma_2) = \theta(\Gamma_1) \diamond \theta(\Gamma_2)$  for  $\diamond \in \{+, \oplus, \sqcup\}$  and any  $\Gamma_1, \Gamma_2$ . This follows from (3) and (3.2).

Therefore, the premises in the derivation  $\theta(\Gamma) \vdash e : \theta(s)$  follow from their counterparts in  $\Gamma \vdash e : s$ , and the lemma can be easily proved by induction on the latter derivation.  $\square$

### 3.5 Case studies

In this section we show some examples of successfully typed programs. The first two examples show the typing derivations of two simple functions.

**Example 3.10.** Given the *appendD* function of Example 2.12, it gets the following type:  $[\alpha]!@_{\rho_1} \rightarrow [\alpha]@\rho \rightarrow \rho \rightarrow [\alpha]@\rho$ . We denote by  $\sigma_{appendD}$  its region-polymorphic variant  $\forall \rho_1 \rho. [\alpha]!@_{\rho_1} \rightarrow [\alpha]@\rho \rightarrow \rho \rightarrow [\alpha]@\rho$ . Let us start typing the recursive application *appendD*  $xx\ ys\ @\ r$ . The sharing analysis gives the following information:

$$sharerec(xx, appendD\ xx\ ys\ @\ r) = \{xs\} \quad isTree(xx)$$

Hence the following relation holds,

$$[xs : [\alpha]\#@_{\rho_1}] = SHR(xx, appendD\ xx\ ys\ @\ r)$$

and the following typing environment is well-defined:

$$[xs : [\alpha]\#@_{\rho_1}] + [f : \sigma_{appendD}] + ([xx : [\alpha]!@_{\rho_1}] \oplus [ys : [\alpha]@\rho]) + [r : \rho]$$

Hence, we can use the [APP] rule for deriving the following judgement:

$$\underbrace{[xs : [\alpha]\#@_{\rho_1}, f : \sigma_{appendD}, xx : [\alpha]!@_{\rho_1}, ys : [\alpha]@\rho, r : \rho]}_{\Gamma_1} \vdash appendD\ xx\ ys\ @\ r : [\alpha]@\rho \quad (3.3)$$

We can apply the [CONS] rule in a similar way to obtain:

$$\underbrace{[x : \alpha, r : \rho]}_{\Gamma_2} + [x_1 : [\alpha]@\rho] \vdash (x : x_1)@r : [\alpha]@\rho \quad (3.4)$$

The result of  $\Gamma_1 \sqcup \Gamma_2$  is well-defined and equal to  $[xs : [\alpha]\#@_{\rho_1}, f : \sigma_{appendD}, xx : [\alpha]!@_{\rho_1}, ys : [\alpha]@\rho, r : \rho, x : \alpha]$ . Besides this, neither  $xs$  nor  $xx$  occur free in  $(x : x_1)@r$ , so we can apply the [LET] rule with

(3.3) and (3.4) in order to type the recursive case of the function:

$$[f : \sigma_{appendD}, xs : [\alpha]\#\rho_1, ys : [\alpha]\rho, r : \rho] + [x : \alpha, xx : [\alpha]!\rho_1] \vdash \mathbf{let} \ x_1 = \dots \mathbf{in} \dots : [\alpha]\rho \quad (3.5)$$

Now we move on to the base case. By applying the [VAR] rule we get,

$$[ys : [\alpha]\rho] \vdash ys : [\alpha]\rho$$

and, by the [EXT] rule,

$$[f : \sigma_{appendD}, xs : [\alpha]\#\rho_1, ys : [\alpha]\rho, r : \rho] \vdash ys : [\alpha]\rho \quad (3.6)$$

Now we use (3.5) and (3.6) in order to apply the [CASE!] rule. It is worth noting that in this case the environment  $\Gamma_R$  is empty, and that the rest of side conditions in that rule hold. Hence we get:

$$[f : \sigma_{appendD}, xs : [\alpha]!\rho_1, ys : [\alpha]\rho, r : \rho] \vdash \mathbf{case} \ xs \ \mathbf{of} \ \dots : [\alpha]\rho$$

Lastly, we apply the [EXT] rule in order to add the binding of the working region variable *self*:

$$[f : \sigma_{appendD}, xs : [\alpha]!\rho_1, ys : [\alpha]\rho, r : \rho, self : \rho_{self}] \vdash \mathbf{case} \ xs \ \mathbf{of} \ \dots : [\alpha]\rho$$

Therefore, the body of the expression is well-typed. This judgement also serves as a base for applying the [FUN] rule, so the function definition is also well-typed.  $\square$

**Example 3.11.** We consider the following variant of the *treesort* function (Example 2.6 on page 26):

$$\begin{aligned} treesortD \ xs \ @ \ r = \\ & \mathbf{let} \ x_1 = mkTreeD \ xs \ @ \ self \\ & \mathbf{in} \ inorder \ x_1 \ @ \ r \end{aligned}$$

where the function *mkTreeD* builds a binary search tree from a list, while destroying the latter.

We assume that *mkTreeD* and *inorder* have already been typed, obtaining their respective type schemes  $\sigma_{mkTreeD}$  and  $\sigma'_{inorder}$ :

$$\begin{aligned} \sigma_{mkTreeD} &= \forall \rho_1 \rho_2. [Int]!\rho_1 \rightarrow \rho_2 \rightarrow BSTree \ Int \ @ \ \rho_2 \\ \sigma'_{inorder} &= \forall \alpha. \sigma_{inorder} \\ \sigma_{inorder} &= \forall \rho_1 \rho_2. BSTree \ \alpha \ @ \ \rho_1 \rightarrow \rho_2 \rightarrow [\alpha]\rho_2 \end{aligned}$$

Firstly we apply the *mkTreeD* application with the [APP] rule:

$$\underbrace{[mkTreeD : \sigma_{mkTreeD}, xs : [Int]!\rho_1, self : \rho_{self}]}_{\Gamma_1} \vdash mkTreeD \ xs \ @ \ self : BSTree \ Int \ @ \ \rho_{self}$$

We proceed similarly with the *inorder* application:

$$\underbrace{[inorder : \sigma_{inorder}, r : \rho]}_{\Gamma_2} + [x_1 : BSTree \ Int \ @ \ \rho_{self}] \vdash inorder \ x_1 \ @ \ r : [Int]\rho$$

With these two judgements, and because the only variable with an unsafe type in  $\Gamma_1$  does not occur free

in the main expression of the **let**, we can apply the [LET] rule so as to get:

$$\underbrace{[mkTreeD : \sigma_{mkTreeD}, inorder : \sigma_{inorder}]}_{\Gamma} + [xs : [Int]!\@_{\rho_1}, r : \rho, self : \rho_{self}] \vdash \mathbf{let} \ x_1 = \dots \mathbf{in} \ \dots : [Int]\@\rho$$

and, by the [EXT] rule:

$$\Gamma + [xs : [Int]!\@_{\rho_1}, r : \rho, self : \rho_{self}] + [treesortD : \sigma] \vdash \mathbf{let} \ x_1 = \dots \mathbf{in} \ \dots : [Int]\@\rho$$

assuming  $\sigma = \forall \rho_1 \rho. [Int]!\@_{\rho_1} \rightarrow \rho \rightarrow [Int]\@\rho$ . We can use this judgement for applying [FUN] and obtain:

$$\{\Gamma\} treesortD \ xs \ @ \ r = \dots \{\Gamma + [treesort : \sigma]\}$$

Therefore, the function is well-typed.

We can build a non-destructive version of *treesortD* by creating a copy of the original list and passing this copy to the *treesortD* function:

$$treesort \ xs \ @ \ r = \mathbf{let} \ xs' = xs \ @ \ self \ \mathbf{in} \ treesortD \ xs' \ @ \ r$$

We get the following type scheme:  $\forall \rho_1 \rho. [Int]\@\rho_1 \rightarrow \rho \rightarrow [Int]\@\rho$ . To type the function's body, we apply the [LET] rule in a similar way as in the destructive version.  $\square$

The following examples are more involved. By the sake of clarity we just show the *Full-Safe* code (annotated with regions in the following example) and give an intuition of the typing derivation, without representing it in detail. All these examples have been accepted by the *Safe* compiler.

**Example 3.12.** In this example we provide an implementation of the insertion sort algorithm, which does not require additional heap memory. We start with the *insertD* function, which destructively inserts an element into an sorted list, such that the result is also a sorted list.

$$\begin{aligned} insertD \ x \ []! \ @ \ r &= (x : [] \ @ \ r) \ @ \ r \\ insertD \ x \ (y : ys)! \ @ \ r & \\ & \quad | \ x \leq \ y = (x : (y : ys) \ @ \ r) \ @ \ r \\ & \quad | \ x > \ y = (y : insertD \ x \ ys \ @ \ r) \ @ \ r \end{aligned}$$

This function destroys the list passed as second parameter as it is traversed. In the recursive case, the cell being destroyed ( $y : ys$ ) can be reused by the result of the function. However, in the base cases, the function needs two available cells to build the result, whereas only one is destroyed in the pattern matching. Thus, in global terms, this function needs a free cell available in the heap in order to insert the element into the list.

The type scheme for this function is  $\forall \rho. Int \rightarrow [Int]!\@\rho \rightarrow \rho \rightarrow [Int]\@\rho$ . The variable *ys* gets a condemned type  $[Int]!\@\rho$  after the pattern matching, since it is the recursive child of the list being destroyed. However, this variable is used to build the  $(y : ys)$  list in the expression guarded by  $x \leq y$ . Since the [CONS] rule demands a safe type for every variable passed to the  $(:)$  constructor we have to apply the [WEAK] rule so the  $(y : ys) \ @ \ r$  expression can be typed under an environment in which *ys* is condemned. With respect to the recursive call, its second parameter must be of a condemned type, and so is *ys*, so the [APP] rule can be applied.

We continue with the *inssortD* function, which repeatedly inserts each element of the input list into the result, which is initially an empty list.

```

data Leftist  $\alpha = LEmpty \mid LNode \text{ Int } (Leftist \alpha) \alpha (Leftist \alpha)$ 

cons (LEmpty)! a r = LNode 1 r a (LEmpty)
cons (LNode n ll la lr)! a r =
  |  $n \geq m = LNode (m + 1) (LNode n ll la lr) a r$ 
  |  $n < m = LNode (n + 1) r a (LNode n ll la lr)$ 
  where (LNode m _ _ ) = r

join LEmpty! r! = r
join (LNode n ll a lr)! LEmpty! = LNode n ll a lr
join (LNode n ll a lr)! (LNode m rl b rr)!
  |  $a < b = cons ll a (join lr (LNode m rl b rr))!$ 
  |  $a \geq b = cons rl b (join rr (LNode n ll b lr))!$ 

emptyPQueue = LEmpty

addPQueue q a = join q (LNode 1 LEmpty a LEmpty)

minPQueue (LNode _ _ a _) = a

delMinPQueue (LNode _ l _ r) = join l r

```

Figure 3.10: Priority queue data type implemented as a leftist tree.

```

inssortD []! @ r = [] @ r
inssortD (x : xs)! @ r = insertD x (inssortD xs @ r) @ r

```

In each recursive call, the first cell of the input list is destroyed. This cell can be reused for building the empty list in the base case, or for calling the *inssortD* function, which needs a free cell available as it has been explained above.

The type of this function is  $\forall \rho . [Int]!@ \rho \rightarrow \rho \rightarrow [Int]@ \rho$ . The *xs* variable is condemned, since it is the recursive child of the cell being destroyed. Hence it can be used in the recursive call to *inssortD*. The result of this call is used destructively in the *inssortD* function.  $\square$

**Example 3.13.** The destructive features of *Safe* can be used to implement data structures whose updating needs no additional heap space. For instance, let us consider an implementation of priority queues based on leftist trees. The source code, taken from [94], is shown in Figure 3.10. The main difference with respect to the version given in [94] is the use of destructive pattern matching in some functions for the sake of constant heap space consumption.

First we define the data type of leftist trees (*Leftist  $\alpha$* ). A leftist tree is a binary tree in which the shortest path from any node to any of its leafs is always the rightmost one. The first argument of the *LNode* constructor contains the length of this path. Leftist trees are maintained in such a way that this length is kept as low as possible. This results in a well-balanced tree. The function *cons* acts as a wrapper of the *LNode* constructor in order to preserve this invariant. This function is destructive on its first parameter, which is reconstructed in the function's body. This is done in order to optimize the memory usage, since otherwise we would need two additional memory cells for executing the base case. By applying destructive pattern matching we can reuse the cell corresponding to the input parameter, so we would only need one memory cell to build the tree. We could have also done without destructive

pattern matching with the following definition:

$$\begin{aligned}
\text{cons } l \ a \ r &= \mathbf{case} \ l \ \mathbf{of} \\
&\quad L\text{Empty} \rightarrow L\text{Node } 1 \ r \ a \ l \\
&\quad L\text{Node } n \ \_ \_ \_ \rightarrow \mathbf{case} \ n \geq m \ \mathbf{of} \\
&\quad\quad \text{True} \rightarrow L\text{Node } (m + 1) \ l \ a \ r \\
&\quad\quad \text{False} \rightarrow L\text{Node } (n + 1) \ r \ a \ l \\
&\quad \mathbf{where} \ (L\text{Node } m \ \_ \_ \_) = r
\end{aligned}$$

With this definition we would still need a single cell, while preserving the tree given as first parameter. This comes at the cost of readability: now the pattern matching is made explicit via **case**. We could avoid this by incorporating the as-patterns facility of Haskell [76]:

$$\begin{aligned}
\text{cons } l@(L\text{Empty}) \ a \ r &= L\text{Node } 1 \ r \ a \ (L\text{Empty}) \\
\text{cons } l@(L\text{Node } n \ \_ \_ \_) \ a \ r &= \\
&\quad | \ n \geq m = L\text{Node } (m + 1) \ l \ a \ r \\
&\quad | \ n < m = L\text{Node } (n + 1) \ r \ a \ l \\
&\quad \mathbf{where} \ (L\text{Node } m \ \_ \_ \_) = r
\end{aligned}$$

Since the current version of *Safe* does not support this kind of as-patterns, we keep the destructive version shown in Figure 3.10.

The *join* function returns the union of two priority queues. Both parameters of this function are condemned, so this operation can be implemented in constant heap space. The fact that both parameters are condemned implies that, in the case in which the tree passed as first parameter is empty, we have to reuse the tree passed as second parameter, since the destruction of the latter does not begin until the second equation of *join*. However, we cannot do the same for the case in which the second parameter is empty, since by the time the execution reaches this equation the destructive pattern matching on the first parameter has already taken place. We are no longer able to reuse a cell that has been destroyed. That is why we have to rebuild the first parameter in the second equation of *join*. The *Core-Safe* translation (without regions) of this function may shed a light on the order in which these destructions are done:

$$\begin{aligned}
\text{join } l \ r &= \mathbf{case!} \ l \ \mathbf{of} \\
&\quad L\text{Empty} \rightarrow r \\
&\quad L\text{Node} \rightarrow \mathbf{case!} \ r \ \mathbf{of} \\
&\quad\quad L\text{Empty} \rightarrow l \ \{\text{error: } l \ \text{has been destroyed}\} \\
&\quad\quad \dots
\end{aligned}$$

The remaining four functions make up the interface of priority queues: creation of an empty queue (*emptyPQueue*), insertion (*addPQueue*), access to the minimal element (*minPQueue*), and removal of the minimal element (*delMinPQueue*). All these functions need constant additional heap space. They are well-typed with respect to the type system explained in this chapter, and they get the following types:

$$\begin{aligned}
\text{cons} &:: \text{Leftist } \alpha \text{ !}@ \rho \rightarrow \alpha \rightarrow \text{Leftist } \alpha \text{ !}@ \rho \rightarrow \rho \rightarrow \text{Leftist } \alpha \text{ @} \rho \\
\text{join} &:: \text{Leftist } \text{Int} \text{ !}@ \rho \rightarrow \text{Leftist } \text{Int} \text{ !}@ \rho \rightarrow \rho \rightarrow \text{Leftist } \text{Int} \text{ @} \rho \\
\text{emptyPQueue} &:: \rho \rightarrow \text{Leftist } \alpha \text{ @} \rho \\
\text{addPQueue} &:: \text{Leftist } \text{Int} \text{ !}@ \rho \rightarrow \text{Int} \rightarrow \rho \rightarrow \text{Leftist } \text{Int} \text{ @} \rho \\
\text{minPQueue} &:: \text{Leftist } \alpha \text{ @} \rho \rightarrow \alpha \\
\text{delMinPQueue} &:: \text{Leftist } \text{Int} \text{ !}@ \rho \rightarrow \rho \rightarrow \text{Leftist } \text{Int} \text{ @} \rho
\end{aligned}$$

All type variables and RTVs in each type are assumed to be universally quantified. The *join* function performs a comparison of the elements of the priority queue with the  $\leq$  operator. *Safe* does not support type classes in this moment (in particular the *Ord* class), so this operator has type  $\text{Int} \rightarrow \text{Int} \rightarrow \text{Int}$ . As a consequence, the *join* function must be applied to priority queues of integers. The same happens with *addPQueue* and *delMinPQueue*, because of their respective calls to *join*. The *cons* and *join* functions forces the input data structures to live in the same region, since they are coalesced in a single data structure. The corresponding RTV  $\rho$  has to be inserted into the signatures of those functions that directly or indirectly build a priority queue. The *minPQueue* function does not build anything, and it does not require this parameter.  $\square$

**Example 3.14.** As another example of a data structure with efficient memory usage, let us consider the following implementation of AVL Trees. Again, the definitions shown here are destructive variants of those appearing in [94]. The AVL tree data type is defined as follows:

$$\mathbf{data} \text{AVLTree } \alpha = \text{AVLEmpty} \mid \text{AVLNode } \text{Int} (\text{AVLTree } \alpha) \alpha (\text{AVLTree } \alpha)$$

The *AVLNode* constructor takes an integer standing for the depth of the tree. This is necessary in order to keep the invariant that the balance factor of each node (that is, the difference between the depths of the left and right subtrees) must be  $-1, 0$  or  $+1$ . Any tree whose nodes satisfy this property is well-balanced. The function *sJoinAVL* is a wrapper of the *AVLTree* constructor and maintains the depth information. It is defined as follows:

$$\text{sJoinAVL } l \ a \ r = \text{AVLNode } (1 + \max (\text{depth } l) (\text{depth } r)) \ l \ a \ r$$

The insertion function in an AVL tree is similar to the insertion in a binary search tree. However, we have to ensure that the invariant of the data structure still holds. If, after insertion, the right subtree of a node outweighs its left subtree (that is, we have a balance factor greater or equal than  $+2$ ) a left rotation or a double left rotation may be needed. This case is handled by the *rJoinAVL* function, which is destructive on its third parameter:

$$\begin{aligned}
&\text{rJoinAVL } l \ a \ (\text{AVLNode } \_ \text{rl } \text{ra } \text{rr})! \\
&\quad | \text{rrd} \geq \text{rld} = \text{sJoinAVL } (\text{sJoinAVL } l \ a \ \text{rl}) \ \text{ra } \text{rr} \\
&\quad | \text{rrd} < \text{rld} = \mathbf{case!} \ \text{rl} \ \mathbf{of} \\
&\quad \quad (\text{AVLNode } \text{rll } \text{rlla } \text{rlr}) \rightarrow \text{sJoinAVL } (\text{sJoinAVL } l \ a \ \text{rll}) \ \text{rlla} \ (\text{sJoinAVL } \text{rlr } \text{ra } \text{rr}) \\
&\quad \mathbf{where} \ \text{rld} = \text{depth } \text{rl} \\
&\quad \quad \text{rrd} = \text{depth } \text{rr}
\end{aligned}$$

Variables  $rl$ ,  $rr$ ,  $rll$  and  $rlr$  get a condemned type, since they are recursive children of data structures being destroyed. As a consequence,  $rl$  can be used in the destructive **case!**.

Similarly, if the left subtree of a node outweighs its right subtree after an insertion, a right rotation or a double right rotation would be necessary. The function  $lJoinAVL$  handles these cases. Its code is similar to that of  $rJoinAVL$  and it will not be shown here. So far, we get the following type signatures:

$$\begin{aligned} sJoinAVL &:: AVLTree\ \alpha\ @\ \rho \rightarrow \alpha \rightarrow AVLTree\ \alpha\ @\ \rho \rightarrow \rho \rightarrow AVLTree\ \alpha\ @\ \rho \\ rJoinAVL &:: AVLTree\ \alpha\ @\ \rho \rightarrow \alpha \rightarrow AVLTree\ \alpha\ !@ \rho \rightarrow \rho \rightarrow AVLTree\ \alpha\ @\ \rho \\ lJoinAVL &:: AVLTree\ \alpha\ !@ \rho \rightarrow \alpha \rightarrow AVLTree\ \alpha\ @\ \rho \rightarrow \rho \rightarrow AVLTree\ \alpha\ @\ \rho \end{aligned}$$

All these functions need constant heap space (one cell) to run. The  $joinAVL$  function is another wrapper for  $AVLNode$ , but, unlike  $sJoinAVL$ , this one ensures that the resulting tree is well-balanced. It is defined as follows:

$$\begin{aligned} &joinAVL\ !\ a\ r! \\ &| \text{abs}(ld - rd) \leq 1 \quad = \quad sJoinAVL\ l\ a\ r \\ &| ld == rd + 2 \quad = \quad lJoinAVL\ l\ a\ r \\ &| ld + 2 == rd \quad = \quad rJoinAVL\ l\ a\ r \\ &| ld > rd + 2 \quad = \quad \mathbf{case!}\ l\ \mathbf{of}\ (AVLNode\ ll\ la\ lr) \rightarrow joinAVL\ ll\ la\ (joinAVL\ lr\ a\ r) \\ &| ld + 2 < rd \quad = \quad \mathbf{case!}\ r\ \mathbf{of}\ (AVLNode\ rl\ ra\ rr) \rightarrow joinAVL\ (joinAVL\ l\ a\ rl)\ ra\ rr \\ &\quad \mathbf{where}\ ld = depth\ l \\ &\quad \quad rd = depth\ r \end{aligned}$$

The *Core-Safe* code corresponding to this function is a sequence of **case** expressions whose branches are the right hand sides of each guard in  $joinAVL$ . The DS pointed to by  $r$  is destroyed in the call to  $rJoinAVL$ . Similarly, the DS corresponding to  $l$  is destroyed in the call to  $lJoinAVL$ . We get the following type for  $joinAVL$ :

$$joinAVL \quad :: \quad AVLTree\ \alpha\ !@ \rho \rightarrow \alpha \rightarrow AVLTree\ \alpha\ !@ \rho \rightarrow \rho \rightarrow AVLTree\ \alpha\ @\ \rho$$

Given the auxiliary functions above, the implementation of the usual insertion and deletion operations is straightforward. Their corresponding functions are accepted by the type system:

$$\begin{aligned} insertAVL &:: Int \rightarrow AVLTree\ Int\ !@ \rho \rightarrow \rho \rightarrow AVLTree\ Int\ @\ \rho \\ deleteAVL &:: Int \rightarrow AVLTree\ Int\ !@ \rho \rightarrow \rho \rightarrow AVLTree\ Int\ @\ \rho \end{aligned}$$

□

### 3.6 Correctness of the type system

The correctness proof of the type system is rather involved. In a first step (Section 3.6.1) we will set up a connection between region type variables and runtime region numbers, and we will show that this connection remains constant during the evaluation of an expression. In Section 3.6.2 we will introduce some notation and definitions in order to formalize the notion of sharing between data structures.

Some of the rules of the type system contain appearances of a *sharerec* function and an *isTree* predicate, which are determined by a set of external static analyses. Although these analyses are not described in this thesis, we have to impose some requirements on them in order to prove the correctness of the type system. Hence, Section 3.6.3 deals with the notion of a *correct* sharing analysis. Finally, in Section 3.6.4 we put all these results together in order to show that the evaluation of a well-typed program does not fail as a consequence of accessing dangling pointers.

### 3.6.1 Preservation of region consistency

A key point in the correctness proof of the type system involves finding a correspondence between static region types and runtime region identifiers. For instance, assume a *Safe* program in which a variable  $x$  has type  $[Int]@ρ$  at a given point. When executing that program,  $x$  must point to a list of integers. Assume that list is located in a region whose identifier is 7. At this point of execution, there is a correspondence between the outermost RTV in the type of  $x$  and the actual region in which the DS pointed to by  $x$  lives. This correspondence is given by the binding  $[ρ \mapsto 7]$ . More generally, we can define this correspondence by means of a *region instantiation*, defined as follows:

**Definition 3.15.** A region instantiation  $\eta : \mathbf{RegType} \rightarrow \mathbb{N}$  is a function from RTVs to natural numbers (interpreted as region identifiers).

An important concept regarding RTVs and region identifiers is the notion of *consistency*. Roughly speaking, the consistency property demands that the region instantiations determined by the types and actual regions of the different variables in scope do not contradict each other. In the example above, if there exists another variable  $y$  of type  $BSTree \alpha @ \rho$ , the DS pointed to by this variable must be located, at runtime, in the same region as  $x$ . Moreover, every DS being constructed with a region variable of type  $\rho$  will also live in that region. Otherwise we would reach an inconsistent configuration.

**Definition 3.16.** Two region instantiations  $\eta$  and  $\eta'$  are said to be consistent if they bind common RTVs to the same number, that is:  $\forall \rho \in \text{dom } \eta \cap \text{dom } \eta'. \eta(\rho) = \eta'(\rho)$ .

The union of two region instantiations  $\eta$  and  $\eta'$  (denoted by  $\eta \uplus \eta'$ ) is defined only if  $\eta$  and  $\eta'$  are consistent and its result is the union of the corresponding bindings. The following definition specifies how to build a region instantiation.

**Definition 3.17.** Given a heap  $h$ , a pointer  $p$ , and a safe type  $s$ , the function *build* is defined as follows:

$$\begin{aligned} \text{build}(h, c, s) &= \emptyset && \text{if } s \in \{\alpha, B\} \\ \text{build}(h, p, T @ \rho) &= \emptyset && \text{if } p \notin \text{dom } h \\ \text{build}(h, p, T @ \rho) &= [\rho \mapsto j] \uplus \biguplus_{i=1}^n \text{build}(h, v_i, s_i) && \text{if } p \in \text{dom } h \\ &\mathbf{where} \quad h(p) = (j, C \bar{v}_i^n) \\ &&& \bar{s}_i^n \rightarrow \rho \rightarrow T @ \rho \preceq \Gamma(C) \end{aligned}$$

provided the resulting region instantiation is well-defined. This means that the  $\uplus$  operator is only applied to consistent region instantiations.

**Example 3.18.** Given the heap in the left-hand side of Figure 3.11, in which the  $x$  variable points to a list of lists, assume  $x$  has type  $[[Int]@ρ_2]@ρ_1$ . Every variable pointing directly to one of the cells of the outermost list must have  $ρ_1$  as the outermost RTV in its type. On the contrary, if a variable points to one

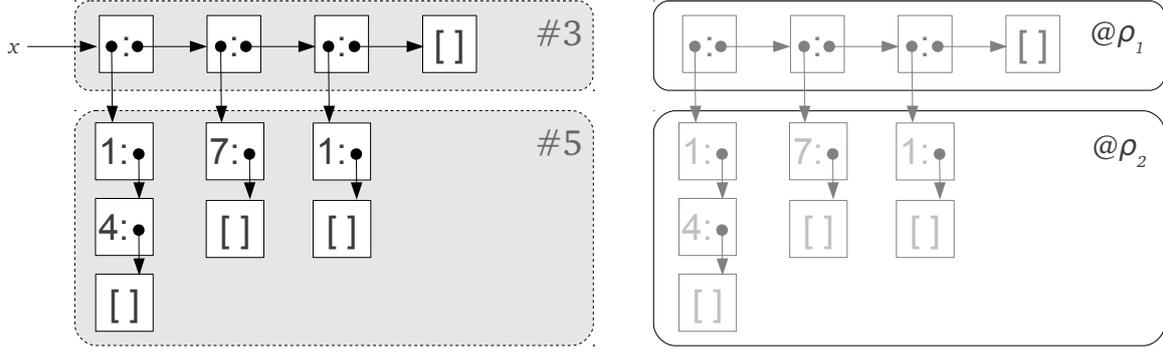


Figure 3.11: Correspondence between the runtime regions in which a DS lives (left) and the RTVs in its type (right).

of the innermost lists, that variable must be of type  $[Int]@_{\rho_2}$ . In this case, the *build* function returns the following region instantiation, which is well-defined:

$$\eta = [\rho_1 \mapsto 3, \rho_2 \mapsto 5]$$

□

**Example 3.19.** Assume the following heap:

$$h = \left[ \begin{array}{l} p_1 \mapsto (2, (:) 0 p_2) \\ p_2 \mapsto (3, []) \end{array} \right]$$

The result of  $build(h, p_1, s)$  is not well-defined for any algebraic type  $s$ . Assume  $s = T @ \rho$ . By applying the definition of *build* we get  $build(h, p_1, T @ \rho) = [\rho \mapsto 2] \uplus [\rho \mapsto 3]$ , which is not well-defined. The reason behind this is that the list pointed to by  $p_1$  is distributed between two different regions (2 and 3), and a DS must live in a single region. □

Besides the triple  $(h, p, s)$ , the result of *build* also depends on a type environment  $\Gamma$  containing the type of the constructors that can be reached by following the pointers that stem from  $p$ . By abuse of notation, we consider this typing environment implicit, as the *build* function only needs the types of the constructors, which are defined globally in the program.

We can extend the *build* function to deal with condemned and in-danger types:

$$\begin{aligned} build(h, p, T !@ \rho) &= build(h, p, T @ \rho) \\ build(h, p, T \# @ \rho) &= build(h, p, T @ \rho) \end{aligned}$$

So far we have determined the region instantiation determined by a single variable. We can extend the *build* function in order to group the region instantiations of all the variables in a runtime environment  $E$ , and whose types can be found in a type environment  $\Gamma$ .

**Definition 3.20.** Given a heap  $h$ , a value environment  $E$ , and a type environment  $\Gamma$ , we define the function  $build^*$  as follows:

$$build^*(h, E, \Gamma) = \bigsqcup_{x \in X} build(h, E(x), \Gamma(x)) \uplus \bigsqcup_{r \in X} [\Gamma(r) \mapsto E(r)] \quad \text{where } X = \text{dom } E \cap \text{dom } \Gamma$$

provided the resulting region instantiation is well-defined.

Notice that the result of  $build^*$  is well-defined only if the region instantiations of the different variables in  $X$  do not contradict each other.

**Example 3.21.** Assume the following heap  $h$ :



Assume a runtime environment  $E$  mapping  $x$  and  $y$  to the cells of the figure above. If we define  $\Gamma = [x : [Int]@p, y : BSTree Int @ p']$ , then  $build^*(h, E, \Gamma)$  results in the following well-defined region instantiation:

$$build^*(h, E, \Gamma) = [\rho \mapsto 1, \rho' \mapsto 2]$$

On the contrary, if we consider the type environment  $\Gamma' = [x : [Int]@p, y : BSTree Int @ p]$ , then  $build^*(h, E, \Gamma')$  is not well-defined, since the tree pointed to by  $y$  does not live in the same region as the list pointed to by  $x$ .  $\square$

Assume a heap  $h$  such that  $build(h, p, \tau)$  is well-defined for some pointer  $p$  and type  $\tau$ . If  $h'$  is a heap that results from removing some pointers from  $h$ , it is easy to see that  $build(h', p, \tau)$  is also well-defined, although it could contain less bindings than the original  $build$ . However, if we extend  $h$  with new bindings we cannot ensure that the corresponding  $build$  is well-defined. For instance, in Example 3.19,  $build(h|_{\{p_1\}}, p_1, [\alpha]@p)$  is well-defined, but  $build(h, p_1, [\alpha]@p)$  is not. When extending a heap, we can ensure the preservation of consistency only if the newly added cells are not reachable from the already existing ones. A heap  $h'$  is said to strictly extend a heap  $h$  if  $h \subseteq h'$  and no pointer in  $\text{dom } h' \setminus \text{dom } h$  is reachable from any pointer in  $h$ . That is, if the pointers we add in  $h'$  are fresh. We denote this by  $h \subseteq_s h'$ . The following lemma formalises all these results.

**Lemma 3.22.** *Let  $h$  and  $h'$  be two heaps. The following two properties hold for each pointer  $p \in \text{dom}(h)$  and region instantiation  $\eta$ :*

1. If  $h \subseteq h'$ , then  $build(h', p, t) \supseteq build(h, p, t)$ , provided  $build(h', p, t)$  is well-defined
2. If  $h \subseteq_s h'$ , then  $build(h', p, t) = build(h, p, t)$

*Proof.* By induction on the size of the structure pointed to by  $p$ .  $\square$

During the evaluation of an expression some cells can be removed from the heap, or new cells can be created, but the contents of a cell are never updated. The following lemma follows from this fact.

**Lemma 3.23.** *Let  $E \vdash h, k, e \Downarrow h', k, v$  be an execution of an expression. For all  $\Gamma$ , if  $build^*(h, E, \Gamma)$  is well-defined, then so is  $build^*(h', E, \Gamma)$ . Moreover, it holds that  $build^*(h', E, \Gamma) \subseteq build^*(h, E, \Gamma)$ .*

*Proof.* This holds because the closure of the variables in  $E$  can only decrease as  $e$  is evaluated (because of destruction), but the contents of the cells in the heap are never updated (see Lemma 3.29 below).  $\square$

The following auxiliary lemma shows that consistency is preserved when copying a data structure to a different region (see Definition 2.13).

**Lemma 3.24.** *Let us assume that  $(h', p') = \text{copy}(h, p, j)$ . If  $\eta$  is consistent with  $\text{build}(h, p, T @ \rho)$  and with  $[\rho' \mapsto j]$ , then  $\text{build}(h', p', T @ \rho')$ , is well-defined and consistent with  $\eta$ .*

*Proof.* By induction on the size of the structure pointed to by  $p$ . Let us assume that  $h(p) = (k, C \bar{v}_i^n)$  and that  $\bar{s}_i^n \rightarrow \rho \rightarrow T @ \rho \trianglelefteq \Gamma(C)$  and  $\bar{s}_i^n \rightarrow \rho' \rightarrow T @ \rho' \trianglelefteq \Gamma(C)$ . From the way in which the types of the constructors are inferred, it follows that  $s_i = s'_i$  if  $i \notin \text{RecPos}(C)$ . We get:

$$\text{build}(h, p, T @ \rho) = [\rho \mapsto k] \uplus \text{build}(h, v_1, s_1) \uplus \dots \uplus \text{build}(h, v_n, s_n)$$

By the definition of *copy*, there is a chain of heaps  $h = h_0 \subseteq_s h_1 \subseteq_s \dots \subseteq_s h_n \subseteq_s h'$  such that for every  $i \in \{1 \dots n\}$ , there exists a value  $v'_i$  such that either  $(h_i, v'_i) = \text{copy}(h_{i-1}, v_i, j)$  or  $(h_i, v'_i) = (h_{i-1}, v_i)$ . Since  $\eta$  is consistent with each  $\text{build}(h, v_i, s_i)$ , it is also consistent with each  $\text{build}(h_{i-1}, v_i, s_i)$  by Lemma 3.22 (2). We have to prove that  $\eta$  is also consistent with  $\text{build}(h_i, v'_i, s'_i)$ . We distinguish cases for each  $i \in \{1 \dots n\}$ :

- $i \notin \text{RecPos}(C)$

Then  $h_i = h_{i-1}$ ,  $s'_i = s_i$  and  $v'_i = v_i$ . Then the consistency of  $\eta$  with  $\text{build}(h_i, v'_i, s'_i)$  follows trivially from the consistency of  $\eta$  with  $\text{build}(h_{i-1}, v_i, s_i)$ .

- $i \in \text{RecPos}(C)$

Then  $(h_i, v'_i) = \text{copy}(h_{i-1}, v_i, j)$ ,  $s_i = T @ \rho$  and  $s'_i = T @ \rho'$ . We apply the induction hypothesis in order to obtain that  $\text{build}(h_i, v'_i, s'_i)$  is well-defined and consistent with  $\eta$ .

Since  $h'$  is a strict extension of every  $h_i$ , we can apply Lemma 3.22 (2) again in order to establish the consistency of  $\eta$  with respect to each  $\text{build}(h', v'_i, s'_i)$ . By the definition of *build* we get:

$$\text{build}(h, p', T @ \rho') = [\rho' \mapsto j] \uplus \text{build}(h', v'_1, s'_1) \uplus \dots \uplus \text{build}(h', v'_n, s'_n)$$

which proves the lemma, since  $\eta$  is consistent with each component in the right-hand side. □

We can prove now that consistency is preserved during the evaluation of an expression  $e$ . As usual, this is proved by induction on the corresponding  $\Downarrow$ -derivation. Nevertheless, we have to be careful when handling the case of function applications. In principle, we can safely assume that the type schemes of the functions being referenced in  $e$  are present in the type environment, but we have no clue on how these type schemes have been obtained. We say that these types schemes are *correct* if they have been obtained via the [FUN] rule of Figure 3.9.

**Definition 3.25.** A typing environment  $\Gamma$  is said to be *correct* with respect to a function signature  $\Sigma$  if for every function symbol  $f \in \text{dom } \Gamma$  it holds that  $f \bar{x}_i @ \bar{r}_j = e_f \in \Sigma$  for some  $\bar{x}_i, \bar{r}_j, e_f$ , and the following judgement is derivable by using the [FUN] rule:

$$\{\Gamma' \setminus f\} f \bar{x}_i @ \bar{r}_j = e_f \{\Gamma'\}$$

for some  $\Gamma' \subseteq \Gamma$  such that  $f \in \text{dom } \Gamma'$ .

The function signature  $\Sigma$  is usually clear from the context, so let us say in the following that a given  $\Gamma$  is correct without making any mention to  $\Sigma$ . Now we are able to establish that consistency is preserved during the evaluation of an expression.

**Theorem 3.26.** *Let us assume the execution of an expression:*

$$E \vdash h, k, e \Downarrow h', k, v \quad (3.7)$$

*Then, for every  $\Gamma, s, \eta$  such that  $\Gamma \vdash e : s$  holds and  $\Gamma$  is a correct environment, if  $\text{build}^*(h, E, \Gamma)$  is well-defined and consistent with  $\eta$ , then  $\text{build}(h', v, s)$  is well-defined and consistent with  $\eta$ , and for every sub-expression  $e_i$  such that the judgement  $E_i \vdash h_i, k, e_i \Downarrow h'_i, k, v_i$  belongs to the derivation of (3.7) and  $\Gamma_i \vdash e_i : s_i$  for some  $E_i, h_i, e_i, h'_i, v_i, \Gamma_i$ , the result of  $\text{build}^*(h_i, E_i, \Gamma_i)$  is well-defined and consistent with  $\eta$ .*

*Proof.* Let us focus on the last rule applied in the derivation of  $\Gamma \vdash e : s$ . If this rule is [EXT] or [WEAK] we obtain  $\Gamma_1 \vdash e : s$  for some  $\Gamma_1 \subset \Gamma$ . By repeating this process we can prove the existence of  $\Gamma'$  and  $\Gamma''$  such that  $\Gamma' \leq \Gamma'' \subseteq \Gamma$ ,  $\Gamma' \vdash e : s$ , and the last rule applied in this latter derivation is neither [EXT] nor [WEAK]. It is easy to see that  $\text{build}^*(E, h, \Gamma')$  is well-defined and consistent with  $\eta$ , since the marks do not play any rule in the definition of  $\text{build}$ .

Given these considerations, we proceed by induction on the depth of the  $\Downarrow$ -derivation, We distinguish cases on the expression being evaluated:

- **Case  $e \equiv c$**

We get  $\Gamma' \vdash c : B$ , and  $\text{build}(h, c, B) = \emptyset$ , which is trivially well-defined and consistent with  $\eta$ .

- **Case  $e \equiv x$**

Since  $x \in \text{dom } E$ ,  $\text{build}(h, E(x), \Gamma'(x))$  is well-defined and consistent with  $\eta$ , and hence  $\text{build}(h, v, s)$  is also well-defined and consistent with  $\eta$ .

- **Case  $e \equiv x @ r$**

We get  $\Gamma' = [x : T @ \rho', r : \rho] \vdash x @ r : T @ \rho$ . Let us assume that  $[r \mapsto j, x \mapsto p] \subseteq E$ . Then  $\eta$  is consistent with  $[\rho \mapsto j]$ . Therefore, by using Lemma 3.24, we prove that  $\eta$  is consistent with  $\text{build}(h', p', T @ \rho)$ , where  $(h', p') = \text{copy}(h', p, j)$ .

- **Case  $e \equiv C \bar{a}_i^n @ r$**

We assume that  $\Gamma' \vdash C \bar{a}_i^n @ r : T @ \rho$  with  $\Gamma'(r) = \rho$  and  $E(r) = j$ . We get:

$$\begin{aligned} \text{build}^*(h, E, \Gamma) &\supseteq [\rho \mapsto j] \uplus \text{build}(h, E(a_1), \Gamma'(a_1)) \uplus \dots \uplus \text{build}(h, E(a_n), \Gamma'(a_n)) \\ &= [\rho \mapsto j] \uplus \text{build}(h', E(a_1), \Gamma'(a_1)) \uplus \dots \uplus \text{build}(h', E(a_n), \Gamma'(a_n)) \end{aligned}$$

The last equality is due to  $h'$  being a strict extension of  $h$ , and Lemma 3.22 (2). All the components of the right-hand side of this equality are pairwise consistent, and also consistent with  $\eta$ . Therefore, the lemma holds, as the result of  $\text{build}(h, v, s)$  is just this right-hand side.

- **Case  $e \equiv g \bar{a}_i^n @ \bar{r}_j^m$**

We assume that  $\Gamma'(g) = \forall \bar{\alpha} \bar{\rho}. \bar{t}_{g,i}^n \rightarrow \bar{\rho}_{g,j}^m \rightarrow s_g$  and that  $g \bar{x}_i^n @ \bar{r}_j^m = e_g \in \Sigma$ . Since  $\Gamma$  is correct with respect to  $\Sigma$  (and hence  $\Gamma'$  also is) we can apply the [FUN] rule, so that  $\{\Gamma_0 \setminus g\} g \bar{x}_i^n @ \bar{r}_j^m = e_g \{\Gamma_0\}$  for some environment  $\Gamma_0$  containing the type of  $g$ . Thus we get:

$$\Gamma_0 \vdash e_g : s_g \quad \text{where } \Gamma_0 \supseteq [\bar{x}_i : \bar{t}_{g,i}^n, \bar{r}_j : \bar{\rho}_{g,j}^m, \text{self} : \rho_{\text{self}}^g] \quad (3.8)$$

The region type  $\rho_{\text{self}}^g$  is a fresh variable, so we can ensure that is different from the  $\bar{\rho}_{g,j}^m$ , from the  $\bar{\Gamma}'(\bar{r}_j)^m$ , and from any variable appearing in the domain of  $\eta$ . From one of the premises of the

[APP] rule we can assume the existence of a type substitution  $\theta$  such that  $\theta(\overline{t_{g,i}}^n \rightarrow \overline{\rho_{g,j}}^m \rightarrow s_g) = \overline{\Gamma'(a_i)}^n \rightarrow \overline{\Gamma'(r_j)}^m \rightarrow s$  and that  $\rho_{self}^g \notin \text{dom } \theta$ . We apply Lemma 3.9 to (3.8) so as to obtain:

$$\Gamma_g \vdash e_g : s \quad \text{where } \Gamma_g = \theta(\Gamma_0) \supseteq [\overline{x_i : \Gamma'(a_i)}^n, \overline{r'_j : \Gamma'(r_j)}^m, self : \rho_{self}^g]$$

Assume the following execution of the function's body:

$$E_g \vdash h, k+1, e_g \Downarrow h^+, k+1, v$$

where  $h' = h^+|_k$  and  $E_g \stackrel{\text{def}}{=} [\overline{x_i \mapsto E(a_i)}^n, \overline{r'_j \mapsto E(r_j)}^m, self \mapsto k+1]$ . In order to apply the induction hypothesis, we have to prove that  $build^*(h, E_g, \Gamma_g)$  is well-defined and consistent with  $\eta$ . This is a consequence of  $build^*(h, E, \Gamma)$  having these properties and  $\rho_{self}^g$  being a fresh variable:

$$\begin{aligned} build^*(h, E_g, \Gamma_g) &= \left( \bigoplus_{i=1}^n build(h, E_g(x_i), \Gamma_g(x_i)) \right) \uplus \left( \bigoplus_{j=1}^m build(h, E_g(r'_j), \Gamma_g(r'_j)) \right) \uplus [\rho_{self}^g \mapsto k+1] \\ &= \left( \bigoplus_{i=1}^n build(h, E(a_i), \Gamma'(r_j)) \right) \uplus \left( \bigoplus_{j=1}^m build(h, E(r_j), \Gamma'(r_j)) \right) \uplus [\rho_{self}^g \mapsto k+1] \end{aligned}$$

Therefore we obtain, by induction hypothesis, a region instantiation  $build(h^+, v, s)$  which is well-defined and consistent with  $\eta$ . The result of  $build(h', v, s)$  also has these properties, by Lemma 3.22 (1).

- **Case  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$**

We assume that  $\Gamma' = \Gamma_1 \sqcup \Gamma_2$ ,  $\Gamma_1 \vdash e_1 : s_1$ ,  $\Gamma_2 + [x_1 : \tau_1] \vdash e_2 : s_2$  and the following executions:

$$E \vdash h, k, e_1 \Downarrow h_1, k, v_1$$

$$E \uplus [x_1 \mapsto v_1] \vdash h_1, k, e_2 \Downarrow h', k, v$$

From the assumptions on  $build^*(h, E, \Gamma)$  we can prove that both  $build^*(h, E, \Gamma_1)$  and  $build^*(h, E, \Gamma_2)$  are well-defined and consistent with  $\eta$ , and so is  $build^*(h_1, E, \Gamma_2)$ , because of Lemma 3.23. By the induction hypothesis,  $build(h_1, v_1, s_1)$  is well-defined and consistent with  $\eta$  and hence the same applies to  $build^*(h_1, E \uplus [x_1 \mapsto v_1], \Gamma_2 + [x_1 : \tau_1])$ . Again, the induction hypothesis on the derivation of  $e_2$  gives the required result.

- **Case  $e \equiv \text{case } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i}^n$**

Let us assume that  $E(x) = p$  and that  $h(p) = (j, C_r \overline{v_j^{n_r}})$ . That is, the  $r$ -th branch is executed:

$$E \uplus [\overline{x_{rj} \mapsto \overline{v_j^{n_r}}}] \vdash h, k, e_r \Downarrow h', k, v$$

By the typing rule for **case** expressions, we obtain  $\Gamma' + [\overline{x_{rj} : \tau_{rj}^{n_r}}] \vdash e_r : s$  where  $\overline{s_j^{n_r}} \rightarrow \rho \rightarrow T @ \rho \leq \Gamma(C_r)$  and  $utype?( \tau_{r_i}, s_j )$  for every  $j \in \{1 \dots n_r\}$ . If  $\Gamma(x) = T @ \rho$ , the region instantiation  $build(h, p, T @ \rho)$  is contained within  $build^*(h, E, \Gamma)$  and hence is consistent with  $\eta$ . Let us unfold

its definition:

$$\begin{aligned} \text{build}(h, p, T @ \rho) &= [\rho \mapsto j] \uplus \text{build}(h, v_1, s_1) \uplus \dots \uplus \text{build}(h, v_{n_r}, s_{n_r}) \\ &= [\rho \mapsto j] \uplus \text{build}(h, v_1, \tau_{r1}) \uplus \dots \uplus \text{build}(h, v_{n_r}, \tau_{r,n_r}) \end{aligned}$$

Each one of the region instantiations of the right-hand side is well-defined and consistent with  $\eta$ . Therefore the result of  $\text{build}^*(h, E \uplus [\overline{x_{rj}} \mapsto \overline{v_j^{n_r}}], \Gamma' + [\overline{x_{rj}} : \tau_{rj}^{n_r}])$  is also well-defined and consistent with  $\eta$ , and we can apply the induction hypothesis on the derivation of the evaluation of  $e_r$ , which leads to the desired result.

- **Case**  $e \equiv \text{case! } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i^n}$

The proof is very similar to that of the non-destructive **case**. The only difference is the fact that  $p$  is now a dangling pointer, but we still can prove that  $\text{build}^*(h \setminus p, E \uplus [\overline{x_{rj}} \mapsto \overline{v_j^{n_r}}], \Gamma' + [\overline{x_{rj}} : \tau_{rj}^{n_r}])$ , because of Lemma 3.22 (1).

□

To sum up, we have set up a correspondence between the actual regions where a DSs resides and the RTVs assigned by the type system: if two variables have the same outer region  $\rho$  at runtime, the cells bound to them at runtime will live in the same actual region. Since the type system (see rule [FUN] in Figure 3.9) demands the  $\rho_{\text{self}}$  variable not to occur in the type of the function's result, every DS returned by the function call does not have cells in its working region (assuming  $\text{self}$  is the only variable pointing to the topmost region, and the only variable of type  $\rho_{\text{self}}$ ). This implies that the deallocation of the  $k + 1$ -th region at the end of a function call does not generate dangling pointers. This will be detailed in the correctness proof of Section 3.6.4.

### 3.6.2 Reachability and harmless semantics

The concepts introduced so far are related to region-based memory management. The next sections on the type system's correctness are devoted to the safety of explicit deallocation via **case!** expressions. A crucial aspect in this part of the proof is the notion of *sharing* between DSs. Two DSs are said to share if there exists a cell in the heap that is reachable from both. Therefore, in order to formally define sharing we have to specify a notion of *reachability* in the heap. It was already pointed out in Section 3.4 that the connections between the cells in a heap can be represented by means of a directed acyclic graph. Thus, our notion of cell reachability is directly related to the notion of reachability in DAGs. However, for the purposes of the type system it is necessary to distinguish the cells that are reachable from a given one by following those pointers occurring in the recursive positions of the constructors being traversed (which we shall call their *recursive descendants*). This motivates the following definition:

**Definition 3.27.** Given a heap  $h$ , we define the child ( $\rightarrow_h$ ) and recursive child ( $\rightarrow_h^*$ ) relations on heap pointers as follows:

$$\begin{aligned} p \rightarrow_h q &\stackrel{\text{def}}{=} h(p) = (j, C \overline{v_i^n}) \wedge q \in \overline{v_i^n} \\ p \rightarrow_h^* q &\stackrel{\text{def}}{=} h(p) = (j, C \overline{v_i^n}) \wedge q = v_i \\ &\quad \text{for some } i \in \text{RecPos}(C) \end{aligned}$$

The reflexive transitive closure of these relations is respectively denoted by  $\rightarrow_h^*$  and  $\rightarrow_h^*$ . If  $p \rightarrow_h^* q$  we say that  $q$  is *reachable* from  $p$  in  $h$ . If  $p \rightarrow_h^* q$  we say that  $q$  is a *recursive descendant* of  $p$  in  $h$ . These definitions serve as a basis for specifying the following notions of reachability:

**Definition 3.28.**

$\mathit{closure}(h, p)$	$\stackrel{\text{def}}{=} \{q \mid p \rightarrow_h^* q\}$	Set of locations reachable in $h$ by location $p$ .
$\mathit{live}(h, E, L)$	$\stackrel{\text{def}}{=} \bigcup_{x \in L} \mathit{closure}(h, E(x))$	Live part of $h$ , i.e. reachable from any free variable.
$\mathit{recReach}(h, p)$	$\stackrel{\text{def}}{=} \{q \mid p \rightarrow_h^* q\}$	Set of recursive descendants of $E(x)$ including itself.
$\mathit{closed}(h, p)$	$\stackrel{\text{def}}{=} \mathit{closure}(h, p) \subseteq \text{dom } h$	If there are no dangling pointers in $\mathit{closure}(h, p)$ .
$\mathit{closed}(h, E, L)$	$\stackrel{\text{def}}{=} \forall x \in L. \mathit{closed}(h, E(x))$	If there are no dangling pointers in $\mathit{live}(h, E, L)$ .

As a convention, we define  $\mathit{closure}(h, c) = \emptyset$  for any heap  $h$  and literal  $c$ . There are several general properties of the language regarding closures. The first one specifies that the closure of a given pointer can decrease during the execution of a program as a consequence of destruction, but it never increases.

**Lemma 3.29.** *Let  $E \vdash h, k, e \Downarrow h', k, v$  be an execution of an expression. Then, for every pointer in  $\text{dom } h$ ,  $\mathit{closure}(h, p) \supseteq \mathit{closure}(h', p)$ .*

*Proof.* By induction on the  $\Downarrow$ -derivation. If  $p$  does not belong to the domain of  $h'$ , then  $\mathit{closure}(h', p) = \emptyset$  and the lemma holds trivially, so let us assume  $p \in \text{dom } h'$ . All cases are straightforward: In rules [Lit], [Var] and [PrimOp] we get  $h = h'$ . In [Copy] and [Cons] only fresh cells are created, and these cells cannot be reachable from those already existing in  $h$ . In [App], [Let], [Case], [Case!] the result follows directly from the induction hypothesis. The only case in which the closure strictly decreases is when removing the location pointed to by the discriminant of a **case!** in [Case!] rule.  $\square$

The second property shows that, if the evaluation of an expression returns a pointer belonging to the initial heap, this pointer must have been accessed through (at least) one of the free variables in that expression. The same applies to the pointers being removed from the initial heap.

**Lemma 3.30.** *Let us assume that  $E \vdash h, k, e \Downarrow h', k, v$ .*

1. *If  $v$  is a pointer in  $\text{dom } h$ , then  $v \in \mathit{live}(h, E, \mathit{fv}(e))$ .*
2. *For every  $p \in \text{dom } h \setminus \text{dom } h'$ ,  $p \in \mathit{live}(h, E, \mathit{fv}(e))$ .*

*Proof.* By induction on the size of the  $\Downarrow$ -derivation. The only base case applicable to (1) is the [Var] rule, in which the result follows trivially. In the recursive cases (1) follows trivially from the induction hypothesis. With regard to (2) the only relevant case is that of [Case!] rule, since none of the pointers removed in the [App] rule as a consequence of deallocating the topmost region, belong to the initial heap. However, the case of the [Case!] rule is also trivial, as the discriminant is a free variable referring to the pointer being removed.  $\square$

Finally, the next property shows that only the live part of the heap is relevant when evaluating an expression. The rest of the heap can be safely discarded from the  $\Downarrow$ -judgements.

**Lemma 3.31.** *Let us assume an execution  $E \vdash h, k, e \Downarrow h', k, v$  and  $p$  a pointer in  $\text{dom } h$ . If  $p \notin \mathit{live}(h, E, \mathit{fv}(e))$ , then we can discard the binding of  $p$  in  $h$  so as to obtain  $E \vdash h \setminus p, k, e \Downarrow h' \setminus p, k, v$ , and this derivation has the same size as the original one.*

*Proof.* By induction on the size of the  $\Downarrow$ -derivation. All cases are straightforward, except in the case in which the last rule applied is [Let]. In this case the first property of Lemma 3.30 is necessary after applying the induction hypothesis to the bound expression. After this, the result holds trivially.  $\square$

In Figure 2.14 we showed the big-step operational semantic rules of *Core-Safe* and we identified the cases in which none of the rules could be applicable for some initial heap  $h$ , expression  $e$  and runtime environment  $E$ . One of these situations is produced when accessing a location  $p$  that does not belong to the domain of the heap (that is, a dangling pointer). We want to show that this situation never takes place when the expression is well-typed w.r.t. the type system shown in this chapter, but we are not concerned about the rest of the pathological cases explained in Section 2.4.1 (pattern-matching errors, non-termination, etc.). For this reason we assume the execution of an expression under a *harmless* big-step operational semantics that does not remove pointers from the heap, thus avoiding the possibility of having dangling pointers. We use the  $\Downarrow^*$  notation to refer to the judgements of this semantics. In this way, given  $E, h, k, e$  of their respective types, if there does not exist a final heap  $h'$  and a value  $v$  such that the judgement  $E \vdash h, k, e \Downarrow^* h', k, v$  is not derivable, we can ensure that this is not due to an access to a dangling pointer, but to another of the reasons explained in Section 2.4.1. On the contrary, in the following sections we will show that  $\Downarrow^*$ -judgements imply  $\Downarrow$ -judgements in the context of well-typed programs (see Appendix C).

The rules of the harmless semantics are shown in Figure 3.12. Notice the similarity with those of Figure 2.14. The only difference is that the topmost region is not removed after executing the function's body in the *[App]* rule, and that the binding pointed to by the discriminant of a destructive **case!** is not removed from the heap. Thus, a destructive **case!** behaves just like its non-destructive counterpart.

It is easy to see that whenever we have a judgement  $E \vdash h, k, e \Downarrow^* h', k, v$ , we get  $h' \supseteq h$ , since nothing is destroyed during  $\Downarrow^*$ -evaluations. Moreover, Lemmas 3.29, 3.30 and 3.31 also hold for  $\Downarrow^*$ -judgements. In particular, we get equality in Lemma 3.29:  $\text{closure}(h, p) = \text{closure}(h', p)$ , and Lemma 3.30(2) holds vacuously.

### 3.6.3 Correctness of the sharing analysis

Recall that our type system depends on some functions and predicates (namely, *sharerrec* and *isTree*) which are defined by some auxiliary analyses whose definition are beyond the scope of this thesis. In particular:

*sharerrec*( $x, e$ ) Approximates the sharing relations between DSs. It contains all the variables in the scope of  $e$  that, at runtime, may point to a recursive descendant of  $x$ .

*isTree*( $x$ ) Approximates the internal sharing property of a DS. This predicate holds for all the variables that, at runtime, point to a DS whose associated graph is a tree.

Although we do not explain the analyses themselves (see [98] for more details) we need to formalize their safety properties in order to prove the correctness of the type system. In other words, we have to determine which properties must satisfy a sharing analysis and an internal sharing analysis so that, when connecting their results to our system, the correctness of the latter is not compromised.

Let us start with the *isTree* predicate. Firstly we have to precise the notion of a DS having internal sharing at runtime. The following definition allows us to describe the path of pointers in the heap that must be taken if we start in a pointer  $p$  and we end in a recursive descendant  $q$  of  $p$ .

**Definition 3.32.** Given two pointers  $p$  and  $q$  in a heap  $h$ , a sequence of natural numbers  $[i_1, \dots, i_m]$  is said to describe the relation  $p \rightarrow_h^* q$  iff there exists a sequence of pointers  $[p_1, \dots, p_m, p_{m+1}] \subseteq \text{dom } h$  and a  $C \in \mathbf{Cons}$  such that:

1.  $p = p_1$  and  $q = p_{m+1}$ .

$$\begin{array}{c}
\frac{}{E \vdash h, k, c \Downarrow^* h, k, c} \text{ [Lit*]} \\
\\
\frac{}{E[x \mapsto v] \vdash h, k, x \Downarrow^* h, k, v} \text{ [Var*]} \\
\\
\frac{}{E \vdash h, k, a_1 \oplus a_2 \Downarrow^* h, k, E(a_1) \oplus E(a_2)} \text{ [PrimOp*]} \\
\\
\frac{j \leq k \quad (h', p') = \text{copy}(h, p, j)}{E[x \mapsto p, r \mapsto j] \vdash h, k, x @ r \Downarrow^* h', k, p'} \text{ [Copy*]} \\
\\
\frac{E(r) \leq k \quad \text{fresh}_h(p)}{E \vdash h, k, C \bar{a}_i^n @ r \Downarrow^* h \uplus [p \mapsto (E(r), C \bar{E}(a_i)^n)], k, p} \text{ [Cons*]} \\
\\
\frac{(g \bar{y}_i^n @ \bar{r}_j^m = e_g) \in \Sigma \quad \overline{[y_i \mapsto E(a_i)^n, r_j' \mapsto E(r_j)^m, \text{self} \mapsto k + 1]} \vdash h, k + 1, e_g \Downarrow^* h', k + 1, v}{E \vdash h, k, g \bar{a}_i^n @ \bar{r}_j^m \Downarrow^* h', k, v} \text{ [App*]} \\
\\
\frac{E \vdash h, k, e_1 \Downarrow^* h_1, k, v_1 \quad E \uplus [x_1 \mapsto v_1] \vdash h_1, k, e_2 \Downarrow^* h', k, v}{E \vdash h, k, \text{let } x_1 = e_1 \text{ in } e_2 \Downarrow^* h', k, v} \text{ [Let*]} \\
\\
\frac{E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, e_r \Downarrow^* h', k, v}{E[x \mapsto p] \vdash h[p \mapsto (j, C_r \bar{v}_i^{nr}), k, \text{case } x \text{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i \Downarrow^* h', k, v]} \text{ [Case*]} \\
\\
\frac{E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, e_r \Downarrow^* h', k, v}{E[x \mapsto p] \vdash h[p \mapsto (j, C_r \bar{v}_i^{nr}), k, \text{case! } x \text{ of } \bar{C}_i \bar{x}_{ij}^{ni} \rightarrow e_i \Downarrow^* h', k, v]} \text{ [Case!*]}
\end{array}$$

Figure 3.12: Harmless operational semantics of *Core-Safe* expressions. The only difference with respect to the standard big-step semantics of Figure 2.14 is the absence of destruction in the cases of function application and destructive **case!**.

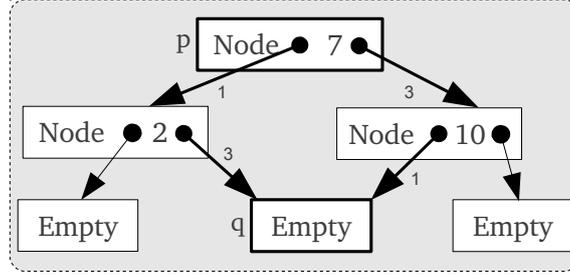


Figure 3.13: DS with internal sharing: there are two paths from  $p$  to  $q$ .

2. For every  $j \in \{1..m\}, i_j \in \text{RecPos}(C)$ .
3. For every  $j \in \{1..m\}, h(p_j) = (l, C \bar{v}_k^n)$  for some  $l \in \mathbb{N}, \bar{v}_k^n \subseteq \mathbf{Val}$ , and  $v_{i_j} = p_{j+1}$ .

**Example 3.33.** Given heap  $h$  of the Figure 3.13, the relation  $p \rightarrow_h^* q$  is described by the sequence  $[1, 3]$ , but also by the sequence  $[3, 1]$ .  $\square$

If a recursive descendant of  $p$  can be reached from that pointer by following two different paths, there is internal sharing (confluence) in the DS pointed to by  $p$ .

**Definition 3.34.** Given two pointers  $p$  and  $p'$  in a heap  $h$ , we say that there is a *confluence* from  $p$  to  $p'$  (denoted  $p \rightrightarrows_h^* p'$ ) iff  $p \rightarrow_h^* p'$  and this relation can be described by two distinct sequences of natural numbers.

**Definition 3.35.** A pointer in a heap  $p$  is said to be a *tree* in  $h$  if, for every  $p' \in \text{recReach}(h, p)$ :  $p \not\rightleftharpoons_h^* p'$ .

**Example 3.36.** The DS pointed to by  $p$  in Figure 3.13 is not a tree, as there is a confluence from  $p$  to  $q$ .  $\square$

Once we have made the notion of internal sharing precise, we can define the correctness of an internal sharing analysis. A sensible correctness statement for such an analysis would be the following:

*Given an execution of an expression  $E \vdash h, k, e \Downarrow h', k, v$ , for every variable  $x \in \text{dom } E$  such that  $E(x)$  is not a tree in  $h$ , the internal sharing analysis determines that the predicate  $\text{isTree}(x)$  does not hold.*

This definition, although reasonable, is too strong for what the type system demands. One can develop an analysis carried out at a function definition level that *assumes that the DSs pointed to by its formal parameters are trees*, and its results would be suitable for the type system. The rationale behind this is the fact that the type system only needs the *isTree* predicate for checking that destructive pattern matching is done to DSs without internal sharing (otherwise the program might destroy a recursive descendant of a DS twice). So, assume a given variable occurs in the discriminant of a **case!**:

1. If that variable is not a formal parameter of the function, the internal sharing analysis will determine whether that variable is a tree or not, and update the sharing information of the pattern variables of the corresponding branches accordingly.
2. If that variable is a formal parameter of the function, the DS to which it points might not be a tree, but the internal sharing analysis would not warn the type system about this, since the analysis assumes the formal parameters to be trees. However, a parameter appearing in a **case!** would

have a condemned type, and the rule for typing function applications [APP] ensures that the arguments of condemned positions are trees, so the assumption of the internal sharing analysis is correct.

Given the above, the correctness of an internal sharing analysis that only approximates the internal sharing property when the type system needs it, would be defined as follows.

**Assumption 3.37.** *Given an expression  $e$  being executed in the context of a function  $f$ .*

$$E \vdash h, k, e \Downarrow^* h', k, v$$

*Let  $P$  be the set of formal parameters of this function. For every  $x \in \text{dom } E$  and pointer  $p$  such that  $E(x) \Rightarrow_h^* p$ , at least one of the following properties hold:*

$$(TREE) \quad \neg \text{isTree}(x)$$

$$(CONF) \quad \exists y_0 \in P. y_0 \Rightarrow_h^* p$$

In other words, if there is internal sharing at runtime, either the analysis warns the type system about it, or this internal sharing occurs in a DS pointed to by a function parameter.

We follow a similar pattern with the *sharerec* function defined by the sharing analysis. One would expect the following correctness property:

*Given an execution of an expression  $E \vdash h, k, e \Downarrow h', k, v$ , for every pair of variables  $x, y \in \text{dom } E$  such that  $\text{recReach}(h, E(x)) \cap \text{closure}(h, E(y)) \neq \emptyset$  the sharing analysis determines that  $y \in \text{sharerec}(x, e)$ .*

But, again, the type system does not need such a strong property. A sharing analysis that approximates this information while assuming that the parameters of the function to which  $e$  belongs do not share with each other would be suitable. The sharing analysis is only needed to determine the variables pointing to a recursive descendant of something being destroyed. So, whenever we have two variables  $x$  and  $y$  such that  $\text{recReach}(h, E(x)) \cap \text{closure}(h, E(y)) \neq \emptyset$  and  $x$  is being destroyed, there are three possibilities:

1. This sharing relation is captured by the sharing analysis, so we get  $y \in \text{sharerec}(x, e)$ .
2. The sharing analysis does not warn the type system about this sharing relation, because the latter is due to the sharing between different parameters which the analysis assumed to be disjoint, but actually they are not (Figure 3.14). This case does not compromise the safety of the type system, since the function application rule [APP] checks that the DSs pointed to by the arguments occurring in condemned positions do not share between them, or with the remaining parameters. Since  $x$  is being destroyed, at least one of these parameters to which the sharing is due is condemned.
3. The sharing analysis does not warn the type system about this sharing relation, since it is due to a parameter from which  $x$  and  $y$  are recursive descendants (Figure 3.15). As a consequence, this parameter would have internal sharing, which the internal sharing analysis does not report, as it was explained above. Again, this does not compromise the safety of the type system, since that parameter would be condemned.

These three possibilities are formalized as follows.

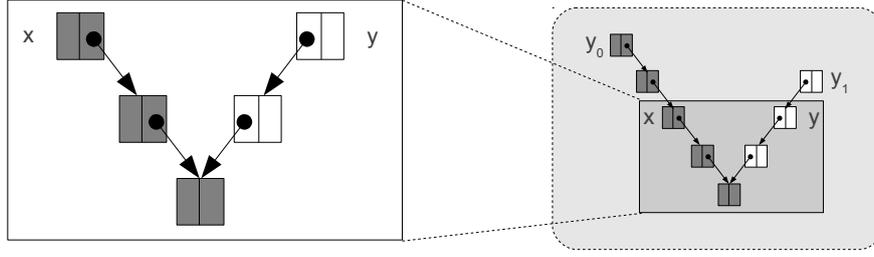


Figure 3.14: If  $y$  points to a recursive descendant of  $x$ , but this relation is not reported by the sharing analysis, it may be due to sharing between some parameters  $y_0, y_1$  in the context function.

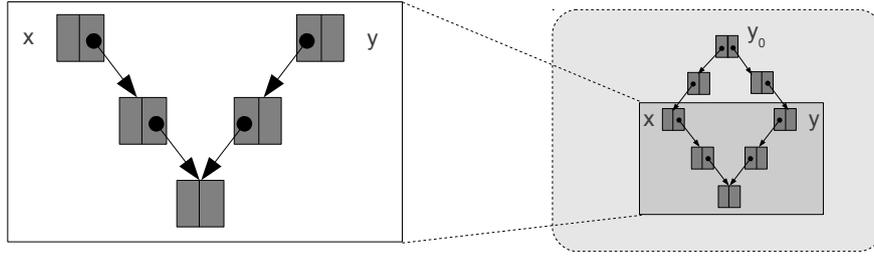


Figure 3.15: If  $y$  points to a recursive descendant of  $x$ , but this relation is not reported by the sharing analysis, it may be due to a parameter  $y_0$  with internal sharing.

**Assumption 3.38.** Given an expression  $e$  being executed in the context of a function  $f$ .

$$E \vdash h, k, e \Downarrow^* h', k, v$$

Let  $P$  be the set of formal parameters of this function. For every  $x, z \in \text{dom } E$  and pointer  $p$  such that  $p \in \text{closure}(h, E(x)) \cap \text{recReach}(h, E(z))$ , at least one of the following properties hold:

- (SHR)  $x \in \text{sharerec}(z, e)$
- (SHP)  $\exists y_0, y_1 \in P. y_0 \neq y_1 \wedge E(y_0) \rightarrow_h^* p \wedge E(y_1) \rightarrow_h^* p$
- (CONF)  $\exists y_0 \in P. y_0 \rightrightarrows_h^* p$

The sharing relations that take place between parameters deserve special attention. Assume that a parameter  $y_0$  points to a recursive descendant of a local variable  $x$  in the context of an expression  $e$ . As it was explained before, the sharing analysis might not reflect this sharing relation, because the DS pointed to by this parameter  $y_0$  is the same as the DS pointed to by another hypothetical parameter  $y_1$ , both of which are assumed to be disjoint by the sharing analysis. However, and since  $x$  is a local variable which has been defined within the context function definition, the sharing analysis should have enough information to determine that at least one of these parameters belongs to  $\text{sharerec}(x, e)$ . That is why the following assumption.

**Assumption 3.39.** Assume the execution of an expression in the context of  $f$ :

$$E \vdash h, k, e \Downarrow^* h', k, v$$

Let  $P$  be the set of formal parameters of  $f$ . For every  $x \in \text{dom } E$  and pointer  $p \in \text{recReach}(h, E(x)) \cap \text{closure}(h, E(y_0))$  for some  $y_0 \in P$ , then there exists a parameter  $y_1 \in P$  such that  $p \in \text{recReach}(h, E(x)) \cap$

$\text{closure}(h, E(y_1))$  and  $y_1 \in \text{sharerec}(x, e)$ .

Notice that, in the case in which  $x$  is not a local variable, but a formal parameter of  $f$ , this assumption holds trivially (we would get  $y_1 = x$ ). The sharing analysis currently implemented in the compiler (and described in [98]) satisfies all these assumptions.

### 3.6.4 Preservation of closedness

The last step is to put together the definitions and results presented in last sections in order to prove the correctness of the type system. The essential fact we aim to prove is the following:

*If we can derive  $E \vdash h, k, e \Downarrow^* h', k, v$  and  $e$  is well-typed, then we can derive  $E \vdash h, k, e \Downarrow h'', k, v$  for some  $h'' \subseteq h'$ .*

In operational terms, this means that the execution of the program cannot fail because of cell destruction. However, we have to impose some constraints on our initial environment  $E$  and heap  $h$  under which the  $\Downarrow^*$ -evaluation is done. Otherwise the claim above might not hold, for instance, when some of the variables occurring free in  $e$  contains a dangling pointer in our initial state. In particular, we demand that the the live part of the heap at the beginning of the evaluation is closed, that is,  $\text{closed}(h, E, \text{fv}(e))$ .

Besides this, some additional constraints are needed on the parameters of the function to which the expression  $e$  belongs. As it was explained in last section, the sharing analysis is not required to capture all the sharing that takes place at runtime, but only the sharing that takes place under certain circumstances (namely, when a variable points to a recursive descendant of a condemned one). The same happens with the internal sharing analysis: it is not required to report the internal sharing occurring in function parameters. The disjointness property shown below specifies these conditions.

**Definition 3.40** (Disjointness property). A value environment  $E$ , a heap  $h$ , and a typing environment  $\Gamma_0$  satisfy the disjointness property (written  $P_{\text{disj}}(E, h, \Gamma_0)$ ) iff

1.  $\Gamma_0$  does not contain variables mapped to in-danger types.
2. For every pair of distinct variables  $y_0, y_1 \in \text{dom } \Gamma_0$  such that  $\Gamma_0(y_0) \in \mathbf{CdmType}$ , it holds that  $\text{recReach}(h, E(y_0)) \cap \text{closure}(h, E(y_1)) = \emptyset$ .
3. For every variable  $y_0 \in \text{dom } \Gamma_0$  such that  $\Gamma_0(y_0) \in \mathbf{CdmType}$ ,  $E(y_0)$  is a tree.

Assume  $\Gamma_0$  contains the types of the parameters of a function, whose body is typeable under this environment. The first condition follows trivially from the [FUN] rule of the type system, which does not allow parameters having in-danger types. The second condition specifies that the recursive spine of condemned parameters cannot share with any other parameter. The last one prevents the DSs pointed to by condemned parameters from having internal sharing. In the correctness proof below we have to prove that the disjointness property is an invariant which is propagated along the execution.

**Example 3.41.** Assume a function definition with three parameters  $y_0, y_1$  and  $y_2$ . The first two parameters have a safe type, whereas  $y_2$  has a condemned type. Consider the three configurations shown in Figure 3.16. The first configuration does not satisfy the disjointness property, since  $y_1$  points to the recursive closure of a condemned parameter ( $y_2$ ). The second configuration does not satisfy the disjointness property, as  $y_2$  is a condemned parameter with internal sharing. Finally, the third configuration does satisfy the disjointness property.  $\square$

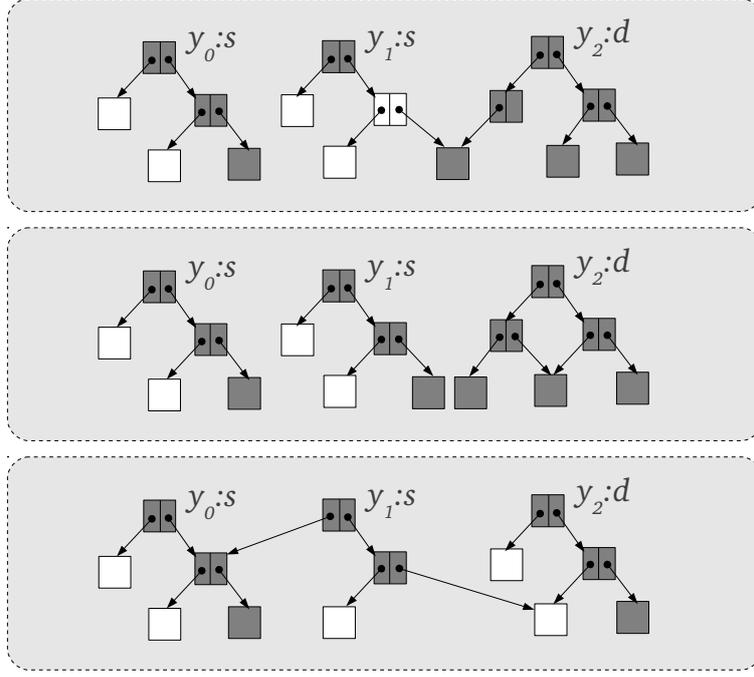


Figure 3.16: Sharing between the parameters of a function. The first two parameters  $y_0$  and  $y_1$  are safe, whereas  $y_2$  is condemned. Gray cells represent the recursive spine of each parameter.

Our next step is to define the invariants that are propagated at the level of expressions, i.e. from a  $\Downarrow^*$ -judgement to the  $\Downarrow^*$ -judgements in its derivation.

**Definition 3.42.** A initial configuration  $(\Gamma, E, h, k, L, \Gamma_0)$  is said to be good whenever:

1.  $closed(h, E, L)$ .
2. The heap  $h$  is acyclic. That is, there is no pointer  $p$  in  $h$  such that  $p \rightarrow_h^+ p$ .
3.  $P_{disj}(E, h, \Gamma_0)$  holds.
4.  $build^*(h, E, \Gamma)$  is well-defined.
5.  $E(self) = k$ , and for every other region variable  $r \in \text{dom } E$ ,  $E(r) < k$ .
6. If  $self \in \text{dom } \Gamma$ , then  $\Gamma(self) = \rho_{self}$ . For every other RTV  $r \in \text{dom } \Gamma$ ,  $\Gamma(r) \neq \rho_{self}$ .

In the context of the proof, the  $\Gamma$  denotes an environment typing the expression being evaluated, and  $\Gamma_0$  an environment typing the body of the function to which the expression being evaluated belongs. The first condition has been introduced above. It is easy to see that the second condition is an invariant, since new cells are constructed only in fresh locations of the heap, and the language does not allow in-place update (Lemma 3.29). The disjointness property (third condition) is propagated between function calls, but it is also preserved during the execution of a given function call, since it refers to the parameters of the context function, which remain constant as this call is evaluated. The fourth condition refers to the consistency of the initial configuration. The fact that consistency is an invariant preserved during evaluation has already been proved in Theorem 3.26. The last two conditions state that  $self$  is the only region variable of type  $\rho_{self}$ , and the only region variable pointing to the topmost region of the heap. These conditions are aimed at proving that the topmost region deallocation does not generate dangling pointers. The propagation of the latter property is a consequence of Proposition 2.15.

The following definition states some properties that hold if the evaluation starts from a good configuration.

**Definition 3.43.** A final configuration  $(s, v, h)$  is said to be good whenever  $build(h, v, s)$  is well-defined and  $closed(h, v)$  holds.

The correctness theorem of the type system shows that the evaluation of an expression transforms good initial configurations into good final configurations. It also proves that  $\Downarrow^*$ -evaluation of well-typed expressions implies  $\Downarrow$ -evaluation. Before proving that theorem, we need the following technical result:

**Lemma 3.44.** *Assume a judgement  $\Gamma_0 + [x : \tau] \vdash e_0 : s_0$ . For every judgement  $\Gamma \vdash e : s$  occurring in its derivation, if  $x \in \text{dom } \Gamma$  then  $\Gamma(x) \leq \tau$ .*

*Proof.* By induction on the depth of the  $\Gamma \vdash e : s$  w.r.t. the judgement  $\Gamma_0 + [x : \tau] \vdash e_0 : s_0$ . All cases are straightforward.  $\square$

Basically, this result states that when we move from the leafs of the typing derivation to its root, the variables in the typing environment can become weaker, but not stronger. Now we can prove the correctness theorem.

**Theorem 3.45 (Correctness of type system).** *For every  $E, h, h', k, e, v, \Gamma_0, \Gamma, s_0, s, L$  of their respective types, and every function definition  $f \bar{x}_i @ \bar{r}_j = e_f$ , if  $\Gamma_0 \vdash e_f : s_0$  and  $e$  is a sub-expression of  $e_f$  such that:*

1.  $E \vdash h, k, e \Downarrow^* h', k, v$ .
2.  $\Gamma \vdash e : s$  belongs to the derivation of  $\Gamma_0 \vdash e_f : s_0$ .
3.  $(\Gamma, E, h, k, L, \Gamma_0)$  is good, being  $L = fv(e)$ .

*it holds that:*

1. There exists a heap  $h'' \subseteq h'$  such that  $E \vdash h, k, e \Downarrow h'', k, v$ .
2. For every pointer  $p \in \text{dom } h \setminus \text{dom } h''$ , and every variable  $z \in \text{dom } E$  such that  $E(z) \rightarrow_h^* p$  we obtain  $\Gamma(z) \in \mathbf{UnsafeType}$ .
3. For every pointer  $p \in \text{dom } h \setminus \text{dom } h''$ , and every variable  $z \in \text{dom } E$  such that  $E(z) \rightarrow_h^* p$ , but  $E(z) \not\rightarrow_h^* p$  we get  $\Gamma(z) \in \mathbf{DgrType}$ .
4.  $(s, v, h'')$  is good.

*Remark.* Before getting into the details of the proof, let us briefly explain the meaning of the second and third conclusions of the theorem. Both are related with the intended semantics of safe, condemned, and in-danger types that was explained in Section 3.3. The second condition states that, if part of the closure of a variable disappears from the heap during the evaluation of  $e$ , that variable must appear in  $\Gamma$  with an unsafe type. The third condition is more specific: if part of the closure of a variable *beyond its recursive spine* is destroyed during the evaluation of  $e$ , that variable must appear in the environment  $\Gamma$  with an in-danger type.

*Proof.* By induction on the size of the  $\Downarrow^*$  derivation. We distinguish cases on the structure of expression  $e$ . In each subcase, we can assume the existence of a  $\Gamma'$  with  $\text{dom } \Gamma' \subseteq \text{dom } \Gamma$ , such that  $\Gamma' \vdash e : s$  and the last rule applied in this derivation is neither [EXT] nor [WEAK]. We can easily prove that, for every  $x \in \text{dom } \Gamma'$ , if  $\Gamma'(x)$  is unsafe, so is  $\Gamma(x)$ . Moreover, if  $\Gamma'(x)$  is in-danger, so is  $\Gamma(x)$ .

One of the key points of the proof is the fact that the goodness property is propagated through the evaluation tree. When proving this we shall leave out the proof of  $h$  being acyclic during the evaluation of  $e$ . This is a general property of the language. The judgement  $P_{disj}(E, h, \Gamma_0)$  also propagates through the evaluation in the current context function, since the closure of the parameters remains unchanged (see Lemma 3.29). Hence we omit the proof of the  $P_{disj}$  propagation, except in the case of function application, where a change of context takes place. The propagation of consistency is also a consequence of Theorem 3.26, and we shall not reproduce its proof again. The fact that *self* is the only region pointing to topmost region in the heap is propagated by Proposition 2.15. The last condition on the definition of good configuration is propagated because the type of region variables remain constant through the typing derivation.

To sum up, for proving the propagation of the goodness property we shall concentrate on the closedness property (and the disjointness property, when calling a function) since the remaining ones either are trivial, or have been proved before. Let us distinguish cases on the expression being evaluated:

- **Case  $e \equiv c$**

Trivial, since a literal is closed by convention and  $\text{dom } h \setminus \text{dom } h' = \emptyset$ .

- **Case  $e \equiv x$**

We get  $L = \{x\}$  and, by hypothesis,  $\text{closed}(h, E(x))$ , which is equivalent to  $\text{closed}(h', v)$ . In this case we also obtain  $\text{dom } h \setminus \text{dom } h' = \emptyset$ .

- **Case  $e \equiv x @ r$**

From the definition of *copy* function it follows that whenever  $(h', p) = \text{copy}(h, E(x), j)$ ,  $p$  is closed in  $h'$  if and only if  $E(x)$  is closed in  $h$ . But  $\text{closed}(h, E(x))$  holds by hypothesis, so  $\text{closed}(h', v)$  also does. We obtain, in this case,  $\text{dom } h \setminus \text{dom } h' = \emptyset$ .

- **Case  $e \equiv C \bar{a}_i^n @ r$**

Each  $a_i$  is closed in  $h$  by hypothesis. Since the result  $v$  is built from closed components, it is also closed. Hence  $\text{closed}(h', v)$  holds. Again,  $\text{dom } h \setminus \text{dom } h'$  is empty.

- **Case  $e \equiv g \bar{a}_i^n @ \bar{r}_j^m$**

Let us assume that  $g \bar{y}_i^n @ \bar{r}'_j^m = e_g \in \Sigma$ . By the  $[App^*]$  rule we get:

$$\underbrace{\overline{[y_i \mapsto E(a_i)^n, r'_j \mapsto E(r_j)^m, self \mapsto k+1]} \vdash h, k+1, e_g \Downarrow_f^* h', k+1, v}_{E_g}$$

Moreover, if  $\Gamma(g) = \forall \bar{\alpha} \bar{\rho}. \bar{t}_{i,g}^n \rightarrow \bar{\rho}_{j,g}^m \rightarrow s_g$ , the following derivation follows from the fact that  $\Gamma$  is correct w.r.t.  $\Sigma$ :

$$\underbrace{\Gamma_\omega + \overline{[y_i : \bar{t}_{i,g}^n]} + \overline{[r'_j : \bar{\rho}_{j,g}^m]} + [self : \rho_{self}]}_{\Gamma_g} \vdash e_g : s_g \quad \text{for some } \Gamma_\omega \subseteq \Gamma \quad (3.9)$$

By the  $[APP]$  rule we know that the type of the application is a concrete instance of  $\Gamma(g)$  and there exists a type substitution  $\theta$  such that:

$$\overline{\theta(t_{i,g}^n)} \rightarrow \overline{\theta(\rho_{j,g}^m)} \rightarrow \theta(s_g) = \overline{\Gamma(a_i)^n} \rightarrow \overline{\Gamma(\rho_j)^m} \rightarrow s$$

Since  $\rho_{self}$  does not occur in  $\Gamma(g)$ , we can safely assume that  $\theta(\rho_{self}) = \rho_{self}$ . By applying the substitution lemma (Lemma 3.9) to (3.9) we obtain:

$$\underbrace{\theta(\Gamma_\omega) + \overline{[y_i : \Gamma(a_i)]^n} + \overline{[r'_j : \Gamma(\rho_j)]^m} + [self : \rho_{self}]}_{\theta(\Gamma_g)} \vdash e_g : s \quad (3.10)$$

From the induction hypothesis we obtain the following implication:

$$\left. \begin{array}{l} E_g \vdash h, k+1, e_g \Downarrow^* h', k+1, v \\ \theta(\Gamma_g) \vdash e_g : s \\ (\theta(\Gamma_g), E_g, h, k+1, fv(e_g), \theta(\Gamma_g)) \text{ good} \end{array} \right\} \Rightarrow \left\{ \begin{array}{l} E_g \vdash h, k+1, e_g \Downarrow h'', k+1, v \text{ for some } h'' \subseteq h' \\ \forall p \in \text{dom } h \setminus \text{dom } h''. \forall z \in \text{dom } E_g. E_g(z) \rightarrow_h^* p \\ \quad \Rightarrow \theta(\Gamma_g)(z) \in \mathbf{UnsafeType} \\ \forall p \in \text{dom } h \setminus \text{dom } h''. \forall z \in \text{dom } E_g. E_g(z) \rightarrow_h^* p \\ \quad \wedge E_g(z) \not\rightarrow_h^* p \\ \quad \Rightarrow \theta(\Gamma_g)(z) \in \mathbf{DgrType} \\ (v, h'') \text{ good} \end{array} \right. \quad (3.11)$$

Let us prove the antecedent of this implication. The first fact follows from the assumption. The second one follows from (3.10). This derivation takes part in a different context definition  $g \overline{y_i} @ \overline{r'_j} = e_g$ . Now we prove that the configuration  $(\theta(\Gamma_g), E_g, h, k+1, fv(e_g), \theta(\Gamma_g))$  is good. The fact  $closed(h, E_g, fv(e_g))$  follows from  $closed(h, E, \{\overline{a_i}^n\})$ , which holds by assumption. The fact  $P_{disj}(E_g, h, \theta(\Gamma_g))$  can be proven by contradiction. It is trivial that the first property of  $P_{disj}(E_g, h, \theta(\Gamma_g))$  holds, so let us firstly assume that the second one does not. That is, there exist distinct parameters  $y_i, y_j \in \text{dom } \theta(\Gamma_g)$  such that  $\theta(\Gamma_g)(y_i) \in \mathbf{CdmType}$  and a pointer  $p$  such that  $E_g(y_i) \rightarrow_h^* p$  and  $E_g(y_j) \rightarrow_h^* p$ . We unfold the definitions of  $\theta(\Gamma_g)$  and  $E_g$  in order to obtain:

$$\Gamma(a_i) \in \mathbf{CdmType} \quad E(a_i) \rightarrow_h^* p \quad E(a_j) \rightarrow_h^* p$$

If  $a_i = a_j$  then  $a_j$  has also a condemned type in  $\Gamma$ , and the result of the operator  $\oplus$  is not defined, contradicting the fact that the expression is well-typed. If  $a_i \neq a_j$  then, by Assumption 3.38, we have three possibilities: (SHR), (SHP), or (CONF). If we are able to prove neither (SHP) nor (CONF) are possible, we will get  $a_j \in \text{sharerec}(a_i, e)$ , which implies  $a_j \in \text{dom } \Gamma_R$  and the disjoint sum of  $\Gamma_R$  with  $\bigoplus_{i=1}^n [a_i : t_i]$  in the [APP] rule would be undefined. So let us prove that assuming (SHP) or (CONF) leads to a contradiction: in any of these cases, there exists a parameter  $y_0$  of the caller function  $f$  such that  $E(y_0) \rightarrow_h^* p$ . By Assumption 3.39, there exists another parameter  $y_2$  such that  $E(y_2) \rightarrow_h^* p$  and  $y_2 \in \text{sharerec}(a_i, e)$ .

- If  $y_2 = a_i$  then  $\Gamma(y_2) = \Gamma(a_i) \in \mathbf{CdmType}$ , which implies, by Lemma 3.44, that  $\Gamma_0(y_2)$  can be condemned or in-danger. If the latter is in-danger we obtain a contradiction, since  $\Gamma_0$  does not have in-danger parameters. If it is condemned, we distinguish cases depending on whether (SHP) or (CONF) holds:

(SHP) In this case there exists a parameter  $y_1$ , different from  $y_0$  (but not necessarily different from  $y_2$ ) such that  $E(y_1) \rightarrow_h^* p$ . This contradicts property (2) of  $P_{disj}(E, h, \Gamma_0)$ , as  $E(y_2) = E(a_i) \rightarrow_h^* p$ , and either  $y_1$  or  $y_0$  is a parameter different from  $y_2$  pointing to its recursive closure.

(CONF) If  $y_2 = y_0$  then  $\Gamma(y_0) = \Gamma(y_2) \in \mathbf{CdmType}$ , which contradicts property (3) of  $P_{disj}(E, h, \Gamma_0)$ . If  $y_2 \neq y_0$ , then we get two different parameters such that  $E(y_0) \rightarrow_h^* p$ ,

$E(y_2) \rightarrow_h^* p$ , and  $\Gamma(y_2)$  is condemned. This contradicts  $P_{disj}(E, h, \Gamma_0)$  (property (2)).

- If  $y_2 \neq a_i$  then  $\Gamma(y_2) \in \mathbf{DgrType}$ , as  $y_2$  belongs to the  $\Gamma_R$  environment of the [APP] rule. This implies that  $\Gamma_0(y_2) \in \mathbf{DgrType}$  as well, which, again, leads to a contradiction.

Now we assume that property (3) of  $P_{disj}(E_g, h, \theta(\Gamma_g))$  does not hold. That is, there exists some parameter  $y_i \in \text{dom } \theta(\Gamma_g)$  with a condemned type and a pointer  $p$  in its recursive closure such that  $E_g(y_i) \Rightarrow_h^* p$ . Equivalently, there exists some argument  $a_i \in \text{dom } \Gamma$  with a condemned type and a pointer  $p$  in its recursive closure such that  $E(a_i) \Rightarrow_h^* p$ . By Assumption 3.37, we get:

$$\begin{aligned} (TREE) \quad & \neg isTree(a_i), \text{ or} \\ (CONF) \quad & \exists y_0. E(y_0) \Rightarrow_h^* p, \text{ where } y_0 \text{ is a parameter of } f \end{aligned}$$

(TREE) clearly contradicts the [APP] rule, so let us assume (CONF). Again, by Assumption 3.39, there exists a parameter  $y_2$  of  $f$  such that  $E(y_2) \rightarrow_h^* p$  and  $y_2 \in sharerec(a_i, e)$ . We distinguish cases as above:

- If  $y_2 = a_i$ , then  $\Gamma(y_2) = \Gamma(a_i) \in \mathbf{CdmType}$ , which implies that  $\Gamma_0(y_2)$  is condemned or in-danger. The latter case leads to a contradiction with property 1 of  $P_{disj}(E, h, \Gamma_0)$ . If  $\Gamma_0(y_2)$  is condemned and  $y_2 \neq y_0$  we get a contradiction with property 2 of  $P_{disj}(E, h, \Gamma_0)$  as  $E(y_2) = E(a_i) \rightarrow_h^* p$  and  $E(y_0) \rightarrow_h^* p$ , where  $y_2$  has a condemned type. If  $\Gamma_0(y_2)$  is condemned and  $y_2 = y_0$ , then  $E(y_2) = E(y_0) \Rightarrow_h^* p$ , which contradicts (3).
- If  $y_2 \neq a_i$ , then  $\Gamma(y_2)$  is in-danger (since  $y_2$  occurs within  $\Gamma_R$  of the [APP] rule), and so is  $\Gamma_0(y_2)$ . This contradicts (1) of  $P_{disj}(E, h, \Gamma_0)$ .

Since we have proven the left-hand side of the implication in (3.11) we get the facts of the right-hand side. The first conclusion of the theorem follow trivially the first fact:

$$E \vdash h, k, g \bar{a}_i^n @ \bar{r}_j^m \Downarrow h''|_k, k, v$$

For proving the second and third conclusions, let us assume a variable  $z \in \text{dom } E$  such that  $E(z) \rightarrow_h^* p$  and  $p \in \text{dom } h \setminus \text{dom } h''|_k$ . Since none of the cells in  $h$  belongs to the  $k + 1$ -th region,  $p \in \text{dom } h \setminus \text{dom } h''$ . By Lemma 3.30  $p$  belongs to  $live(h, E, \{\bar{a}_i\})$ , so there exists an argument  $a_i$  such that  $E(a_i) = E_g(y_i) \rightarrow_h^* p$ . Since  $y_i$  has an unsafe type in  $\theta(\Gamma_g)$  (by (3.11)),  $a_i$  has an unsafe type in  $\Gamma$ . But, since  $a_i$  is an argument of the function, that type cannot be in-danger, so it is condemned. Let us distinguish cases:

- If  $p$  is a recursive descendant of  $a_i$  we have the same situation as before:  $z$  points to a recursive child of a condemned argument. This implies (SHR), (SHP) or (CONF). Again, the only possibility that does not lead to a contradiction is the first one, so we can safely assume it to hold. Since  $z \in sharerec(a_i, e)$  holds,  $z$  belongs to the  $\Gamma_R$  environment in the [APP] rule, so  $\Gamma(z) \in \mathbf{DgrType}$ , and we are done.
- The case in which  $p$  is not a recursive descendant of  $a_i$  leads to a contradiction: it would imply that  $E_g(y_i) \rightarrow_h^* p$ , but  $E_g(y_i) \not\rightarrow_h^* p$ , from which it follows that  $\theta(\Gamma_g)(y_i) \in \mathbf{DgrType}$  by (3.11), but  $\theta(\Gamma_g)$  cannot have parameters with an in-danger type.

For proving the fourth conclusion we have to prove that  $closed(h'', v)$  implies  $closed(h''|_k, v)$ . That is,  $closure(h'', v)$  does not contain cells in the  $k + 1$ -th region of the heap. By Theorem 3.26,

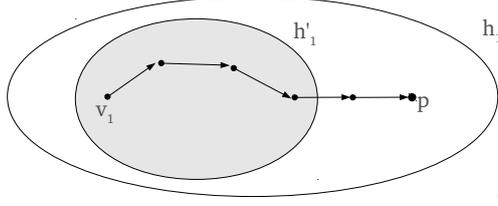


Figure 3.17: If  $p$  is reachable from  $v_1$  and  $p \notin \text{dom } h'_1$ , then  $v_1$  is not closed in  $h'_1$ .

$\text{build}(h'', v, \theta(s))$  is well-defined and consistent with  $\text{build}^*(h, E_g, \theta(\Gamma_g))$ . If there existed a cell in  $\text{closure}(h'', v)$  in the topmost region of the heap, we would obtain the binding  $[\rho_{\text{self}} \mapsto k + 1] \in \text{build}(h'', v, \theta(s))$ , since  $\text{self}$  is the only region variable such that  $E_g(\text{self}) = k + 1$  and  $\theta(\Gamma_g)(\text{self}) = \rho_{\text{self}}$ . But this contradicts the fact that  $\theta(s)$  does not contain the  $\rho_{\text{self}}$  RTV, a constraint imposed by the [FUN] rule when applied to the definition of  $g$ . Hence,  $\text{closed}(h''|_k, v)$  holds.

- **Case  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$**

Assume the following execution for  $e_1$

$$E \vdash h, k, e_1 \Downarrow_f^* h_1, k, v_1$$

and that  $\Gamma' = \Gamma_1 \sqcup \Gamma_2$  where  $\Gamma_1$  and  $\Gamma_2 + [x_1 : \tau_1]$  are typing environments of the corresponding expressions  $e_1$  and  $e_2$ . The facts  $\text{closed}(h, E, \text{fv}(e_1))$  and  $P_{\text{disj}}(E, h, \Gamma_0)$  follow from the assumptions of the theorem. So, the initial configuration of  $e_1$  is good, and we can apply the induction hypothesis in order to get:

$$E \vdash h, k, e_1 \Downarrow h'_1, k, v_1 \quad \text{for some } h'_1 \subseteq h_1 \quad (3.12)$$

$$\text{closed}(h'_1, v_1)$$

$$\forall p \in \text{dom } h \setminus \text{dom } h'_1. \forall z \in \text{dom } E.E(z) \rightarrow_h^* p \Rightarrow \Gamma_1(z) \in \mathbf{UnsafeType} \quad (3.13)$$

$$\forall p \in \text{dom } h \setminus \text{dom } h'_1. \forall z \in \text{dom } E.E(z) \rightarrow_h^* p \wedge E(z) \not\rightarrow_h^* p \Rightarrow \Gamma_1(z) \in \mathbf{DgrType} \quad (3.14)$$

With regard to the main expression  $e_2$ , we assume the following  $\Downarrow^*$ -execution:

$$\underbrace{E \uplus [x_1 \mapsto v_1]}_{E_1} \vdash h_1, k, e_2 \Downarrow_f^* h', k, v \quad (3.15)$$

Firstly we should prove that we can safely substitute  $h'_1$  for  $h_1$  in this judgement. However, this task becomes easier if we prove  $\text{closed}(h'_1, E_1, \text{fv}(e_2))$  first. Let  $z \in \text{fv}(e_2)$ . We already know that  $\text{closed}(h'_1, E_1(x_1))$ , so let us assume that  $z \neq x_1$ . If  $E_1(z)$  is not closed in  $h'_1$ , there must exist a pointer  $p$  such that  $E_1(z) \rightarrow_{h'_1}^* p$ , but  $p \notin \text{dom } h'_1$ . Notice that we get  $E_1(z) \rightarrow_{h_1}^* p$  as well, since  $h'_1 \subseteq h_1$  and, by Lemma 3.29,  $E_1(z) \rightarrow_h^* p$ . Let us distinguish cases:

- $p \notin \text{dom } h$

Since  $z \in \text{fv}(e_2)$  and is distinct from  $x_1$ , we obtain  $z \in \text{fv}(e)$ . This case leads to a contradiction because of the assumption  $\text{closed}(h, E, \text{fv}(e))$ .

- $p \in \text{dom } h$

By (3.13) the  $z$  variable gets an unsafe type in  $\Gamma_1$ , so it cannot occur free in  $e_2$  by the side condition of [LET]. This case also leads to a contradiction.

Thus  $\text{closed}(h'_1, E_1, \text{fv}(e_2))$  holds. Now we prove that  $p \notin \text{live}(h_1, E_1, \text{fv}(e_2))$  for every pointer  $p \in \text{dom } h_1 \setminus \text{dom } h'_1$ . Again, we prove this by contradiction: if  $p \notin \text{dom } h'_1$  but  $p \in \text{live}(h_1, E_1, \text{fv}(e_2))$ , then  $\text{closed}(h'_1, E_1, \text{fv}(e_2))$  would not hold, but we just have proved that it does hold, so  $p \notin \text{live}(h_1, E_1, \text{fv}(e_2))$ . Thus we can substitute  $h'_1$  for  $h_1$  in (3.15) (by Lemma 3.31), and we get:

$$E_1 \vdash h'_1, k, e_2 \Downarrow_f^* h'_2, k, v \quad \text{for some } h'_2 \subseteq h' \quad (3.16)$$

It is easy to prove that the configuration  $(\Gamma_2, E_1, h'_1, k, \text{fv}(e_2), \Gamma_0)$  is good, as we already know that  $\text{closed}(h'_1, E_1, \text{fv}(e_2))$  holds. Thus we can apply the induction hypothesis to (3.16) in order to obtain,

$$E_1 \vdash h'_1, k, e_2 \Downarrow h'', k, v \quad \text{for some } h'' \subseteq h'_2 \subseteq h' \quad (3.17)$$

$$\text{closed}(h'', v)$$

It is straightforward to see that  $(\Gamma_2, E_1, h_1, k, \text{fv}(e_2), \Gamma_0)$  is good, so the following can be obtained by instantiating the induction hypothesis and by taking  $h_1$  as the initial heap, instead of  $h'_1$ :

$$\forall p \in \text{dom } h_1 \setminus \text{dom } h''. \forall z \in \text{dom } E_1. E_1(z) \rightarrow_{h_1}^* p \Rightarrow \Gamma_2(z) \in \mathbf{UnsafeType} \quad (3.18)$$

$$\forall p \in \text{dom } h_1 \setminus \text{dom } h''. \forall z \in \text{dom } E_1. E_1(z) \rightarrow_{h_1}^* p \wedge E_1(z) \not\rightarrow_{h_1}^* p \Rightarrow \Gamma_2(z) \in \mathbf{DgrType} \quad (3.19)$$

Thus we can derive, from (3.12) and (3.17),  $E \vdash h, k, e \Downarrow h'', k, v$ . The final step is to prove that for every pointer  $p \in \text{dom } h \setminus \text{dom } h''$  and every variable  $z \in \text{dom } E$  such that  $E(z) \rightarrow_h^* p$  has an unsafe type in  $\Gamma_1 \sqcup \Gamma_2$ . Moreover, if  $E(z) \not\rightarrow_h^* p$ , then  $z$  has an in-danger type in that environment.

–  $p \notin \text{dom } h'_1$

Then  $p$  has been removed from the heap during the evaluation of  $e_1$ . By (3.13) we get  $\Gamma_1(z) \in \mathbf{UnsafeType}$  and hence  $(\Gamma_1 \sqcup \Gamma_2)(z) \in \mathbf{UnsafeType}$ . In the case that  $E(z) \not\rightarrow_h^* p$  we get, more specifically,  $\Gamma_1(z) \in \mathbf{DgrType}$  by (3.14) and, hence,  $(\Gamma_1 \sqcup \Gamma_2)(z) \in \mathbf{DgrType}$ .

–  $p \in \text{dom } h'_1$

In this case  $p$  has been removed from the heap during the evaluation of  $e_2$ . Since  $h'_1 \subseteq h_1$ , we know that  $p \in \text{dom } h_1$ . Besides this, the  $\Downarrow^*$ -derivation does not remove pointers from the heap, so  $h \subseteq h_1$ . This means that  $E(z) \rightarrow_h^* p$  implies  $E(z) \rightarrow_{h_1}^* p$ . We can apply (3.18) and obtain  $\Gamma_2(z) \in \mathbf{UnsafeType}$ , which implies  $(\Gamma_1 \sqcup \Gamma_2)(z) \in \mathbf{UnsafeType}$ . Similarly,  $E(z) \not\rightarrow_h^* p$  implies  $E(z) \not\rightarrow_{h_1}^* p$ , so by applying (3.19) we get  $\Gamma_2(z) \in \mathbf{DgrType}$ , from which it follows that  $(\Gamma_1 \sqcup \Gamma_2)(z) \in \mathbf{DgrType}$ .

- **Case  $e \equiv \text{case } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i^n}$**

We assume the execution of the  $r$ -th branch:

$$\underbrace{E \uplus [\overline{x_{rj} \mapsto \overline{v_j^{n_r}}}]}_{E_r} \vdash h, k, e_r \Downarrow_f^* h', k, v$$

From the assumption  $\text{closed}(h, E, \text{fv}(e))$  it follows that  $\text{closed}(h, E_r, \text{fv}(e_r))$ . If we denote by  $\Gamma_r$  the typing environment of  $e_r$ , the configuration  $(\Gamma_r, E_r, h, k, \text{fv}(e_r), \Gamma_0)$  is good. By induction hypothesis we get:

$$E \uplus [\overline{x_{rj} \mapsto \overline{v_j^{n_r}}}] \vdash h, k, e_r \Downarrow h'', k, v \quad \text{for some } h'' \subseteq h'$$

$$\text{closed}(h'', v)$$

$$\forall p \in \text{dom } h \setminus \text{dom } h'', \forall z \in \text{dom } E_r. E_r(z) \rightarrow_h^* p \Rightarrow \Gamma_r(z) \in \mathbf{UnsafeType} \quad (3.20)$$

$$\forall p \in \text{dom } h \setminus \text{dom } h'', \forall z \in \text{dom } E_r. E_r(z) \rightarrow_h^* p \wedge E_r(z) \not\rightarrow_h^* p \Rightarrow \Gamma_r(z) \in \mathbf{DgrType} \quad (3.21)$$

From the first fact we can derive  $E \vdash h, k, e \Downarrow h'', k, v$ . Now we prove that for every variable  $z \in \text{dom } E$  pointing to a location  $p \in \text{dom } h \setminus \text{dom } h''$  has an unsafe type. Since  $E(z) = E_r(z) \rightarrow_h^* p$ , by (3.20) we get  $\Gamma_r(z) \in \mathbf{UnsafeType}$ . Since  $z$  cannot be any of the pattern variables, we get  $z \in \text{dom } \Gamma$  and the required result follows trivially. In a similar way we can apply (3.21) in order to obtain  $\Gamma(z) \in \mathbf{DgrType}$  in the case where  $E(z) \not\rightarrow_h^* p$ .

- **Case**  $e \equiv \text{case! } x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i}} \rightarrow e_i^n$

Assume the  $r$ -th branch is executed:

$$\underbrace{E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}]}_{E_r} \vdash \underbrace{h \uplus [p \mapsto w]}_{h^+}, k, e_r \Downarrow_f^* h', k, v$$

By the [CASE!] rule,  $x$  has an condemned type in  $\Gamma'$ , and hence an unsafe type in  $\Gamma$ .

Let us prove that  $E(x) = p \notin \text{live}(h^+, E_r, \text{fv}(e_r))$  by contradiction: if  $p \in \text{live}(h^+, E_r, \text{fv}(e_r))$ , there exists a variable  $z \in \text{fv}(e_r)$  such that  $E_r(z) \rightarrow_{h^+}^* p$ . The  $z$  variable cannot be one of the pattern variables  $\bar{x}_{rj}^{nr}$ , since otherwise we would have  $E(x) = p \rightarrow_{h^+} E_r(z) \rightarrow_{h^+}^* p$ , contradicting the fact that the initial heap is acyclic. Therefore,  $z$  is not a pattern variable, so  $E_r(z) = E(z)$ . Obviously,  $E(x) \rightarrow_{h^+}^* p$ . By Assumption 3.38 we have three possibilities: (SHR), (SHP), or (CONF). We can prove that neither (SHP) nor (CONF) are possible by following a similar reasoning as in the end of the case  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$  (we only have to substitute  $x$  for  $a_i$ ). Thus, the only possibility is (SHR), that is,  $z \in \text{sharerec}(x, e)$ . Then  $z$  belongs to the  $\Gamma_R$  environment of the [CASE!] rule, and  $z \notin \text{fv}(e_r)$ , leading to a contradiction.

Since  $p \notin \text{live}(h^+, E_r, \text{fv}(e_r))$ , by Lemma 3.31 we can safely discard the binding in the evaluation of  $e_r$  in order to get a  $\Downarrow^*$ -derivation of the same size as the previous one:

$$E_r \vdash h, k, e_r \Downarrow_f^* h' \setminus p, k, v$$

Moreover, since the initial configuration is good and  $p \notin \text{live}(h^+, E_r, \text{fv}(e_r))$ ,  $\text{closed}(h, E_r, \text{fv}(e_r))$  holds and hence the configuration  $(\Gamma_r, E_r, h, k, \text{fv}(e_r), \Gamma_0)$  is good, being  $\Gamma_r$  the typing environment of the  $e_r$ . We can apply the induction hypothesis in order to get:

$$E_r \vdash h, k, e_r \Downarrow h'', k, v \quad \text{for some } h'' \subseteq h' \setminus p \subseteq h'$$

$$\text{closed}(h'', v)$$

$$\forall q \in \text{dom } h \setminus \text{dom } h''. \forall z \in \text{dom } E_r. E_r(z) \rightarrow_h^* q \Rightarrow \Gamma_r(z) \in \mathbf{UnsafeType} \quad (3.22)$$

$$\forall q \in \text{dom } h \setminus \text{dom } h''. \forall z \in \text{dom } E_r. E_r(z) \rightarrow_h^* 1 \wedge E_r(z) \not\rightarrow_h^* q \Rightarrow \Gamma_r(z) \in \mathbf{DgrType} \quad (3.23)$$

Thus  $(v, h'')$  is good. Now we prove that for every pointer  $q \in \text{dom } h^+ \setminus \text{dom } h''$  and every variable  $z \in \text{dom } E$  such that  $E(z) \rightarrow_{h^+}^* q$ , then  $z$  occurs in  $\Gamma$  with an unsafe type. If  $q \neq p$  (i.e.,  $q$  is not the pointer being destroyed by this case!) we further distinguish cases:

- If  $E(z) \rightarrow_{h^+}^* p$ , we have already proved that  $z \in \text{sharerec}(x, e)$ , so  $z$  must occur in  $\Gamma_R$  with an in-danger type.
- If  $E(z) \not\rightarrow_{h^+}^* p$ , then  $p$  cannot be in the middle of the path between  $E(z)$  and  $q$  in  $h^+$ , so  $E(z) \rightarrow_h^* q$ , and we proceed as in the non-destructive case.

If  $q = p$ , let  $z$  such that  $E(z) \rightarrow_{h^+}^* p$ . If  $z$  is the discriminant of the **case!**, then it has a condemned type. If  $z \neq x$  then we have already shown that  $z \in \text{sharerec}(x, e)$ . Hence  $z$  gets an in-danger type, as it belongs to the  $\Gamma_R$  environment. Finally we prove, under the same conditions, that  $\Gamma(z) \in \mathbf{DgrType}$  whenever  $E(z) \rightarrow_{h^+}^* q$ , but  $E(z) \not\rightarrow_{h^+}^* q$ . We distinguish cases:

$q = p$  The  $z$  variable cannot be the **case!** discriminant, since  $E(x) \rightarrow_{h^+} p$  and we are assuming that  $E(z) \not\rightarrow_{h^+} p$ . So, we safely assume that  $z \neq x$ . We have already shown that  $z \in \text{sharerec}(x, e)$ , which forces  $z$  to appear with an in-danger type in the  $\Gamma_R$  environment of [CASE!]. Hence  $\Gamma(z) \in \mathbf{DgrType}$ .

$q \neq p$

- If  $z$  is not the **case!** discriminant  $x$ , then  $\Gamma_r(z) \leq \Gamma(z)$ , since  $z$  cannot be a pattern variable. Again, we have to distinguish cases depending on whether  $E(z)$  points to  $p$  in  $h^+$ . If it does, then  $z \in \text{sharerec}(x, e)$ , as we have proved before, and belongs to  $\Gamma_R$ , so it has an in-danger type. If it does not, we get  $E(z) \rightarrow_h^* q$ , and we just apply (3.23) to get  $\Gamma_r(z) \in \mathbf{DgrType}$ , which implies  $\Gamma(z) \in \mathbf{DgrType}$ .
- If  $z$  is the **case!** discriminant  $x$ , we cannot ensure that  $\Gamma_r(x) \leq \Gamma(x)$ , since the [CASE!] rule replaces the binding of the  $x$  variable in  $\Gamma_r$  with the binding  $[x : T !@ \rho]$ . However, we shall see that this case leads to a contradiction. Since  $q \neq p$ , we know that  $E(x)$  reaches  $q$  through some of its pattern variables, that is,  $E(x) \rightarrow_{h^+} E_r(x_{ri}) \rightarrow_{h^+}^* q$  for some  $i \in \{1..n_r\}$ . Since  $q \neq p$ , this is equivalent to  $E(x) \rightarrow_h E_r(x_{ri}) \rightarrow_h^* q$ . We distinguish cases:
  - \* If  $i \in \text{RecPos}(C)$  we get  $E(x) \rightarrow_{h^+} E_r(x_{ri}) \rightarrow_h^* q$ . Since we are assuming that  $E(z) \not\rightarrow_{h^+}^* q$  we know that  $E_r(x_{ri}) \not\rightarrow_{h^+}^* q$ . By (3.23),  $\Gamma_r(x_{ri}) \in \mathbf{DgrType}$ , which contradicts the *inh* predicate in [CASE!].
  - \* If  $i \notin \text{RecPos}(C)$  we get  $E_r(x_{ri}) \rightarrow_h^* q$ , which implies, by (3.22),  $\Gamma_r(x_{ri}) \in \mathbf{UnsafeType}$  which, again, contradicts the *inh* assumptions in [CASE!].

□

This theorem can be instantiated to show that well-typed programs do not access dangling pointers.

**Corollary 3.46.** *Assume a Core-Safe program  $\text{prog}$ . Let  $\Sigma$  be an environment containing its function definitions and  $e$  the main expression of the program. If the following judgement is derivable:*

$$[\text{self} \mapsto 0] \vdash [], 0, e \Downarrow^* h', k, v$$

*and  $\Gamma + [\text{self} : \rho_{\text{self}}] \vdash e : s$  holds for some type  $s$  and some type environment  $\Gamma$  which only contains functional type schemes and is correct w.r.t.  $\Sigma$ , then we obtain:*

$$[\text{self} \mapsto 0] \vdash [], 0, e \Downarrow h'', k, v$$

*for some  $h'' \subseteq h$ .*

*Proof.* If we denote by  $\Gamma_0$  the environment  $\Gamma + [self : \rho_{self}]$ , we have to show that the initial configuration

$$(\Gamma_0, [self \mapsto 0], [], 0, fv(e), \Gamma_0)$$

is good. Obviously, the empty heap is closed and acyclic. The disjointness property holds vacuously, since there are not function parameters in the main expression. Moreover,  $build^*([], [self \mapsto 0], \Gamma_0) = [\rho_{self} \mapsto 0]$ , which is well-defined. Finally, the only region variable in the runtime environment pointing to the topmost region (0) is *self*, and this is the only region variable of type  $\rho_{self}$  in  $\Gamma_0$ , since  $\Gamma$  only contain functional type schemes. Therefore, the initial configuration is good, and we can apply Theorem 3.45 to get the desired result.  $\square$

### 3.7 Towards the type inference of function definitions

The type system shown in this chapter provides us a way to establish the correctness of a *Core-Safe* program, subject to the existence of a typing derivation for it. However, it does not give any clue on how to build these typing derivations. Moreover, it assumes that the program to be typed is annotated with region variables, even when these variables are not specified by the programmer.

The aim of the next two chapters is to devise a mechanical way to annotate a *Core-Safe* program with region variables, and to construct its corresponding typing derivation. This is achieved by two algorithms:

- A region inference algorithm which, given a region-free *Core-Safe* program, annotates its abstract syntax tree with region variables. This algorithm is only concerned with Hindley-Milner types and region type variables, regardless of whether these types are safe, in-danger or condemned.
- A destruction safety inference algorithm, relying on an ancillary sharing analysis and concerning exclusively the destructive pattern matching feature of the language, without regard to regions. It determines whether the type of a given variable should be safe, in-danger or condemned.

None of these algorithms can, on its own, guarantee the existence of a typing derivation. We can only build a correct typing derivation from the results of *both* algorithms. In order to prove the correctness of these algorithms we would need to consider them jointly, as a single algorithm. But, in this case, the correctness proof and the algorithm itself would become unwieldy.

Instead of putting the two algorithms together, we split the type system in two simpler type systems in such a way that, if a program is typable according to these two simpler systems, then so it is according to the initial one. As a consequence, we obtain:

- A type system dealing exclusively with regions, and leaving out the in-danger or condemned nature of types. It is defined as a set of judgements  $\Gamma_R \vdash_{Reg} e : s$ .
- A type system dealing exclusively with safe, in-danger or condemned marks, without regard to regions. It is defined as a set of judgements  $\Gamma_D \vdash_{Dst} e : s$ .

The environments occurring in the  $\vdash_{Reg}$  derivations are called *region environments*. They map variables and function and constructor symbols to safe types and safe type signatures respectively. On the other hand, the environments in the  $\vdash_{Dst}$  derivations are called *mark environments*. A mark environment maps type variables to marks in the set  $\mathbf{Mark} = \{s, r, d\}$  and function and constructor names to mark signatures. A mark signature is an expression of the form  $\overline{m}_i^n \rightarrow s$ , in which each  $m_i$  stand for the

$$\begin{array}{c}
\frac{}{\Gamma \vdash_{Reg} c : B} [LIT_{Reg}] \\
\\
\frac{}{\Gamma + [x : s] \vdash_{Reg} x : s} [VAR_{Reg}] \\
\\
\frac{}{\Gamma + [x : T@r', r : \rho] \vdash_{Reg} x@r : T@r} [COPY_{Reg}] \\
\\
\frac{\Gamma \vdash_{Reg} e_1 : s_1 \quad \Gamma + [x_1 : s_1] \vdash_{Reg} e : s}{\Gamma \vdash_{Reg} \mathbf{let} x_1 = e_1 \mathbf{in} e_2 : s} [LET_{Reg}] \\
\\
\frac{\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s \trianglelefteq \sigma \quad \Gamma \vdash_{Reg} a_i : s_i}{\Gamma + [f : \sigma] + [\bar{r}_j : \bar{\rho}_j^m] \vdash_{Reg} f \bar{a}_i^n @ \bar{r}_j^m : s} [APP_{Reg}] \\
\\
\frac{\Gamma(C) = \sigma \quad \bar{s}_i^n \rightarrow \rho \rightarrow s \trianglelefteq \sigma \quad \forall i \in \{1..n\}. \Gamma \vdash_{Reg} a_i : s_i}{\Gamma + [r : \rho] \vdash_{Reg} C \bar{a}_i^n @ r : s} [CONS_{Reg}] \\
\\
\frac{\Gamma(x) = s_x \quad \forall i \in \{1..n\}. \bar{s}_{ij}^{n_i} \rightarrow \rho \rightarrow s_x \trianglelefteq \Gamma(C_i) \quad \forall i \in \{1 \dots n\}. \Gamma + [x_{ij} : s_{ij}^{n_i}] \vdash_{Reg} e_i : s}{\Gamma \vdash_{Reg} \mathbf{case}(!) x \mathbf{of} \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n : s} [CASE(!)_{Reg}]
\end{array}$$

Figure 3.18: Typing rules for deriving  $\vdash_{Reg}$  judgements.

mark of the corresponding function parameter, whereas the  $s$  mark at the end stands for the result of the function (which is always safe).

Formally, we have two kinds of judgements:

$$\begin{array}{l}
\Gamma_R \vdash_{Reg} e : s \quad \text{where } \Gamma_R : (\mathbf{Var} \rightarrow \mathbf{SafeType}) \times (\mathbf{RegVar} \rightarrow \mathbf{RegType}) \times (\mathbf{Fun} \cup \mathbf{Cons} \rightarrow \mathbf{SafeFunType}) \\
\Gamma_D \vdash_{Dst} e : s \quad \text{where } \Gamma_D : (\mathbf{Var} \rightarrow \mathbf{Mark}) \times (\mathbf{Fun} \rightarrow \mathbf{MarkSig})
\end{array}$$

Figures 3.18 and 3.19 contain the typing rules for the  $\vdash_{Reg}$  and  $\vdash_{Dst}$  type systems respectively. All these rules are simplified versions of those occurring in Figure 3.6 and hence they will not be explained here again. The  $inh_{i,C}$  predicate appearing in  $[CASE!_{Dst}]$  rule is defined in Figure 3.20. Again, this predicate is a simplified version of its counterpart in Figure 3.8. The operators defined in Section 3.3 also apply to region and type environments:  $+$  forbids disjoint domains, whereas  $\oplus$  allows them if they bound common variables to a safe mark. The result of  $\Gamma_{D,1} \sqcup \Gamma_{D,2}$  maps common variables in  $\Gamma_{D,1}$  and  $\Gamma_{D,2}$  to the maximal mark, assuming the following order between marks:  $s \leq d \leq r$ . This operator demands the functions belonging to the domain of  $\Gamma_{D,1}$  and  $\Gamma_{D,2}$  to be bound to the same mark signature in both of them.

For proving that the union of the  $\vdash_{Reg}$  and  $\vdash_{Dst}$  type systems are equivalent to the original  $\vdash$  type system, we have to specify how to combine the region environments with the mark environments. This is done by means of an operator  $\circ$  defined in Figure 3.21. By abuse of notation, we extend the domain of this operator to safe type schemes and mark signatures as follows:

$$(\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s) \circ (\bar{m}_i^n \rightarrow s) \stackrel{\text{def}}{=} (\bar{s}_i \circ \bar{m}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s)$$

If the number of  $s_i$  is different from the number of  $m_i$ , the result is undefined.

In the following,  $\Gamma_R$  always denotes a region environment and  $\Gamma_D$  denotes a mark environment.

$$\begin{array}{c}
\frac{\Gamma' \vdash_{Dst} e : s \quad \Gamma \supseteq \Gamma'}{\Gamma \vdash_{Dst} e : s} [EXT_{Dst}] \quad \frac{\Gamma + [x : m_1] \vdash_{Dst} e : s \quad m_1 \leq m_2}{\Gamma + [x : m_2] \vdash_{Dst} e : s} [WEAK_{Dst}] \\
\\
\frac{}{\emptyset \vdash_{Dst} c : s} [LIT_{Dst}] \quad \frac{}{[x : s] \vdash_{Dst} x : s} [VAR_{Dst}] \quad \frac{}{[x : s] \vdash x @ r : s} [COPY_{Dst}] \\
\\
\frac{\Gamma_1 \vdash_{Dst} e_1 : s \quad \Gamma_2 + [x_1 : m] \vdash_{Dst} e_2 : s \quad \forall x \in \text{dom } \Gamma_1. \Gamma_1(x) \in \{r, d\} \Rightarrow x \notin \text{fv}(e_2)}{\Gamma_1 \sqcup \Gamma_2 \vdash_{Dst} \mathbf{let } x_1 = e_1 \mathbf{ in } e_2 : s} [LET_{Dst}] \\
\\
\frac{R = \bigcup_{i=1}^n \{ \text{sharerec}(a_i, f \bar{a}_i^n @ \bar{r}_j^m) \setminus \{a_i\} \mid m_i = d \} \quad \bigwedge_{m_i=d} \text{isTree}(a_i) \quad \Gamma = \Gamma_R + [f : \bar{m}_i^n \rightarrow s] + \bigoplus_{i=1}^n [a_i : m_i] \quad \Gamma_R = [y : r \mid y \in R]}{\Gamma \vdash_{Dst} f \bar{a}_i^n @ \bar{r}_j^m : s} [APP_{Dst}] \\
\\
\frac{\Gamma = \bigoplus_{i=1}^n [a_i : s]}{\Gamma \vdash_{Dst} C \bar{a}_i^n @ r : s} [CONS_{Dst}] \\
\\
\frac{\forall i \in \{1..n\}. \Gamma + [\bar{x}_{ij} : \bar{m}_{ij}^{n_i}] \vdash_{Dst} e_i : s}{\Gamma \sqcup [x : s] \vdash_{Dst} \mathbf{case } x \mathbf{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i} : s} [CASE_{Dst}] \\
\\
\frac{R = \text{sharerec}(x, \mathbf{case } x \mathbf{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i} \setminus \{x\} \quad \forall i \in \{1..n\}. \forall j \in \{1..n_i\}. \text{inh}_{i,C_i}(m_{ij}) \quad \forall z \in R \cup \{x\}. \forall i \in \{1..n\}. z \notin \text{fv}(e_i) \quad \forall i \in \{1..n\}. \Gamma + [\bar{x}_{ij} : \bar{m}_{ij}^{n_i}] \vdash_{Dst} e_i : s \quad \Gamma_R = [y : r \mid y \in R]}{\Gamma_R \sqcup (\Gamma \setminus x) + [x : d] \vdash_{Dst} \mathbf{case! } x \mathbf{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i} : s} [CASE!_{Dst}]
\end{array}$$

Figure 3.19: Typing rules for deriving  $\vdash_{Dst}$  judgements.

$$\begin{array}{ll}
\text{inh}_{i,C}(s) & \text{if } i \notin \text{RecPos}(C) \\
\text{inh}_{i,C}(d) & \text{if } i \in \text{RecPos}(C)
\end{array}$$

Figure 3.20: Inheritance compatibility definitions for the  $\vdash_{Dst}$  rules.

$\circ : \text{SafeType} \times \text{Mark} \rightarrow \text{ExpType}$

$$\begin{array}{l}
B \circ m = B \\
\alpha \circ m = \alpha \\
T \bar{s}_i^n @ \bar{\rho}_j^m \circ s = T \bar{s}_i^n @ \bar{\rho}_j^m \\
T \bar{s}_i^n @ \bar{\rho}_j^m \circ r = T \bar{s}_i^n \# @ \bar{\rho}_j^m \\
T \bar{s}_i^n @ \bar{\rho}_j^m \circ d = T \bar{s}_i^n ! @ \bar{\rho}_j^m
\end{array}$$

Figure 3.21: Compositionality of types.

**Definition 3.47.** Two environments  $\Gamma_R$  and  $\Gamma_D$  are composable if the following conditions hold:

1.  $\text{dom } \Gamma_D = \text{dom } \Gamma_R \setminus (\mathbf{RegType} \cup \mathbf{Cons})$
2. For every variable  $x \in \text{dom } \Gamma_D$ ,  $\Gamma_R(x) \circ \Gamma_D(x)$  is well-defined.
3. For every function symbol  $f \in \text{dom } \Gamma_D$ ,  $\Gamma_R(f) \circ \Gamma_D(f)$  is well-defined.

In this case, we define the composition of two typing environments  $\Gamma_R \circ \Gamma_D$  as follows:

$$\begin{aligned} \Gamma_R \circ \Gamma_D &\stackrel{\text{def}}{=} [x \mapsto \Gamma_R(x) \circ \Gamma_D(x) \quad | \quad x \in \text{dom } \Gamma_D] \\ &+ [f \mapsto \Gamma_R(f) \circ \Gamma_D(f) \quad | \quad f \in \text{dom } \Gamma_D] \\ &+ [r \mapsto \Gamma_R(r) \quad | \quad r \in \text{dom } \Gamma_R] \\ &+ [C \mapsto \Gamma_R(C) \quad | \quad C \in \text{dom } \Gamma_R] \end{aligned}$$

Now we can state formally the equivalence between the two smaller type systems and the original one. Before this, we need two auxiliary lemmas, whose proof is straightforward and will be omitted here.

**Lemma 3.48** (Environment reduction). *If  $x$  (resp.  $r$ ,  $f$ ) does not occur in  $e$ , and it does belong to the domain of  $\Gamma$ , then for any  $s' \in \mathbf{SafeType}$ :*

$$\Gamma + [x : s'] \vdash_{\text{Reg}} e : s \Leftrightarrow \Gamma \vdash_{\text{Reg}} e : s$$

and respectively with  $[r : \rho]$  and  $[f : sf]$  for any  $\rho \in \mathbf{RegType}$  and  $sf \in \mathbf{SafeFunType}$ .

*Proof.* By induction on the size of the corresponding  $\vdash_{\text{Reg}}$  derivation. □

**Lemma 3.49.** *Given a region environment  $\Gamma$  (resp. mark environment) such that  $\Gamma \vdash_{\text{Reg}} e : s$  (resp.  $\Gamma \vdash_{\text{Dst}} e : s$ ). If  $x \in \text{fv}(e)$ , then  $x \in \text{dom } \Gamma$ . Similarly for the function and constructor symbols being applied in  $e$ .*

*Proof.* By induction on the size of the  $\vdash_{\text{Reg}}$  (resp.  $\vdash_{\text{Dst}}$ ) derivation. □

**Theorem 3.50.** *For any two composable environments  $\Gamma_R$  and  $\Gamma_D$ , any expression  $e$  and any safe type  $s$ :*

$$\Gamma_R \circ \Gamma_D \vdash e : s \iff \Gamma_R \vdash_{\text{Reg}} e : s \wedge \Gamma_D \vdash_{\text{Dst}} e : s$$

*Proof.* Let us denote by  $\Gamma$  the result of  $\Gamma_R \circ \Gamma_D$ , which exists because  $\Gamma_R$  and  $\Gamma_D$  are composable. First we prove the  $(\Rightarrow)$  implication by induction on the  $\vdash$  derivation. We distinguish cases on the last rule applied:

- **Case [EXT]**

From one of the premises of this rule we get  $\Gamma' \vdash e : s$ , where  $\Gamma' \subseteq \Gamma$ . We can break down  $\Gamma'$  as follows  $\Gamma' = \Gamma'_R \circ \Gamma'_D$ . By induction hypothesis we obtain  $\Gamma'_R \vdash_{\text{Reg}} e : s$  and  $\Gamma'_D \vdash_{\text{Dst}} e : s$ . From the latter judgement we can derive  $\Gamma_D \vdash_{\text{Dst}} e : s$  by applying  $[\text{EXT}_{\text{Dst}}]$ , if necessary. The variables occurring in the domain of  $\Gamma$ , but not in the domain of  $\Gamma'$  do not appear free in  $e$ . Hence, by applying Lemmas 3.49 and 3.48 we obtain  $\Gamma_R \vdash_{\text{Reg}} e : s$ .

- **Case [WEAK]**

In this case we get  $\Gamma' \vdash e : s$  for some  $\Gamma' \leq \Gamma$ . If we split  $\Gamma'$  into  $\Gamma'_R \circ \Gamma'_D$  it holds that  $\Gamma'_R = \Gamma_R$ , since  $\Gamma$  and  $\Gamma'$  only differ in their marks. Hence, by induction hypothesis we get  $\Gamma'_R \vdash_{\text{Reg}} e : s$  and  $\Gamma'_D \vdash_{\text{Dst}} e : s$ . From the first fact we get  $\Gamma_R \vdash_{\text{Reg}} e : s$ , whereas in the second we can apply the  $[\text{WEAK}_{\text{Dst}}]$  rule.

- **Cases [LIT], [VAR], and [COPY]**

All these cases are trivial.

- **Case [LET]**

We get, for some  $\Gamma_1$  and  $\Gamma_2$  such that  $\Gamma = \Gamma_1 \sqcup \Gamma_2$ , the following judgements:

$$\Gamma_1 \vdash e_1 : s_1 \quad \Gamma_2 + [x_1 : \tau_1] \vdash e_2 : s$$

Each typing environment can be separated as follows:

$$\Gamma_1 = \Gamma_{1,R} \circ \Gamma_{1,D} \quad \Gamma_2 = \Gamma_{2,R} \circ \Gamma_{2,D}$$

By applying the induction hypothesis to each we obtain:

$$\begin{array}{ll} \Gamma_{1,R} \vdash_{Reg} e_1 : s_1 & \Gamma_{1,D} \vdash_{Dst} e_1 : s \\ \Gamma_{2,R} + [x_1 : s_1] \vdash_{Reg} e_2 : s & \Gamma_{2,D} + [x_1 : m] \vdash_{Dst} e_2 : s \end{array}$$

where  $s_1 \circ m = \tau_1$ . It is easy to show that  $\Gamma_{1,D} \sqcup \Gamma_{2,D}$  is equivalent to  $\Gamma_D$ . Moreover, the side condition of [LET] implies its counterpart in  $[LET_{Dst}]$ , as, for every  $x$ ,  $\Gamma_{1,D}(x) \in \{d, r\}$  implies  $\Gamma_1(x) \in \mathbf{UnsafeType}$ . Hence, by applying the  $[LET_{Dst}]$  rule to the judgements on the right-hand side, we get  $\Gamma_D \vdash_{Dst} e : s$ . Moreover, we know that the variables occurring in both  $\Gamma_{1,R}$  and  $\Gamma_{2,R}$  get the same type in both environments, because of the *utype?* constraint imposed by the  $\sqcup$  operator. Hence we can extend both environments by using Lemma 3.48 in order to get  $\Gamma_R \vdash_{Reg} e_1 : s_1$  and  $\Gamma_R + [x_1 : s_1] \vdash_{Reg} e_2 : s$ . Now we can apply the  $[LET_{Reg}]$  rule and obtain  $\Gamma_R \vdash_{Reg} e : s$ .

- **Case [APP]**

We get  $\Gamma = \Gamma_0 + [f : \forall \bar{\alpha} \bar{\rho}. tf] + [\bar{r}_j : \bar{\rho}_j^m] + \bigoplus_{i=1}^n [\bar{a}_i : \bar{t}_i^n]$ , where the  $\Gamma_R$  of the [APP] rule has been substituted by  $\Gamma_0$  to avoid confusion with the region environment  $\Gamma_R$ . The  $\Gamma$  environment is split into:

$$\begin{aligned} \Gamma_R &= \Gamma_{0,R} + [f : \forall \bar{\alpha} \bar{\rho}. sf] + [\bar{r}_j : \bar{\rho}_j^m] + \bigoplus_{i=1}^n [\bar{a}_i : \bar{s}_i^n] \\ \Gamma_D &= \Gamma_{0,D} + [f : \bar{m}_i \rightarrow m] + \bigoplus_{i=1}^n [\bar{a}_i : \bar{m}_i^n] \end{aligned}$$

where  $t_i = s_i \circ m_i$  for all  $i \in \{1..n\}$ . From the premises of the [APP] rule it follows  $\Gamma_D \vdash_{Dst} e : s$ . For applying the  $[APP_{Reg}]$  rule we have to show that  $\Gamma_{0,R} + \bigoplus_{i=1}^n [\bar{a}_i : \bar{s}_i^n] \vdash_{Reg} a_i : s_i$  for each  $i \in \{1..n\}$ , but we can prove this easily by case distinction on whether each  $a_i$  is a literal or a variable.

- **Case [CONS]**

Similar to the [APP] case.

- **Cases [CASE] and [CASE!]**

Again let us rename the  $\Gamma_R$  environment of [CASE!] by a less confusing name  $\Gamma_0$ . The premises of both judgements  $\Gamma_R \vdash_{Reg} e : s$  and  $\Gamma_D \vdash_{Dst} e : s$  follow trivially from their counterparts in the  $\vdash$  derivation. Before applying the  $[CASE!_{Reg}]$  rule, we may need to apply Lemma 3.48 to the

$\vdash_{Reg}$  typing of each **case** branch, in order to append the  $\Gamma_0$  environment and be able to derive  $\Gamma_R \vdash_{Reg} e : s$ .

Now we prove the ( $\Leftarrow$ ) implication by induction on the  $\vdash_{Dst}$  derivation. In some cases we will prove that  $\Gamma'_R \circ \Gamma_D \vdash e : s$  holds for some subset  $\Gamma'_R$  of  $\Gamma_R$  such that  $\text{dom } \Gamma_R \setminus \text{dom } \Gamma'_R \subseteq \mathbf{RegVar} \cup \mathbf{Cons}$ . From this judgement we can apply the [EXT] rule of the type system in order to obtain the desired result:  $\Gamma_R \circ \Gamma_D \vdash e : s$ . Again, we distinguish cases on the last rule applied in the  $\vdash_{Dst}$  derivation:

- **Case** [EXT<sub>Dst</sub>]

From this rule we get  $\Gamma'_D \vdash_{Dst} e : s$  for some  $\Gamma'_D \subseteq \Gamma_D$ . Since every variable  $x$  occurring in  $\text{dom } \Gamma_D \setminus \text{dom } \Gamma'_D$  does not occur free in  $e$ , we can remove the bindings of those variables in  $\text{dom } \Gamma_D \setminus \text{dom } \Gamma'_D$  from  $\Gamma_R$  by applying Lemma 3.49 in order to obtain an environment  $\Gamma'_R$  composable with  $\Gamma'_D$  such that  $\Gamma'_R \vdash_{Reg} e : s$ . By induction hypothesis,  $\Gamma'_R \circ \Gamma'_D \vdash e : s$ . Moreover,  $\Gamma'_R \circ \Gamma'_D \subseteq \Gamma$ . Hence we can apply [EXT] so as to get  $\Gamma \vdash e : s$ .

- **Case** [WEAK<sub>Dst</sub>]

Assume  $\Gamma_D + [x : m_2] \vdash e : s$  for some  $x$  and  $m_2$ . By the [WEAK<sub>Dst</sub>] rule there exists another mark  $m_1 \leq m_2$  such that  $\Gamma_D + [x : m_1] \vdash e : s$ . If  $\Gamma_D + [x : m_2]$  is composable with  $\Gamma_R$ , so is  $\Gamma_D + [x : m_1]$ , since they only differ in their marks. By induction hypothesis we get  $(\Gamma_D + [x : m_1]) \circ \Gamma_R \vdash e : s$ , in which  $\Gamma_R(x) \circ m_1 \leq \Gamma_R(x) \circ m_2$ . We can apply the [WEAK] rule of the type system so as to get the desired result.

- **Cases** [LIT<sub>Dst</sub>], [VAR<sub>Dst</sub>], and [COPY<sub>Dst</sub>]

The theorem holds trivially in all these cases. It is possible that the environment  $\Gamma_R$  contains more bindings than  $\Gamma_D$ , but these extra bindings must be of the form  $[r : \rho]$  or  $[C : \sigma]$ , since  $\Gamma_R$  and  $\Gamma_D$  are composable. We can discard these bindings (Lemma 3.48) from  $\Gamma_R$  before applying the corresponding  $\vdash$  rule ([LIT], [VAR], [COPY]) so as to get  $\Gamma'_R \circ \Gamma_D \vdash e : s$  for some  $\Gamma'_R \subseteq \Gamma_R$ . After this, we apply the [EXT] rule to recover these bindings and get  $\Gamma_R \circ \Gamma_D \vdash e : s$ .

- **Case** [LET<sub>Dst</sub>]

In this case  $\Gamma_D = \Gamma_{1,D} \sqcup \Gamma_{2,D}$  for some  $\Gamma_{1,D}, \Gamma_{2,D}$ . With the first environment we can type the auxiliary expression:  $\Gamma_{1,D} \vdash_{Dst} e_1 : s$ , whereas we get  $\Gamma_R \vdash_{Reg} e_1 : s_1$ . However, since  $\text{dom } \Gamma_{1,D} \subseteq \text{dom } \Gamma_D \subseteq \text{dom } \Gamma_R$ , we can remove bindings from  $\Gamma_R$  (by using Lemma 3.48) in order to get a typing environment  $\Gamma_{1,R}$  composable with  $\Gamma_{1,D}$ , since every variable not occurring in  $\Gamma_{1,D}$  is not free in  $e_1$  (as Lemma 3.49 states).

On the other hand, we have  $\Gamma_{2,D} + [x_1 : m] \vdash_{Dst} e_2 : s$  for some mark  $m$  and we can also remove bindings from  $\Gamma_R$  in order to obtain a  $\Gamma_{2,R}$  composable with  $\Gamma_{2,D}$  in which  $\Gamma_{2,R} + [x_1 : s] \vdash_{Reg} e_2 : s$ .

By applying the induction hypothesis on the two  $\vdash_{Dst}$  judgements, we get  $\Gamma_{1,R} \circ \Gamma_{1,D} \vdash e_1 : s_1$  and  $\Gamma_{2,R} \circ \Gamma_{2,D} + [x_1 : \tau_1] \vdash e_2 : s$ , where  $\tau_1 = s_1 \circ m$ . Moreover, if  $x$  has an unsafe type in  $\Gamma_{1,R} \circ \Gamma_{1,D}$ , then  $\Gamma_{1,D}(x) \in \{d, r\}$ , so the side condition of [LET] follows from its counterpart in [LET<sub>Dst</sub>]. It is obvious that  $utype?(s_1, \tau_1)$ , so let us apply the [LET] rule in order to obtain:

$$(\Gamma_{1,R} \circ \Gamma_{1,D}) \sqcup (\Gamma_{2,R} \circ \Gamma_{2,D}) \vdash e : s$$

Now we have to prove that  $(\Gamma_{1,R} \circ \Gamma_{1,D}) \sqcup (\Gamma_{2,R} \circ \Gamma_{2,D})$  is well-defined and equal to  $\Gamma_R \circ \Gamma_D$ . The well-definedness follows from  $\Gamma_{1,R}$  and  $\Gamma_{2,R}$  mapping common variables to the same safe type

(because both environments are subsets of a same  $\Gamma_R$ ). For the inclusion  $(\Gamma_{1,R} \circ \Gamma_{1,D}) \sqcup (\Gamma_{2,R} \circ \Gamma_{2,D}) \subseteq \Gamma_R \circ \Gamma_D$ , we distinguish cases on whether a given variable belongs to the domain of  $(\Gamma_{1,R} \circ \Gamma_{1,D})$ , or the domain of  $(\Gamma_{2,R} \circ \Gamma_{2,D})$ , or both. In any case the inclusion holds. The  $\supseteq$  inclusion is a direct consequence of the fact that  $\text{dom } \Gamma_D = \text{dom } \Gamma_{1,D} \cup \text{dom } \Gamma_{2,D}$ , so each variable in the domain of  $\Gamma_R \circ \Gamma_D$  must also appear in the domain of  $(\Gamma_{1,R} \circ \Gamma_{1,D}) \sqcup (\Gamma_{2,R} \circ \Gamma_{2,D})$ , and also associated to the same type, because of the  $\subseteq$  inclusion.

- **Case**  $[APP_{Dst}]$

By the rule  $[APP_{Reg}]$  we get  $\Gamma_R \vdash_{Reg} a_i : s_i$ , so the binding  $[a_i : s_i]$  is contained within  $\Gamma_R$  for those  $a_i$  that are not literals. However, because our convention of allowing literals in an environment with their respective type, we shall consider that  $\bigoplus_{i=1}^n [a_i : s_i] \subseteq \Gamma_R$ . By Lemma 3.48, we can discard bindings from  $\Gamma_R$  until we obtain:

$$\Gamma'_R = \Gamma_{0,R} + \bigoplus_{i=1}^n [a_i : s_i] + [\overline{r_j} : \overline{\rho_j^m}] + [f : \forall \overline{\alpha} \overline{\rho} . \overline{s_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s]$$

whereas:

$$\Gamma_D = \Gamma_{0,D} + \bigoplus_{i=1}^n [a_i : m_i] + [f : \overline{m_i^n} \rightarrow s]$$

and  $\Gamma_{0,R}$  is composable with  $\Gamma_{0,D}$ . Therefore:

$$\Gamma'_R \circ \Gamma_D = (\Gamma_{0,R} \circ \Gamma_{0,D}) + \bigoplus_{i=1}^n [a_i : s_i \circ m_i] + [\overline{r_j} : \overline{\rho_j^m}] + [f : \forall \overline{\alpha} \overline{\rho} . \overline{s_i^n} \circ \overline{m_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s]$$

A given variable has a  $d$  mark in  $\Gamma_D$  if and only if it has a condemned type in  $\Gamma'_R \circ \Gamma_D$ . So, the set domain of the  $\Gamma_R$  environment in the  $[APP]$  rule (which we have denoted  $\Gamma_0$  in this proof, to avoid confusion) is the same as that its counterpart in the  $[APP_{Dst}]$  rule. Thus we can apply  $[APP]$  in order to get  $\Gamma'_R \circ \Gamma_D \vdash e : s$ . Since  $\Gamma'_R \subseteq \Gamma_R$  and  $\Gamma_R$  is composable with  $\Gamma_D$ , the only bindings that can appear in  $\Gamma_R$  but not in  $\Gamma'_R$  are those of the form  $[r : \rho]$  and  $[C : \sigma]$ . We can extend the environment  $\Gamma'_R \circ \Gamma_D$  with the help of the  $[EXT]$  rule in order to get  $\Gamma_R \circ \Gamma_D \vdash e : s$ .

- **Case**  $[CONS_{Dst}]$

The proof is very similar to that of  $[APP_{Dst}]$  rule.

- **Case**  $[CASE_{Dst}]$

Let  $n$  be the number of case branches. From the premises in the  $[CASE_{Dst}]$  and  $[CASE_{Reg}]$  rules we obtain, for each  $i \in \{1..n\}$ :

$$\Gamma_R + [x_{ij} : s_{ij}] \vdash_{Reg} e_i : s$$

$$\Gamma_D + [x_{ij} : m_{ij}] \vdash_{Dst} e_i : s$$

Both typing environments are composable, and hence we can apply the induction hypothesis in order to infer, for each  $i \in \{1..n\}$ :  $\Gamma_R \circ \Gamma_D + [x_{ij} : s_{ij} \circ m_{ij}] \vdash e_i : s$ . The *utype?* predicates in the  $[CASE]$  rule hold because the  $s_{ij}$  in those judgements match the types of their respective constructors, as the  $[CASE_{Reg}]$  rule demands. The rest of the premises of the rule  $[CASE]$  are direct consequences of their counterparts in  $[CASE_{Reg}]$  and  $[CASE_{Dst}]$ , so the theorem holds by applying  $[CASE]$ .

- **Case**  $[CASE!_{Dst}]$

Let  $n$  be the number of case branches. We obtain  $\Gamma_D = \Gamma_{0,D} \sqcup (\Gamma'_D \setminus x) + [x : d]$  for some  $\Gamma'_D$ . We can also split  $\Gamma_R$  in the same way so that  $\Gamma_R = \Gamma_{0,R} \sqcup (\Gamma'_R \setminus x) + [x : T @ \rho]$  such that  $\Gamma'_R$  and  $\Gamma'_D$  are composable. From the premises of each respective rules we obtain:

$$\begin{aligned} & \Gamma'_D \setminus x + [x : m_i] + [\overline{x_{ij} : m_{ij}^{n_i}}] \vdash_{Dst} e_i : s \\ & \Gamma_{0,R} \sqcup (\Gamma'_R \setminus x) + [x : T @ \rho] + [\overline{x_{ij} : s_{ij}^{n_i}}] \vdash_{Reg} e_i : s \end{aligned}$$

for each  $i \in \{1..n\}$ . In case  $m_i$  does not occur in the  $\vdash_{Dst}$  judgement we can add it via the  $[EXT_{Dst}]$  rule. We can also remove the  $\Gamma_{0,R}$  from the  $\vdash_{Reg}$  judgements by using Lemma 3.48 (since the corresponding variables do not appear in the environment of the  $\vdash_{Dst}$  derivations, they do not occur free in the  $e_i$ ). For each branch, the resulting region environments are composable with the mark environments, so we can apply the induction hypothesis and obtain, for each  $i \in \{1..n\}$

$$(\Gamma'_R \circ \Gamma'_D) \setminus x + [x : \tau] + [\overline{x_{ij} : s_{ij} \circ m_{ij}^{n_i}}] \vdash e_i : s$$

for some  $\tau$ . The *inh* predicates in rule  $[CASE!]$ , as well as the rest of the premises in this rule can be easily derived from their counterparts in  $[CASE!_{Dst}]$  and  $[CASE(!)_{Reg}]$ .

□

### 3.8 Conclusions and related work

In this chapter we have introduced a destruction-aware type system for *Safe* and proved it correct, in the sense that the live part of the heap will never contain dangling pointers. We have shown the expressiveness of the type system by means of several examples. Finally, we have defined two auxiliary systems which allow us to consider regions and explicit destruction separately, and will serve as a basis for the correctness results of the inference algorithms explained in the next two chapters.

Regarding regions, our type system presents obvious similarities with that of Tofte and Talpin [116], since our region model is inspired by the latter. Their correctness proof uses an involved coinductive approach for showing the preservation of consistency under evaluation. Most of the technical difficulty lies in the presence of arrow effects to reflect which regions are being read and written in each sub-expression. These effects are not necessary in *Safe*'s type system, since there is a single local RTV in every function ( $\rho_{self}$ ). When considering this function from outside, the read and write effects on this RTV are hidden (i.e. are not reflected in the function's type). This considerably simplifies the consistency preservation proof (Theorem 3.26) and the region inference algorithm.

There has been research work aimed at providing a more intuitive understanding of the proofs in [116]. In particular, Boudol [21] extends Tofte and Talpin's calculus with a region deallocation construct which, on the one hand, allows the programmer to break the nested nature of regions, and, on the other hand, allows to simplify the original proof of [116]. A difference with respect our system is the presence of effects (Boudol introduces a new kind of negative effects), and the lack of an inference algorithm.

With regard to explicit destruction, our safety type system has some characteristics of linear types (see [118] as a basic reference). A number of variants of linear types have been developed for years for coping with the related problems of achieving safe in-place updates in functional languages [95] or detecting program sites where values could be safely deallocated [66]. The work closest to our system

is that of Aspinall and Hofmann [10], which proposes a type system for a language explicitly reusing heap cells. They prove that well-typed programs can be safely translated into an imperative language with an explicit deallocation/reusing mechanism. Their typing scheme includes three usage aspects for the variables occurring in the type environment. These are: (1) variable used destructively, (2) variable used read-only, but shares a component with the result, and (3) variable used read-only, and not shared with the result. We summarise here the differences and similarities with our work.

There are non-essential differences such as: (1) they only admit algorithms running in constant heap space, i.e. for each allocation there must exist a previous deallocation; (2) they use at the source level an explicit parameter  $d$  representing a pointer to the cell being reused; and (3) they distinguish two different Cartesian products depending on whether there is sharing or not between the tuple components. A major difference with respect to our initial type system of [84] is the use of a total order in their usage aspects, whereas we had defined a partial order in our marks. In this thesis we have improved the work in [84] in order to incorporate a total order in this analysis. In fact, the [WEAK] rule of Figure 3.6 has been inspired by the [DROP] rule of [10]. This change has had a positive impact in our system: on the one hand, it allows the type system to accept more pointer-safe functions, and, on the other hand, it considerably simplifies the inference algorithm (see Section 5.5). Besides this, there are some other essential differences with respect to [10]:

1. Their aspects 2 and 3 (read-only and shared, or just read-only) could be roughly assimilated to our use  $s$  (read-only), and their aspect 1 (destructive), to our use  $d$  (condemned), both defined in Section 3.3. We add a third use  $r$  (in-danger) arising from a sharing analysis based on abstract interpretation [98]. This use allows us to know more precisely which variables are in danger when a DS is destroyed.
2. We make a distinction between the recursive spine of a data structure, and the cells beyond that recursive spine. As a consequence, variables pointing to a non-recursive sub-structure of a destroyed cell can be safely accessed. Konečný introduces in [69] a set of separation assertions for distinguishing between the data and the spine in a DS. However, it is unclear how this system interacts with the different usage aspects of [10].
3. We use a separate analysis [98] for approximating the sharing relations, whereas in [10] sharing is tracked by means of the usage aspect 2.
4. Our typing rule for **let**  $x_1 = e_1$  **in**  $e_2$  allows more usage aspects combinations than theirs. Let  $i \in \{1, 2, 3\}$  the aspect assigned to  $x_1$ ,  $j$  the aspect of a variable  $z$  in  $e_1$ , and  $k$  the aspect of the variable  $z$  in  $e_2$ . We allow the following combinations  $(i, j, k)$  that they forbid:  $(1, 2, 2)$ ,  $(1, 2, 3)$ ,  $(2, 2, 2)$ ,  $(2, 2, 3)$ . The deep reason is our more precise sharing information and the new in-danger type. In a more recent version of this system [11] combination  $(2, 2, 3)$  is allowed.

An example of a *Safe* program using the combination  $(1, 2, 3)$  is the following:

$$\mathbf{let} \ x = z : [] \ \mathbf{in} \ \mathbf{case!} \ x \ \mathbf{of} \ \dots \ \mathbf{case} \ z \ \mathbf{of} \ \dots$$

Variable  $x$  is destroyed, but a sharing variable  $z$  can be read both in the auxiliary and in the main expression. An example of *Safe* program using the combination  $(1, 2, 2)$  is the following:

$$\mathbf{let} \ x = z : [] \ \mathbf{in} \ \mathbf{case!} \ x \ \mathbf{of} \ \dots \ z$$

Here, the result  $z$  shares the destroyed variable  $x$ . Both programs take profit from the fact that the

sharing variable  $z$  is not a recursive descendant of  $x$ . Our type system assigns an  $s$ -type to these variables.

# Chapter 4

## Region Inference

### 4.1 Introduction

The type system of Chapter 3 establishes the safety of region-annotated *Core-Safe* programs. Given a *Safe* program without regions, our aim in this chapter is to annotate this program with region variables, so that the result is typable by using the  $\vdash_{Reg}$  rules of Figure 3.18. We define a region inference algorithm that decorates data types and function definitions by maximizing the number of cells built in the working region *self*. Regarding this, the inference algorithm yields *optimal* programs with respect to the type system. This means that, if the programmer were allowed to annotate *Safe* programs by himself, he would not achieve a better result than the inference algorithm. Any attempt to improve the results of the inference algorithm (for example, by placing data structures in the *self* region, whereas the algorithm places them in an output region) leads to an ill-typed program.

**Example 4.1.** Let us consider the following function definition implementing the Quicksort algorithm. We show here the *Full-Safe* code, rather than its desugared version, for the sake of clarity:

$$\begin{aligned} \text{qsort } [] &= [] \\ \text{qsort } (x : xs) &= \text{append } ls' (x : gs') \\ &\quad \text{where } (ls, gs) = \text{partition } x \ xs \\ &\quad \quad ls' = \text{qsort } ls \\ &\quad \quad gs' = \text{qsort } gs \end{aligned}$$

Let us assume that the region inference algorithm has already been applied to the *partition* and *append* functions, obtaining the following type schemes:

$$\begin{aligned} \text{partition} &:: \text{Int} \rightarrow [\text{Int}]@_{\rho_1} \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow \rho_4 \rightarrow ([\text{Int}]@_{\rho_2}, [\text{Int}]@_{\rho_3})@_{\rho_4} \\ \text{append} &:: [\alpha]@_{\rho_1} \rightarrow [\alpha]@_{\rho_2} \rightarrow \rho_2 \rightarrow [\alpha]@_{\rho_2} \end{aligned}$$

The *partition* function receives three region arguments: the first and second one are the regions where the resulting lists containing the elements that are lower or greater than the pivot (first parameter) will be allocated. The last region argument specifies the region in which the tuple grouping these results is stored. The *append* function receives a parameter indicating where to build the result, which has to be placed in the same region as the second parameter.

There are several ways in which a given function can be region-annotated. In particular, the follow-

ing region-annotated version of  $qsort$  would be accepted by the type system:

$$\begin{aligned}
qsort &:: [Int]@_{\rho_0} \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow [Int]@_{\rho_2} \\
qsort [] &@_{r_1 r_2 r_3} = []@_{r_2} \\
qsort (x : xs) &@_{r_1 r_2 r_3} = \text{append } ls' (x : gs')@_{r_2} @_{r_2} \\
&\mathbf{where} (ls, gs) = \text{partition } x \text{ } xs @_{r_1 r_2 r_3} \\
&ls' = qsort \text{ } ls @_{r_1 r_2 r_3} \\
&gs' = qsort \text{ } gs @_{r_1 r_2 r_3}
\end{aligned}$$

The elements of the input list being lower than the pivot are stored in the region corresponding to  $r_1$ , whereas  $r_3$  contains the region in which the tuples returned by the  $partition$  function are built. However, these DS are intermediate results that are not part of the sorted list returned by the  $qsort$  function. Consequently, the following choice of region annotations is better from the point of view of memory consumption, while still being accepted by the type system:

$$\begin{aligned}
qsort [] &@_{r_2} = []@_{r_2} \\
qsort (x : xs) &@_{r_2} = \text{append } ls' (x : gs')@_{r_2} @_{r_2} \\
&\mathbf{where} (ls, gs) = \text{partition } x \text{ } xs @_{self \text{ } self \text{ } self} \\
&ls' = qsort \text{ } ls @_{self} \\
&gs' = qsort \text{ } gs @_{r_2}
\end{aligned}$$

Since  $r_1$  and  $r_3$  are not used anymore in the function's body, we can remove them from the list of parameters. Now the result of  $partition$  resides completely in the working region. This is the version inferred by the algorithm subject of this chapter. If we tried to improve the results by placing the  $gs'$  list in the  $self$  region, then the  $(x : gs')$  list would also be forced to reside in this region. The same would happen with the result of  $append$  and the empty list being built in the base case. The region  $r_1$  would not be used anymore, so we would get the following region-annotated program:

$$\begin{aligned}
qsort [] &= []@_{self} \\
qsort (x : xs) &= \text{append } ls' ((x : gs')@_{self}) @_{self} \\
&\mathbf{where} (ls, gs) = \text{partition } x \text{ } xs @_{self \text{ } self \text{ } self} \\
&ls' = qsort \text{ } ls \\
&gs' = qsort \text{ } gs
\end{aligned}$$

which is ill-typed, since the type of the result is  $[Int]@_{\rho_{self}}$ , and the [FUNB] rule of the type system (Figure 3.9) could not be applied, because  $\rho_{self}$  occurs in the type of the result. This means that, at runtime, the resulting sorted list would be built in the working region and would be disposed of immediately after the call to  $qsort$ .  $\square$

As it was explained in Section 3.4, *Safe* supports polymorphic recursion on RTVs. This implies that, given a recursive function, the RTVs in the type of the recursive calls need not be the same as the RTVs in the type of the function. This can significantly decrease the memory needs of some functions, as the following example (taken from [116]) shows:

**Example 4.2.** Assume the following data type for heap integers consisting of a single constructor  $HInt$  taking an integer as a parameter:

$$\mathbf{data} \text{ } HInt @ \rho = HInt \text{ } Int @ \rho$$

The addition function on heap integers is defined as follows:

$$\text{add} (\text{HInt } n) (\text{HInt } m) = \text{HInt } (n + m)$$

Our aim is to annotate the following function for computing Fibonacci numbers:

$$\begin{aligned} \text{fib} (\text{HInt } 0) &= \text{HInt } 0 \\ \text{fib} (\text{HInt } 1) &= \text{HInt } 1 \\ \text{fib} (\text{HInt } n) &= \text{add} (\text{fib} (\text{HInt } (n - 1))) (\text{fib} (\text{HInt } (n - 2))) \end{aligned}$$

Without allowing polymorphic recursion, the input given to the recursive calls to *fib* would be forced to live in the same region as the input given to the root call. Similarly, the partial results returned by the recursive calls would have to live in the same region as the global result given by the root call. As a consequence, we would obtain the following region-annotated definition:

$$\begin{aligned} \text{fib} &:: \text{HInt } @ \rho_1 \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \text{HInt } @ \rho_2 \\ \text{fib} (\text{HInt } 0) @ r_1 r_2 &= \text{HInt } 0 @ r_2 \\ \text{fib} (\text{HInt } 1) @ r_1 r_2 &= \text{HInt } 1 @ r_2 \\ \text{fib} (\text{HInt } n) @ r_1 r_2 &= \text{add} (\text{fib} (\text{HInt } (n - 1) @ r_1) @ r_2) (\text{fib} (\text{HInt } (n - 2) @ r_1) @ r_2) @ r_2 \end{aligned}$$

Although the algorithm produces a great amount of intermediate results (namely, the heap integers passed to the recursive calls and the integers returned by those calls), none of these results is stored in the working region. As a consequence, the number of memory cells needed for evaluating the call  $\text{fib} (\text{HInt } n)$  is proportional to  $2^n$ . If we allow polymorphic recursion, we get the following region-annotated version:

$$\begin{aligned} \text{fib} &:: \text{HInt } @ \rho_1 \rightarrow \rho_2 \rightarrow \text{HInt } @ \rho_2 \\ \text{fib} (\text{HInt } 0) @ r_2 &= \text{HInt } 0 @ r_2 \\ \text{fib} (\text{HInt } 1) @ r_2 &= \text{HInt } 1 @ r_2 \\ \text{fib} (\text{HInt } n) @ r_2 &= \text{add} (\text{fib} (\text{HInt } (n - 1) @ \text{self}) @ \text{self}) (\text{fib} (\text{HInt } (n - 2) @ \text{self}) @ \text{self}) @ r_2 \end{aligned}$$

In this case, the RTVs in the type of the recursive occurrences of *fib* are region instances of the general type scheme inferred for this function:

$$\begin{aligned} \text{fib function} &: \forall \rho_1 \rho_2. \text{HInt } @ \rho_1 \rightarrow \rho_2 \rightarrow \text{HInt } @ \rho_2 \\ \text{Recursive calls} &: \text{HInt } @ \rho_{\text{self}} \rightarrow \rho_{\text{self}} \rightarrow \text{HInt } @ \rho_{\text{self}} \end{aligned}$$

The intermediate heap integers generated by *fib* are stored in the working region. This results in an algorithm of linear memory complexity, instead of exponential. As an additional advantage, the latter version does not need the region parameter  $r_1$ .  $\square$

The region inference problem was already addressed by Tofte and Talpin in the context of the MLKit language [116, 113]. Their inference algorithm, described in [112], deals with higher-order programs written in Standard ML. Since (at this moment) *Safe* is a first-order language, region inference can be expected to be simpler and more efficient than that of MLKit. Their algorithm runs in time  $\mathcal{O}(n^4)$  in the worst case, where  $n$  is the size of the term, including in it the Hindley-Milner type annotations. The explanation of the algorithm and of its correctness arguments [112] needed around 40 pages of dense writing. So, it is not an easy task to incorporate the MLKit ideas into a new language.

<b>Prog</b> <sub>0</sub> ∋	<i>prog</i> <sub>0</sub>	→	$\overline{data_{0,i}; def_{0,i}}; \epsilon_0$	
<b>DecData</b> <sub>0</sub> ∋	<i>data</i> <sub>0</sub>	→	<b>data</b> $T \overline{a_i} = C_i \overline{t_{ij}^{n_i}}$	
<b>DecFun</b> <sub>0</sub> ∋	<i>def</i> <sub>0</sub>	→	$f \overline{x_i} = \epsilon_0$	
<b>Exp</b> <sub>0</sub> ∋	$\epsilon_0$	→	<i>a</i>	{literal <i>c</i> or variable <i>x</i> }
			<i>x</i> @	{copy}
			$C \overline{a_i}$	{constructor}
			$f \overline{a_i}$	{function application}
			<b>let</b> $x_1 = \epsilon_{0,1}$ <b>in</b> $\epsilon_{0,2}$	{let declaration}
			<b>case</b> <i>x</i> <b>of</b> $C_i \overline{x_{ij}^{n_i}} \rightarrow \epsilon_{0,i}$	{read only pattern matching}
			<b>case!</b> <i>x</i> <b>of</b> $C_i \overline{x_{ij}^{n_i}} \rightarrow \epsilon_{0,i}$	{destructive pattern matching}

Figure 4.1: Input language definition: *Core-Safe* without regions.

$$\begin{aligned}
&inferProgram :: \mathbf{SafeTypeEnv} \rightarrow \mathbf{Prog}_0 \rightarrow (\mathbf{SafeTypeEnv}, \mathbf{Prog}) \\
&inferProgram \Gamma (\overline{data_{0,i}; def_{0,i}}; \epsilon_0) = (\Gamma', \overline{data_i; def_i}; e) \\
&\quad \text{where } (\Gamma_d, \overline{data_i}) = \text{mapAccumL } inferDecData \Gamma \overline{data_{0,i}} \\
&\quad \quad (\Gamma', \overline{def_i}) = \text{mapAccumL } inferDef \Gamma_d \overline{def_{0,i}} \\
&\quad \quad e = inferMainExpression \Gamma' \epsilon_0
\end{aligned}$$

Figure 4.2: A high-level view of the Hindley-Milner and region inference algorithm.

This chapter is an extended version of [85], whose contribution is a simple region inference algorithm for *Safe*. If polymorphic recursion is not inferred, it runs in time  $\mathcal{O}(n)$  in the worst case, being  $n$  as above, while if polymorphic recursion appears, it needs time  $\mathcal{O}(n^2)$  in the worst case. Moreover, the first phase of the algorithm can be directly integrated in the usual Hindley-Milner type inference algorithm, just by considering RTVs as ordinary polymorphic type variables. The second phase involves very simple set operations and the computation of a fixed point. Unlike [112], termination is always guaranteed without special provisions. There, they had to sacrifice principal types in order to ensure termination. Due to its simplicity, we believe that our algorithm can be easily reused in a different first-order functional language featuring Hindley-Milner types.

The algorithm described in this chapter has been implemented as a part of the *Safe* compiler. Although the implementation works at the *Full-Safe* level, the description given here applies only to *Core-Safe*. Its extension to *Full-Safe* is straightforward. In Section 4.2 we give an overview of the inference algorithm. Sections 4.3 and 4.4 deal with the different phases of the algorithm, whereas in Section 4.5 we prove its correctness. Finally, we give some examples in Section 4.6.

## 4.2 A high level view of the algorithm

The algorithm receives a region-free *Core-Safe* program  $prog_0 \in \mathbf{Prog}_0$  and produces a region-annotated *Core-Safe* program  $prog \in \mathbf{Prog}$  as a result. The syntax of region-annotated programs has already been described in Section 2.3, whereas the syntax of the region-free programs is shown in Figure 4.1. The only difference between the two syntaxes is the absence of region types and region variables in the latter. We use  $data_0$ ,  $def_0$  and  $\epsilon_0$  to denote respectively data definitions, function definitions and expressions in this region-free language.

In Figure 4.2 we show, in a Haskell-like pseudo-code, the definition of the whole process. The function  $inferProgram$  receives, besides the region-free input program, a typing environment containing the types of every builtin constructor and function (e.g. arithmetic/logic operations and list constructors).

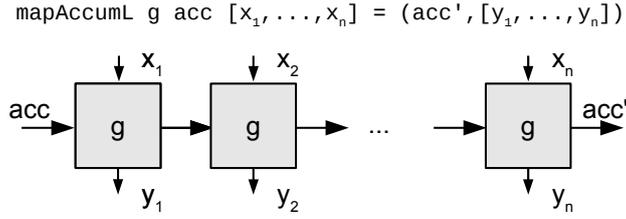


Figure 4.3: Visual description of the `mapAccumL` function, which behaves like `map`, but propagating an accumulator. It returns the transformed list and the final accumulator.

In addition to the final region-annotated program, this function returns the extension of the initial  $\Gamma$  with the types of each constructor or function being defined in the program. The function `mapAccumL`, described in [76], applies a function to each element of a list while passing an accumulator parameter from left to right (see Figure 4.3). Firstly, each **data** declaration is processed by *inferDecData*, which also extends the input environment  $\Gamma$  with the types of the constructors being defined in each data type. After this, the function definitions are processed by the function *inferDef*, which also augments the typing environment resulting from the previous phase with the inferred types for each function definition. Finally, the region annotations of the main expression are inferred.

### 4.3 Region inference of data declarations

In this phase the algorithm decorates the type and data constructors with region type variables. Each **data** declaration is processed individually. Our aim is to generate as many distinct RTVs as possible for each data constructor, since this implies that, at runtime, the data structures of this type can be distributed among several regions. In this way, we achieve more granularity for deciding which part of a data structure should be placed in the *self* region in order to be disposed of as soon as possible.

This phase also generates a typing environment  $\Gamma$  that maps every constructor name being defined to its type signature. We start from an initial environment containing the type of the following built-in constructors:

<code>[]</code>	:: $\forall \alpha \rho : \rho \rightarrow [\alpha]@ \rho$	{empty list constructor}
<code>(:)</code>	:: $\forall \alpha \rho : \alpha \rightarrow [\alpha]@ \rho \rightarrow \rho \rightarrow [\alpha]@ \rho$	{list constructor (head:tail)}
<code>(,)</code>	:: $\forall \alpha \beta \rho : \alpha \rightarrow \beta \rightarrow \rho \rightarrow (\alpha, \beta)@ \rho$	{pair constructor}

In this section we give an informal description of the region inference algorithm for the **data** declarations, together with a running example. Let us consider the following declaration:

$$\mathbf{data} \ T \ \alpha \ \beta = C \ (\alpha, \alpha) \ [\beta] \ | \ D \ \alpha \ (T \ \alpha \ \beta)$$

The steps of the algorithm are as follows:

1. Decorate each non-recursive nested data type with fresh RTVs. All these data types are assumed to have been inferred previously, so we know the exact number of RTVs needed in each occurrence. Basic types *Int* and *Bool* do not require a RTV, as their values do not reside by themselves in the heap. Built-in data types (tuples and lists) require one RTV.

In our example, we obtain:

$$\mathbf{data} \ T \ \alpha \ \beta = C \ (\alpha, \alpha)@_{\rho_1} \ [\beta]@_{\rho_2} \ | \ D \ \alpha \ (T \ \alpha \ \beta)$$

2. Generate an additional fresh RTV and decorate each constructor with it. This is the *outermost* RTV of the type being defined.

$$\mathbf{data} \ T \ \alpha \ \beta = C \ (\alpha, \alpha)@_{\rho_1} \ [\beta]@_{\rho_2} \ @ \ \rho_3 \ | \ D \ \alpha \ (T \ \alpha \ \beta) \ @ \ \rho_3$$

3. Add the RTVs generated in the two previous steps as extra region type parameters for the type constructor being defined. The last parameter must correspond to the outermost RTV obtained in step 2.

$$\mathbf{data} \ T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3 = C \ (\alpha, \alpha)@_{\rho_1} \ [\beta]@_{\rho_2} \ @ \ \rho_3 \ | \ D \ \alpha \ (T \ \alpha \ \beta) \ @ \ \rho_3$$

4. Now that we know which RTVs are associated with the current data type, we can annotate the recursive occurrences of the latter. Polymorphic recursion over region types is supported in function signatures, but not in data type definitions. Therefore the list of RTVs in the recursive occurrences must be the same as in the left-hand side of the definition.

In our example we get the following result:

$$\mathbf{data} \ T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3 = C \ (\alpha, \alpha)@_{\rho_1} \ [\beta]@_{\rho_2} \ @ \ \rho_3 \ | \ D \ \alpha \ (T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3) \ @ \ \rho_3$$

5. Finally, we compute the type signature for the data constructors from the types of their arguments.

$$\begin{aligned} C &:: \forall \alpha \ \beta \ \rho_1 \ \rho_2 \ \rho_3 : (\alpha, \alpha)@_{\rho_1} \rightarrow [\beta]@_{\rho_2} \rightarrow \rho_3 \rightarrow T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3 \\ D &:: \forall \alpha \ \beta \ \rho_1 \ \rho_2 \ \rho_3 : \alpha \rightarrow (T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3) \rightarrow \rho_3 \rightarrow T \ \alpha \ \beta \ @ \ \rho_1 \ \rho_2 \ \rho_3 \end{aligned}$$

This phase extends the typing environment  $\Gamma$  with the types of these constructors.

**Example 4.3.** Consider the following data type declarations without regions:

$$\begin{aligned} \mathbf{data} \ HInt &= HInt \ Int \\ \mathbf{data} \ Table &= TBL \ [(Int, HInt)] \end{aligned}$$

In the first declaration we have to add a RTV for the data constructor *HInt*:

$$\mathbf{data} \ HInt \ @ \ \rho = HInt \ Int \ @ \ \rho$$

In the second declaration we need four RTVs: one for each nested data type and another one for the data being defined:

$$\mathbf{data} \ Table \ @ \ \rho_1 \ \rho_2 \ \rho_3 \ \rho_4 = TBL \ [(Int, HInt \ @ \ \rho_1) \ @ \ \rho_2] \ @ \ \rho_3 \ @ \ \rho_4$$

□

$$\begin{aligned}
& inferDef :: \mathbf{SafeTypeEnv} \rightarrow \mathbf{DecFun}_0 \rightarrow (\mathbf{SafeTypeEnv}, \mathbf{DecFun}) \\
& inferDef \Gamma (f \bar{x}_i^n = \epsilon_0) = (\Gamma + [f : \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s], f \bar{x}_i^n @ \bar{r}_j^m = e) \\
& \quad \text{where } fresh(\alpha_0, \bar{\alpha}_i^n, \bar{r}_j^m) \\
& \quad \quad \Gamma + [f : \alpha_0] + [\bar{x}_i : \bar{\alpha}_i^n] \vdash_f \epsilon_0 \longrightarrow \epsilon : \alpha \mid E \quad (A) \\
& \quad \quad \langle \theta, \bar{\varphi}_i^p \rangle = solveEqs E \quad (B) \\
& \quad \quad \bar{s}_i^n \rightarrow s = \theta(\alpha_0) \quad (C) \\
& \quad \quad (R_{arg}, R_{self}, -) = inferRegions (\bar{s}_i^n \rightarrow s) \theta(\epsilon) \bar{\varphi}_i^p \quad (D) \\
& \quad \quad \{\bar{\rho}_j^m\} = R_{arg} \quad (E) \\
& \quad \quad \mathcal{R} = [\bar{\rho}_i \mapsto r_i^m] + [\rho \mapsto self \mid \rho \in R_{self}] \quad (F) \\
& \quad \quad e = annotateExp \mathcal{R} \epsilon \quad (G)
\end{aligned}$$

Figure 4.4: Region inference algorithm for function definitions.

## 4.4 Region inference of function definitions

Our region inference algorithm for function definitions is implemented as an extension of Hindley-Milner inference (HM in the following). In Figure 4.4 we give its definition. The first two steps (A) and (B) deal with HM inference. As usual in many of its presentations [99], we consider two sub-phases: generation of a system of equations between types and its later resolution via unification. This phase produces two results:

1. An intermediate expression  $\epsilon$  decorated with type information, which will be used in the subsequent phases.
2. A solution  $\langle \theta, \bar{\varphi}_i^p \rangle$  to the system of equations, whose meaning will be made precise later.

Section 4.4.1 describes this phase in detail. With the resulting substitution  $\theta$  we can obtain (step (C)) a preliminary type  $\bar{s}_i^n \rightarrow s$  (without regions) for the function. The next step (D), which is explained in Section 4.4.2, determines which RTVs have to appear as additional parameters in the function's signature ( $R_{arg}$ ) and which ones must be unified with  $\rho_{self}$  ( $R_{self}$ ). Once we have computed the corresponding  $R_{arg}$  and  $R_{self}$ , we build in (F) a correspondence  $\mathcal{R}$  between region types and freshly generated region variables. We apply this correspondence to the annotations of the intermediate expression  $\epsilon$  in order to obtain the final *Core-Safe* expression (step (G)). Besides this, the type of the function is augmented with the variables occurring in  $R_{arg}$  and included into the resulting typing environment.

### 4.4.1 HM Inference

Our implementation of this phase extends the usual HM inference in the following ways:

- **Generation of equations between region types:** Algebraic data types have already been annotated with RTVs by means of the algorithm explained in Section 4.3. The equations being generated in this phase include these region types, which are treated like ordinary polymorphic variables. In particular, some of them may be unified as a consequence of solving the system of equations.
- **Weak forms of unification equations:** In absence of polymorphic recursion, the type of the function being defined must be equal to the type of the function in each call. Since we allow polymorphic recursion over regions, we can relax this requirement, so the type of the function in each call site may be a *region type instance* of the function's type. This means that both types need not be syntactically equal after unification, but there must exist a list  $\bar{\varphi}_i$  of substitutions mapping the RTVs

$\mathbf{Exp}^1 \ni \epsilon \rightarrow$	$a$	{literal $c$ or variable $x$ }
	$  x@_{[\rho]}$	{copy}
	$  C \bar{a}_i_{[\rho]}$	{constructor}
	$  f \bar{a}_i_{[\bar{\rho}_j]}$	{nonrecursive function application}
	$  f \bar{a}_i_{[k, \bar{t}_i \rightarrow t]}$	{recursive function application}
	$  \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2$	{let declaration}
	$  \mathbf{case} \ x \ \mathbf{of} \ \bar{C}_i \ \bar{x}_{ij} \rightarrow \epsilon_i$	{read only pattern matching}
	$  \mathbf{case!} \ x \ \mathbf{of} \ C_i \ \bar{x}_{ij} \rightarrow \epsilon_i$	{destructive pattern matching}

Figure 4.5: Intermediate language for expressions.

of the function’s type into RTVs of the type of each recursive application. Besides polymorphic recursion, the *Safe* language has a copy expression, in which the types of source and destination DS may differ in the outermost region. For handling these cases, we introduce different kinds of constraints which impose fewer restrictions than unification equations.

- **Decoration of the source expression:** The abstract syntax tree of the input definition has to be decorated with typing information, which will be used in subsequent phases of the algorithm. For this purpose we define the set  $\mathbf{Exp}_1$  of intermediate expressions by the grammar in Figure 4.5 and we use  $\epsilon, \epsilon_1, \dots$  to denote such expressions.
  - Copy expressions  $x@$  are attached a region type variable  $\rho$ , which is the type of the region in which the destination DS will reside at runtime.
  - Constructor applications  $C \bar{a}_i^n$  are decorated with another variable  $\rho$ , which is the outermost region type of the DS being created.
  - Function calls  $g \bar{a}_i^n$  that are different from the function being defined (i.e. nonrecursive calls) receive a sequence of RTVs  $\bar{\rho}_j^m$  as decoration. These types correspond to the region variables that will be added as additional arguments in this function call. Since the regions of these functions have already been inferred, the number of needed RTVs is known by the compiler.
  - Recursive function calls  $f \bar{a}_i^n$  are decorated with an identifier  $k \in \mathbb{N}$  and the concrete instantiation of  $f$ ’s type that is used in this recursive call. If the function being inferred has  $p$  recursive calls, we assume that each call has a unique number ranging from 1 to  $p$ .

The generation of unification equations is defined by a set of rules with operate on judgements of this form:

$$\Gamma \vdash_f \epsilon^0 \longrightarrow \epsilon : \alpha \mid E \tag{4.1}$$

where  $\Gamma$  is a pair of functions in  $(\mathbf{Fun} \cup \mathbf{Cons} \rightarrow \mathbf{SafeFunType}) \times (\mathbf{Var} \rightarrow \mathbf{TypeVar})$  that map, respectively, function and constructor symbols to safe type schemes  $\sigma$  and program variables  $x$  to type variables  $\alpha$ . Again, by abuse of notation, we will not make an explicit distinction between these mappings, and both of them will be denoted by  $\Gamma$ . It will be clear from the context which mapping we are referring to, depending on whether we write  $\Gamma(f)$ ,  $\Gamma(C)$  or  $\Gamma(x)$ . The intuitive meaning of (4.1) is that, under the environment  $\Gamma$ , the region-free expression  $\epsilon_0 \in \mathbf{Exp}_0$  is transformed into the intermediate expression  $\epsilon \in \mathbf{Exp}_1$  of type  $\alpha$ . As a side effect, the set  $E$  of constraints between types is generated. Each constraint may be of one of these types:

$$s_1 = s_2$$

$$\begin{array}{c}
\frac{c \in \mathbf{Int}}{\Gamma \vdash_f c \longrightarrow c : \alpha \mid \{\alpha = \mathit{Int}\}} \text{ [LITI}_{\text{Gen}}] \quad \frac{c \in \mathbf{Bool}}{\Gamma \vdash_f c \longrightarrow c : \alpha \mid \{\alpha = \mathit{Bool}\}} \text{ [LITB}_{\text{Gen}}] \\
\\
\frac{}{\Gamma \vdash_f x \longrightarrow x : \alpha \mid \{\Gamma(x) = \alpha\}} \text{ [VAR}_{\text{Gen}}] \\
\\
\frac{\mathit{fresh}(\rho)}{\Gamma \vdash_f x@ \longrightarrow x@_{[\rho]} : \alpha \mid \{\alpha = \mathit{isData}(\Gamma(x), \rho)\}} \text{ [COPY}_{\text{Gen}}] \\
\\
\frac{\mathit{fresh}(\rho) \quad \forall i \in \{1 \dots n\} : \Gamma \vdash_f a_i \longrightarrow a_i : \alpha_i \mid E_i \quad \bar{s}_i^n \rightarrow \rho' \rightarrow s \trianglelefteq \Gamma(C)}{\Gamma \vdash_f C \bar{a}_i^n \longrightarrow C \bar{a}_i^n_{[\rho]} : \alpha \mid \{\bar{s}_i^n \rightarrow \rho' \rightarrow s = \bar{a}_i^n \rightarrow \rho \rightarrow \alpha\} \cup \bigcup_{i=1}^n E_i} \text{ [CONS}_{\text{Gen}}] \\
\\
\frac{\mathit{fresh}(k) \quad \forall i \in \{1 \dots n\} : \Gamma \vdash_f a_i \longrightarrow a_i : \alpha_i \mid E_i}{\Gamma \vdash_f f \bar{a}_i^n \longrightarrow f \bar{a}_i^n_{[k, \bar{a}_i^n \rightarrow \alpha]} : \alpha \mid \{\Gamma(f) \approx_k \bar{a}_i^n \rightarrow \alpha\} \cup \bigcup_{i=1}^n E_i} \text{ [APP-R}_{\text{Gen}}] \\
\\
\frac{g \neq f \quad \forall i \in \{1 \dots n\} : \Gamma \vdash_f a_i \longrightarrow a_i : \alpha_i \mid E_i \quad \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s \trianglelefteq \Gamma(g)}{\Gamma \vdash_f g \bar{a}_i^n \longrightarrow g \bar{a}_i^n_{[\bar{\rho}_j^m]} : \alpha \mid \{\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s = \bar{a}_i^n \rightarrow \bar{\rho}_j^m \rightarrow \alpha\} \cup \bigcup_{i=1}^n E_i} \text{ [APP-NR}_{\text{Gen}}] \\
\\
\frac{\Gamma \vdash_f \epsilon_1^0 \longrightarrow \epsilon_1 : \alpha_1 \mid E_1 \quad \Gamma \setminus x_1 + [x_1 : \alpha_1] \vdash_f \epsilon_2^0 \longrightarrow \epsilon_2 : \alpha_2 \mid E_2}{\Gamma \vdash_f \mathbf{let} \ x_1 = \epsilon_1^0 \ \mathbf{in} \ \epsilon_2^0 \longrightarrow \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2 : \alpha \mid E_1 \cup E_2 \cup \{\alpha = \alpha_2\}} \text{ [LET}_{\text{Gen}}] \\
\\
\frac{\Gamma(x) = \alpha_x \quad \forall i \in \{1 \dots n\}. \begin{cases} \mathit{fresh}(\bar{\alpha}_{ij}^{n_i}) \wedge \bar{s}_{ij}^{n_i} \rightarrow \rho \rightarrow s_i \trianglelefteq \Gamma(C_i) \\ \Gamma \setminus \bar{x}_{ij}^{n_i} + [\bar{x}_{ij} : \alpha_{ij}^{n_i}] \vdash_f \epsilon_i^0 \longrightarrow \epsilon_i : \alpha_i \mid E_i' \\ E_i = E_i' \cup \{\bar{s}_{ij}^{n_i} \rightarrow s_i = \bar{\alpha}_{ij}^{n_i} \rightarrow \alpha_x\} \cup \{\alpha_i = \alpha\} \end{cases}}{\Gamma \vdash_f \mathbf{case}(!) \ x \ \mathbf{of} \ C_i \ \bar{x}_{ij}^{n_i} \rightarrow \epsilon_i^0 \longrightarrow \mathbf{case}(!) \ x \ \mathbf{of} \ C_i \ \bar{x}_{ij}^{n_i} \rightarrow \epsilon_i : \alpha \mid \bigcup_{i=1}^n E_i} \text{ [CASE}_{\text{Gen}}]
\end{array}$$

Figure 4.6: Rules for generating equations.

It specifies that the types at both sides of the = must be equal after unification.

$$s_1 = \mathit{isData}(s_2, \rho)$$

This means that the types of  $s_1$  and  $s_2$  may only differ in the outermost region, which must be  $\rho$  in the type  $s_1$ .

$$s_1 \approx_k s_2$$

It specifies that  $s_2$  must be a region instance of  $s_1$ , but the HM types (i.e. those without considering regions) must be equal. This allows us to have polymorphic recursion over regions. We annotate each constraint of this form with a number  $k$  which is the identifier of the recursive call from which this constraint is generated (see below).

The rules that define the equation generation can be found in Figure 4.6. Although not stated explicitly, we assume that  $\alpha$  denotes a *fresh* type variable in every rule. Moreover, we assume that each occurrence of  $\trianglelefteq$  in the rules means that the type on the left-hand side is a *fresh* instance of the type scheme on the right-hand side. Most of the rules are the usual ones in HM inference, with some minor differences. Rule [COPY<sub>Gen</sub>] generates an *isData* constraint, in order to specify that the outermost region type of the DS being copied may differ from  $\rho$ , which is the outermost region type of the result. We make distinction between recursive and non-recursive function applications. In the latter case, the function

being called has been already inferred, so we have to generate as many fresh RTVs as in the function's signature. In recursive function applications we generate an  $\approx$  equation and a fresh identifier.

**Example 4.4.** Given the following function for appending two lists:

$$\begin{aligned} \mathit{append} \ xs \ ys &= \mathbf{case} \ xs \ \mathbf{of} \\ & \quad [] \rightarrow ys \ @ \\ & \quad (x : xx) \rightarrow \mathbf{let} \ x_1 = \mathit{append} \ xx \ ys \ \mathbf{in} \ (x : x_1) \end{aligned}$$

We start with an environment  $\Gamma \supseteq [\mathit{append} : \alpha_0, xs : \alpha_1, ys : \alpha_2]$ . The rules of Figure 4.6 generate the following constraints:

$$\begin{array}{lll} [\alpha_3]@_{\rho_1} = \alpha_1 & & \{[] \text{ pattern}\} \\ \alpha_6 \rightarrow [\alpha_6]@_{\rho_2} \rightarrow [\alpha_6]@_{\rho_2} = \alpha_4 \rightarrow \alpha_5 \rightarrow \alpha_1 & & \{(x : xx) \text{ pattern}\} \\ \alpha_7 = \mathit{isData}(\alpha_2, \rho_3) & & \{ys \ @\} \\ \alpha_0 \approx_1 \alpha_5 \rightarrow \alpha_2 \rightarrow \alpha_8 & & \{\text{call to } \mathit{append}\} \\ \alpha_9 \rightarrow [\alpha_9]@_{\rho_4} \rightarrow \rho_4 \rightarrow [\alpha_9]@_{\rho_4} = \alpha_4 \rightarrow \alpha_8 \rightarrow \rho_5 \rightarrow \alpha_{10} & & \{(x : x_1)\} \\ \alpha_{11} = \alpha_7 & & \{\mathbf{case} \ \text{branch}\} \\ \alpha_{11} = \alpha_{10} & & \{\mathbf{case} \ \text{branch}\} \\ \alpha_0 = \alpha_1 \rightarrow \alpha_2 \rightarrow \alpha_{11} & & \{\text{type of } \mathit{append}\} \end{array}$$

We obtain the following intermediate expression with decorations:

$$\begin{aligned} \epsilon_{\mathit{append}} &= \mathbf{case} \ xs \ \mathbf{of} \\ & \quad [] \rightarrow ys \ @_{[\rho_3]} \\ & \quad (x : xx) \rightarrow \mathbf{let} \ x_1 = \mathit{append} \ xx \ ys_{[1, \alpha_5 \rightarrow \alpha_2 \rightarrow \alpha_8]} \ \mathbf{in} \ (x : x_1)_{[\rho_5]} \end{aligned}$$

The copy expression and the application of the constructor are annotated with the region type variables corresponding to their respective results, whereas the call to  $\mathit{append}$  is annotated with its concrete type.  $\square$

Once we have generated a set  $E$  of constraints, we can solve them by unification. The technique we use is a slight variation of original method defined by Robinson [101] with some extensions for handling  $\mathit{isData}$  constraints and  $\approx$ -equations, which are treated in this context like latent unifications. These extensions are straightforward to define and implement, and they are shown in Appendix A.

The unification algorithm returns a substitution  $\theta$  mapping type variables  $\alpha$  (resp. RTVs  $\rho$ ) to types or type signatures (resp. to RTVs). It also yields, for each equation of the form  $s_1 \approx_k s_2$ , a substitution  $\varphi_k$  between region types which maps the region type variables of the left-hand side to the ones in the right-hand side. If the input function definition has  $p$  recursive calls (and hence it involves the generation of  $p$  equations of this kind) then the unification method yields a list of  $p$  substitutions  $\overline{\varphi}_i^p$ . In our implementation, both  $\theta$  and the  $\overline{\varphi}_i^p$  are computed simultaneously, but the computation of the  $\overline{\varphi}_i^p$  can also be deferred after the unification.

Although we do not explain in this chapter how to obtain a solution via unification, we must specify the meaning of a solution in order to prove the correctness of the algorithm. In the following definition we use the notation  $s_1 \equiv s_2$  to denote syntactic equality of types. We also assume that the application of substitutions  $\theta$  and  $\varphi_i$  to type expressions is defined in the usual way.

**Definition 4.5.** A pair  $\langle \theta, \overline{\varphi_i^p} \rangle$  is a solution to a set of constraints  $E$  if and only if the following conditions hold:

1. For each  $s_1 = s_2 \in E$ , then  $\theta(s_1) \equiv \theta(s_2)$ . The same applies to equations of the form  $sf_1 = sf_2$ .
2. For each  $s = isData(s', \rho) \in E$  then  $\theta(s) \equiv T \overline{s_i^n} @ \overline{\rho_j^m} (\theta(\rho))$  and  $\theta(s') \equiv T \overline{s_i^n} @ \overline{\rho_j^m} \rho'$  for some  $\overline{s_i^n}, \overline{\rho_j^m}$  and  $\rho'$ .
3. For each  $s_1 \approx_k s_2$  then  $\varphi_k(\theta(s_1)) \equiv \theta(s_2)$ . The same applies to equations of the form  $sf_1 \approx_k sf_2$ .

**Example 4.6.** The unification method yields the following substitution for the Example 4.4 above:

$$\theta = \left[ \begin{array}{lll} \alpha_1 \mapsto [\alpha_3]@_{\rho_1} & \alpha_2 \mapsto [\alpha_3]@_{\rho_6} & \alpha_4 \mapsto \alpha_3 \\ \alpha_5 \mapsto [\alpha_3]@_{\rho_1} & \alpha_6 \mapsto \alpha_3 & \alpha_7 \mapsto [\alpha_3]@_{\rho_4} \\ \alpha_8 \mapsto [\alpha_3]@_{\rho_4} & \alpha_9 \mapsto \alpha_3 & \alpha_{10} \mapsto [\alpha_3]@_{\rho_4} \\ & \alpha_{11} \mapsto [\alpha_3]@_{\rho_4} & \\ \rho_2 \mapsto \rho_1 & \rho_3 \mapsto \rho_4 & \rho_5 \mapsto \rho_4 \\ & \alpha_0 \mapsto [\alpha_3]@_{\rho_1} \rightarrow [\alpha_3]@_{\rho_8} \rightarrow [\alpha_3]@_{\rho_6} & \end{array} \right]$$

We compare the type of the function with the type of the recursive application:

$$\begin{array}{ll} \text{Function: } \textit{append} & :: [\alpha_3]@_{\rho_1} \rightarrow [\alpha_3]@_{\rho_6} \rightarrow [\alpha_3]@_{\rho_4} \\ \text{Recursive application: } \textit{append} & :: [\alpha_3]@_{\rho_1} \rightarrow [\alpha_3]@_{\rho_6} \rightarrow [\alpha_3]@_{\rho_4} \end{array}$$

Both types are equal, and hence there is no polymorphic recursion in this example. Therefore, we get the identity region mapping for this recursive call:

$$\varphi_1 = [\rho_1 \mapsto \rho_1, \rho_6 \mapsto \rho_6, \rho_4 \mapsto \rho_4]$$

By applying  $\theta$  to the intermediate expression we obtain:

$$\begin{aligned} \theta(\epsilon_{\textit{append}}) &= \text{case } xs \text{ of} \\ & \quad [] \rightarrow ys @_{[\rho_4]} \\ & \quad (x : xx) \rightarrow \text{let } x_1 = \textit{append } xx \text{ } ys_{[1, [\alpha_3]@_{\rho_1} \rightarrow [\alpha_3]@_{\rho_6} \rightarrow [\alpha_3]@_{\rho_4}]} \text{ in } (x : x_1)_{[\rho_4]} \end{aligned}$$

□

#### 4.4.2 Kernel of the algorithm

Once we have obtained a decorated expression and a solution  $\langle \theta, \overline{\varphi_i^p} \rangle$  for the type equations, we apply the substitution  $\theta$  to the type decorations of the expression, obtaining another decorated expression  $\epsilon \in \mathbf{Exp}_1$  as a result. We also apply this substitution to the fresh type variable  $\alpha_0$  assigned to the function being inferred in order to obtain its type  $\theta(\alpha_0) \equiv \overline{s_i^n} \rightarrow s$  with RTVs. Let us denote by  $R_{in}$  the region type variables occurring in the type of the input arguments (i.e. the  $\overline{s_i^n}$ ), and by  $R_{out}$  those occurring in the result of the function (i.e. the  $s$ ). Our next step is to classify these RTVs, along with those that have appeared in the decorations of  $\epsilon$ . Some of these RTVs must be added to the type of the function as extra parameters; some others will be unified with the region type variable  $\rho_{self}$ . Once we have done this classification, the generation of region variables in order to produce the final *Core-Safe* expression is straightforward.

$$\begin{aligned}
& \text{inferRegions } (\bar{s}_i^n \rightarrow s) \in \bar{\varphi}_j^p = (R_{arg}, R_{self}, R_{expl}) \\
& \quad \text{where } R_{out} = \text{regions}(s) \\
& \quad \quad R_{in} = \text{regions}(\bar{s}_i^n) \\
& \quad \quad R_0 = \text{regions}(\epsilon) \\
& \quad \quad (R_{arg}, R_{expl}) = \text{computeRargFP } R_{in} R_{out} \oslash R_0 \bar{\varphi}_j^p \\
& \quad \quad R_{self} = R_{expl} \setminus (R_{in} \cup R_{out}) \\
\\
& \text{computeRargFP } R_{in} R_{out} R_{arg} R_0 \bar{\varphi}_j^p \\
& \quad | R_{arg} == R'_{arg} = (R'_{arg}, R'_{expl}) \\
& \quad | \text{otherwise} = \text{computeRargFP } R_{in} R_{out} R'_{arg} R_0 \bar{\varphi}_j^p \\
& \quad \text{where } R'_{expl} = R_0 \cup \bigcup_{j=1}^p \{ \varphi_j^+(\rho) \mid \rho \in R_{arg} \} \\
& \quad \quad R'_{arg} = R'_{expl} \cap (R_{in} \cup R_{out})
\end{aligned}$$

Figure 4.7: Second phase of the region inference algorithm.

The main phase of the region inference algorithm computes three sets of RTVs:

$R_{arg}$  It contains the types of the region variables that will eventually be added as extra parameters of the function definition. In other words, if  $\bar{s}_i^n \rightarrow s$  is the type we have obtained for the function in the previous phase, and  $R_{arg} = \{\bar{\rho}_j^m\}$ , then the function's signature will be extended as  $\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s$ .

$R_{self}$  It contains the RTVs that will be unified with the  $\rho_{self}$  region. This implies that the corresponding region variable in the *Core-Safe* program will be *self*.

$R_{expl}$  It is the set of *explicit* RTVs. A region type variable  $\rho$  is considered to be explicit if there exists a region variable  $r$  in the resulting *Core-Safe* program with type  $\rho$ .

These definitions are far from being constructive, since they all refer to the resulting region-annotated program, which is exactly what we aim for. However, they provide enough intuition in order to characterize some properties of these sets. We have to find  $R_{arg}$ ,  $R_{self}$  and  $R_{expl}$  subject to the following restrictions:

1.  $R_{expl} \subseteq R_{self} \cup R_{arg}$
2.  $R_{self} \cap R_{arg} = \emptyset$
3.  $R_{self} \cap (R_{in} \cup R_{out}) = \emptyset$
4.  $R_{expl} = \text{regions}(\epsilon) \cup \bigcup_{j=1}^p \varphi_j^+(R_{arg})$ , where  $\text{regions}(\epsilon)$  is the set of RTVs appearing in the annotations of those subexpressions of  $\epsilon$  of the form  $C \bar{a}_{i[\rho]}$  and  $g \bar{a}_{i[\bar{\rho}_j]}$ , and  $\varphi_j^+(\rho)$  is defined for every  $j \in \{1..p\}$  as follows:

$$\varphi_j^+(\rho) = \begin{cases} \varphi_j(\rho) & \text{if } \rho \in \text{dom } \varphi_j \\ \rho & \text{if } \rho \notin \text{dom } \varphi_j \end{cases}$$

We use the notation  $\varphi_j^+(R_{arg})$  to denote the set  $\{\varphi_j(\rho) \mid \rho \in R_{arg}\}$ .

The first one expresses that the region variables occurring in the final *Core-Safe* program (and hence belonging to  $R_{expl}$ ) must be variables in scope. The types of region variables in scope belong either to  $R_{arg}$  or  $R_{self}$ , depending on whether they are additional parameters of the function definition, or the identifier *self*. The second and third restrictions are forced by the fact that  $\rho_{self}$  cannot appear in the

resulting type signature of the function, as the rule [FUNB] in Figure 3.9 demands. The last restriction specifies that the RTVs appearing in the annotations of  $\epsilon$  are considered explicit. Moreover, if  $f$  is the function being inferred, this restriction also requires that the type of  $f$  extended with the region arguments in  $R_{arg}$  can produce type instances for typing all the recursive applications of  $f$ , each one extended with as many region arguments as the number of elements in  $R_{arg}$ .

In Section 4.5 we show that any sets  $R_{expl}$ ,  $R_{arg}$  and  $R_{self}$  satisfying these restrictions lead to a region-annotated *Core-Safe* expression which is well-typed according to  $\vdash_{Reg}$  rules. Notice that, provided the restriction (4) holds, an algorithm choosing any  $R_{arg} \supseteq R_{expl}$  and  $R_{self} = \emptyset$  would be correct with respect to this specification, but this solution would be very poor as, on the one hand, no construction would ever be done in the *self* region and, on the other hand, there might be region arguments never used. We look for an optimal solution in two senses: we want  $R_{arg}$  to be as small as possible, so that only those regions where data are built are given as arguments, but we also want  $R_{self}$  to be as big as possible, so that the maximum amount of memory is deallocated at function termination.

Our algorithm initially computes  $R_{arg} = R_0 \cap (R_{in} \cup R_{out})$ , by using the set  $R_0$  of initial explicit RTVs, that is,  $R_0 \stackrel{\text{def}}{=} \text{regions}(\epsilon)$ . Then, it starts a fixpoint algorithm *computeRargFP* (Figure 4.7) trying to get the type of  $f$ 's recursive applications as instances of the type of  $f$  extended with the current set  $R_{arg}$  of arguments. It may happen that the set  $R_{expl}$  may grow when considering different applications (see the Example 4.17). Adding more explicit RTVs to one recursive application will influence the type of the rest. As  $R_{arg}$  depends on  $R_{expl}$ , it may also grow. So, a fixpoint is used in order to obtain the final  $R_{arg}$  and  $R_{expl}$  from the initial ones. Due to our solution above,  $R_{arg}$  cannot grow greater than  $R_{in} \cup R_{out}$ , so termination of the fixpoint is guaranteed. We have the following invariant in the algorithm:

$$R_{arg} = R_{expl} \cap (R_{in} \cup R_{out}) \quad (4.2)$$

Once obtained the final  $R_{arg}$  and  $R_{expl}$ , the set  $R_{self}$  is computed as follows:

$$R_{self} = R_{expl} \setminus (R_{in} \cup R_{out}) \quad (4.3)$$

We show in Section 4.5 that these choices maximise the data allocated to the *self* region, which in turn maximises the amount of memory reclaimed at runtime when the corresponding function call finishes. With respect to the remaining DSs not being inferred to live in *self*, they will be allocated to the regions which are parameters to the function being called. It is the *caller* function's responsibility to determine where to put these DSs by passing the suitable arguments. Since the caller function is also inferred by the algorithm, the parameter assignment is done in such a way that the data allocated in the caller's *self* region is also maximized. From a global point of view, every cell not being created in the current topmost region (i.e. the region bound to the *self* identifier) will be created in the highest possible region and hence, will be deallocated at the earliest time allowed by the type system.

**Example 4.7.** Continuing with the Example 4.6 above, we start iterating with the following sets:

$$R_{in} = \{\rho_1, \rho_6\} \quad R_{out} = \{\rho_4\} \quad R_{arg} = \emptyset \quad R_0 = \{\rho_4\}$$

After the first iteration, we get  $R_{expl} = \{\rho_4\}$  and  $R_{arg} = \{\rho_4\}$ . A further iteration leads to the same solution, which is hence the fixpoint. Finally, we compute  $R_{self} = \emptyset$  by using (4.3). This implies that nothing is going to be built in the *self* region during the evaluation of this function. From the RTVs in

$R_{arg}$  we can extend the type signature of *append* as follows:

$$append :: [\alpha_3]@_{\rho_1} \rightarrow [\alpha_3]@_{\rho_6} \rightarrow \rho_4 \rightarrow [\alpha_3]@_{\rho_4}$$

□

**Example 4.8.** Given the following definition:

$$f \ xs \ ys = \mathbf{let} \ zs = (1 : xs) \ \mathbf{in} \ f \ ys \ zs$$

After the HM inference phase, we obtain the preliminary type  $[Int]@_{\rho_1} \rightarrow [Int]@_{\rho_2}$  for  $f$ , the region substitution  $\varphi_1 = [\rho_1 \mapsto \rho_2, \rho_2 \mapsto \rho_1]$  corresponding to the first recursive call, and the following intermediate definition:

$$f \ xs \ ys = \mathbf{let} \ zs = (1 : xs)_{[\rho_1]} \ \mathbf{in} \ f \ ys \ zs_{[1, [Int]@_{\rho_2} \rightarrow [Int]@_{\rho_1}]}$$

The initial call to *computeRargFP* is done with the following sets:

$$R_{in} = \{\rho_1\} \quad R_{out} = \{\rho_2\} \quad R_{arg} = \emptyset \quad R_0 = \{\rho_1\}$$

After the first iteration,  $R_{expl}$  and  $R_{arg}$  are both equal to  $\{\rho_1\}$ . However, since  $\varphi_1(\rho_1) = \rho_2$ , the RTV  $\rho_2$  is added to both sets in the second iteration. By iterating a third time we obtain the same result, so we have the following fixed point:

$$R_{arg} = \{\rho_1, \rho_2\} \quad R_{expl} = \{\rho_1, \rho_2\}$$

□

### 4.4.3 Annotating function definitions with region variables

Once we know the exact number of RTVs in  $R_{arg}$ , we create a fresh region parameter for each RTV in this set. Let us assume that  $R_{arg} = \{\bar{\rho}_j^m\}$  and that  $\bar{r}_j^m$  are the newly created region parameters. We define a mapping  $\mathcal{R} : R_{arg} \cup R_{self} \rightarrow \mathbf{RegVar}$  as follows:

$$\mathcal{R} \stackrel{\text{def}}{=} [\bar{\rho}_j \mapsto \bar{r}_j^m] \uplus [\rho \mapsto self \mid \rho \in R_{self}]$$

The last step is to traverse the expression  $\epsilon$  in order to transform its decorations into region variables, so that we get a *Core-Safe* expression as a result. This transformation is done by the *annotateExp* function, defined in Figure 4.8. Although not explicitly shown, we assume two extra parameters in this function that are propagated downwards to each recursive call: the  $R_{arg}$  and  $\bar{\varphi}_j^p$  computed in the previous phases. Most cases in the definition of *annotateExp* are straightforward, since we only have to apply the  $\mathcal{R}$  mapping to the decorations in  $\epsilon$  in order to get the actual region annotations in the program. In the case of expressions of the form  $f \ \bar{a}_i^n_{[k, \bar{s}_i^n \rightarrow s]}$ , which correspond to recursive calls, the types of the region variables to be inserted may be distinct from the variables in  $R_{arg}$ , since we allow polymorphic recursion. Hence we proceed in two phases: firstly we obtain the concrete region instance corresponding to this call site (this is given by the  $\varphi_k^+$ ), and then, we apply the  $\mathcal{R}$  mapping which translates each RTV of this instance into a region variable.

$$\begin{array}{lll}
\text{annotateExp } \mathcal{R} & (c) & = c \\
\text{annotateExp } \mathcal{R} & (x) & = x \\
\text{annotateExp } \mathcal{R} & (x@_{[\rho]}) & = x @ \mathcal{R}(\rho) \\
\text{annotateExp } \mathcal{R} & (C \bar{a}_i^n_{[\rho]}) & = C \bar{a}_i^n @ \mathcal{R}(\rho) \\
\text{annotateExp } \mathcal{R} & (g \bar{a}_i^n_{[\bar{\rho}_j^m]}) & = g \bar{a}_i^n @ \mathcal{R}(\bar{\rho}_j^m) \\
\text{annotateExp } \mathcal{R} & (f \bar{a}_i^n_{[k, \bar{s}_i^n \rightarrow s]}) & = \text{annotateExp } \mathcal{R} (f \bar{a}_i^n_{[\varphi_k^+(\bar{\rho}_j^m)]}) \\
& & \mathbf{where} \{ \bar{\rho}_j^m \} = R_{arg} \\
\text{annotateExp } \mathcal{R} & (\mathbf{let} x_1 = \epsilon_1 \mathbf{in} \epsilon_2) & = \mathbf{let} x_1 = e_1 \mathbf{in} e_2 \\
& & \mathbf{where} e_1 = \text{annotateExp } \mathcal{R} \epsilon_1 \\
& & \quad e_2 = \text{annotateExp } \mathcal{R} \epsilon_2 \\
\text{annotateExp } \mathcal{R} & (\mathbf{case}(!) x \mathbf{of} \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow \epsilon_i^n}) & = \mathbf{case}(!) x \mathbf{of} \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow \epsilon_i^n} \\
& & \mathbf{where} \forall i : e_i = \text{annotateExp } \mathcal{R} \epsilon_i
\end{array}$$

Figure 4.8: Definition of *annotateExp*.

**Example 4.9.** From the results of the Example 4.7 we build the mapping  $\mathcal{R} = [\rho_4 \mapsto r]$  and obtain the following *Core-Safe* definition for *append*:

$$\begin{aligned}
\text{append } xs \ ys @ r &= \mathbf{case} \ xs \ \mathbf{of} \\
& [] \rightarrow ys @ r \\
& (x : xx) \rightarrow \mathbf{let} \ x_1 = \text{append } xx \ ys \ \mathbf{in} \ (x : x_1) @ r
\end{aligned}$$

□

**Example 4.10.** Given the definition of *f* in Example 4.8, we apply *annotateExp* with the mapping  $\mathcal{R} = [\rho_1 \mapsto r_1, \rho_2 \mapsto r_2]$ , with  $R_{arg} = \{\rho_1, \rho_2\}$  and  $\varphi_1 = [\rho_1 \mapsto \rho_2, \rho_2 \mapsto \rho_1]$ :

$$\begin{aligned}
& \text{annotateExp } \mathcal{R} \left( \mathbf{let} \ zs = (1 : xs)_{[\rho_1]} \ \mathbf{in} \ f \ ys \ zs_{[1, [Int]@_{\rho_2} \rightarrow [Int]@_{\rho_1}]} \right) \\
&= \mathbf{let} \ zs = \text{annotateExp } \mathcal{R} \left( (1 : xs)_{[\rho_1]} \right) \ \mathbf{in} \ \text{annotateExp } \mathcal{R} \left( f \ ys \ zs_{[1, [Int]@_{\rho_2} \rightarrow [Int]@_{\rho_1}]} \right) \\
&= \mathbf{let} \ zs = (1 : xs) @ r_1 \ \mathbf{in} \ \text{annotateExp } \mathcal{R} \left( f \ ys \ zs_{[\rho_2, \rho_1]} \right) \\
&= \mathbf{let} \ zs = (1 : xs) @ r_1 \ \mathbf{in} \ f \ ys \ zs @ r_2 \ r_1
\end{aligned}$$

We obtain the following region-annotated definition:

$$f \ xs \ ys @ r_1 \ r_2 = \mathbf{let} \ zs = (1 : xs) @ r_1 \ \mathbf{in} \ f \ ys \ zs @ r_2 \ r_1$$

□

## 4.5 Correctness and optimality

The aim of this section is to prove that the final *Core-Safe* program resulting from the algorithm is typable with respect to the  $\vdash_{Reg}$  rules in Figure 3.18. This fact can be established in three steps:

1. The HM inference algorithm described in Section 4.4.1 returns a *well-formed* intermediate expression  $\epsilon \in \mathbf{Exp}_1$ , provided the unification algorithm returns a correct solution to the generated constraints.

$$\begin{array}{c}
\frac{}{\Gamma \Vdash_{\bar{\varphi}} c : B} \text{[LIT}_{\text{HM}}] \quad \frac{}{\Gamma + [x : s] \Vdash_{\bar{\varphi}} x : s} \text{[VAR}_{\text{HM}}] \\
\\
\frac{\Gamma(x) = T \bar{s}_i^n @_{\bar{\rho}_j^m} \rho'}{\Gamma \Vdash_{\bar{\varphi}} x @_{[\rho]} : T \bar{s}_i^n @_{\bar{\rho}_j^m} \rho} \text{[COPY}_{\text{HM}}] \\
\\
\frac{\bar{s}_i^n \rightarrow \rho \rightarrow s \leq \Gamma(C) \quad \forall i \in \{1 \dots n\}. \Gamma \Vdash_{\bar{\varphi}} a_i : s_i}{\Gamma \Vdash_{\bar{\varphi}} C \bar{a}_i^n : s} \text{[CONS}_{\text{HM}}] \\
\\
\frac{\varphi_k(\Gamma(f)) = \bar{s}_i^n \rightarrow s \quad \forall i \in \{1 \dots n\}. \Gamma \Vdash_{\bar{\varphi}} a_i : s_i}{\Gamma \Vdash_{\bar{\varphi}} f \bar{a}_i^n_{[k, \bar{s}_i^n \rightarrow s]} : s} \text{[APP-R}_{\text{HM}}] \\
\\
\frac{\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s \leq \sigma \quad \forall i \in \{1 \dots n\}. \Gamma \Vdash_{\bar{\varphi}} a_i : s_i}{\Gamma + [g : \sigma] \Vdash_{\bar{\varphi}} g \bar{a}_i^n_{[\bar{\rho}_j^m]} : s} \text{[APP-NR}_{\text{HM}}] \\
\\
\frac{\Gamma \Vdash_{\bar{\varphi}} \epsilon_1 : s_1 \quad \Gamma \setminus x_1 + [x_1 : s_1] \Vdash_{\bar{\varphi}} \epsilon_2 : s}{\Gamma \Vdash_{\bar{\varphi}} \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2 : s} \text{[LET}_{\text{HM}}] \\
\\
\frac{\Gamma(x) = s_x \quad \forall i \in \{1..n\} : \left\{ \begin{array}{l} \bar{s}_{ij}^{n_i} \rightarrow \rho \rightarrow s_x \leq \Gamma(C_i) \\ \Gamma \setminus \bar{x}_{ij}^{n_i} + [\bar{x}_{ij} : \bar{s}_{ij}^{n_i}] \Vdash_{\bar{\varphi}} \epsilon_i : s \end{array} \right.}{\Gamma \Vdash_{\bar{\varphi}} \mathbf{case}(!) \ x \ \mathbf{of} \ C_i \ \bar{x}_{ij}^{n_i} \rightarrow \epsilon_i^{n_i} : s} \text{[CASE(!)}_{\text{HM}}]
\end{array}$$

Figure 4.9: Rules for  $\Vdash$ .

2. Given a *well-formed* intermediate expression  $\epsilon \in \mathbf{Exp}_1$  and three sets  $(R_{arg}, R_{self}, R_{expl})$  such that the four conditions (1), (2), (3) and (4) of Section 4.4.2 hold, the result of *annotateExp* (as described in Section 4.4.3) is a well-typed *Core-Safe* expression.
3. The solution of  $(R_{arg}, R_{self}, R_{expl})$  returned by the function *inferRegions* satisfies the above mentioned four conditions.

The link between these steps is not only the intermediate expression  $\epsilon$ , but also the fact that it is well-formed. We define this notion by means of an intermediate type system, defined by the rules in Figure 4.9. This type system enforces that the annotations in  $\epsilon$  do not contradict its type. Most rules are similar to those in Section 3.18 and hence they will not be explained here. The only detail worth noting is that there are two different  $\Vdash_{\bar{\varphi}}$  rules for dealing with function applications, [APP-R<sub>HM</sub>] and [APP-NR<sub>HM</sub>], since recursive and nonrecursive function applications are decorated in different ways. In the case of recursive function applications, the functional type in the decoration is a region instance of the type bound to  $f$  within the environment  $\Gamma$ . The concrete instantiation is given by the corresponding  $\varphi_k$  substitution. The list of region substitutions  $\bar{\varphi}_i^p$  (abbreviated in the rules as  $\bar{\varphi}$ ) is propagated through the whole derivation.

Given these considerations, we are ready to prove the first step. The proof of the following Lemma is similar to that of Hindley-Milner inference. The main difference is that the  $\Vdash_{\bar{\varphi}}$  typing rules involve types with regions.

**Lemma 4.11.** *For every  $\Gamma, \epsilon_0, \epsilon, \alpha, f$  and  $E$  of their respective types, if  $\Gamma \vdash_f \epsilon_0 \longrightarrow \epsilon : \alpha \mid E$  and  $\langle \theta, \bar{\varphi}_i^p \rangle$  is a*

solution to  $E$ , we can derive the following judgement:

$$\theta(\Gamma) \Vdash_{\overline{\varphi}_i^p} \theta(\epsilon) : \theta(\alpha)$$

where  $\theta(\Gamma)$  denotes the application of  $\theta$  to every type in the range of  $\Gamma$ .

*Proof.* By induction on the structure of  $\epsilon_0$ . We distinguish cases:

- **Case**  $\epsilon_0 \equiv c$

We get trivially  $\theta(\Gamma) \Vdash_{\overline{\varphi}} c : \text{Int}$  or  $\theta(\Gamma) \Vdash_{\overline{\varphi}} c : \text{Bool}$  depending on whether  $c \in \mathbf{Int}$  or  $c \in \mathbf{Bool}$ .

- **Case**  $\epsilon_0 \equiv x$

We have to prove  $\theta(\Gamma) \Vdash_{\overline{\varphi}} x : s$  where  $s = \theta(\alpha)$ . Since  $\theta$  is a solution for the equation  $\Gamma(x) = \alpha$  we get  $s = \theta(\alpha) = \theta(\Gamma(x))$ . In order to apply the rule [VAR<sub>HM</sub>] we have to show that  $x \in \text{dom}(\theta(\Gamma))$ , but this follows trivially from the fact that  $x \in \text{dom}(\Gamma)$ .

- **Case**  $\epsilon_0 \equiv x@$

The resulting decorated expression is  $\epsilon \equiv x@_{[\rho]}$ , where  $\rho$  is a fresh region type variable. Since  $\theta$  is the solution of the generated equation, we get  $\theta(\alpha) \equiv T \overline{s}_i^n @ \overline{\rho}_j^m \theta(\rho)$ . Besides this, since  $x \in \text{dom}(\Gamma)$ , we get  $x \in \text{dom}(\theta(\Gamma))$  and hence  $\theta(\Gamma(x)) \equiv T \overline{s}_i^n @ \overline{\rho}_j^m \rho'$  which is exactly the premise of the rule [COPY<sub>HM</sub>]. We can apply this rule in order to obtain  $\theta(\Gamma) \Vdash_{\overline{\varphi}} x@_{[\theta(\rho)]} : \theta(\alpha)$ .

- **Case**  $\epsilon_0 \equiv C \overline{a}_i^n$

In order to apply the rule [CONS<sub>HM</sub>] so as to obtain  $\theta(\Gamma) \Vdash_{\overline{\varphi}} C \overline{a}_i^n_{[\theta(\rho)]} : \theta(\alpha)$  we need to prove its premises:  $\overline{\theta(\alpha_i)}^n \rightarrow \theta(\rho) \rightarrow \theta(\alpha) \leq \theta(\Gamma(C))$  and  $\forall i \in \{1 \dots n\} : \theta(\Gamma) \Vdash a_i : \theta(\alpha_i)$ . The former follows from the definition of  $\theta$  being a solution, since  $\theta(sf_0) = \overline{\theta(\alpha_i)}^n \rightarrow \theta(\rho) \rightarrow \theta(\alpha)$  and  $\theta(sf_0) \leq \Gamma(C)$ . In fact, if  $sf_0$  is an instance of  $\Gamma(C)$ , so it is  $\theta(sf_0)$ . Moreover,  $\Gamma(C)$  does not contain any free variables, so  $\theta(\Gamma(C)) = \Gamma(C)$ . The latter premise is a direct consequence of applying the induction hypothesis to each  $a_i$ .

- **Case**  $\epsilon_0 \equiv f \overline{a}_i^n$

We have to find a  $\Vdash_{\overline{\varphi}}$  typing for the decorated expression  $f \overline{a}_i^n_{[k, \theta(\overline{\alpha}_i^n \rightarrow \alpha)]}$ . Firstly, we can apply the induction hypothesis to each argument in order to get, for all  $i \in \{1 \dots n\}$ ,  $\theta(\Gamma) \Vdash_{\overline{\varphi}} a_i : \theta(\alpha_i)$ . Moreover, and because of  $\langle \theta, \overline{\varphi}_i^p \rangle$  being a solution to  $E$ , we get:

$$\varphi_k(\theta(\Gamma(f))) = \theta(\overline{\alpha}_i^n \rightarrow \alpha)$$

which is the remaining premise that we need for applying the rule [APP-R<sub>HM</sub>].

- **Case**  $\epsilon_0 \equiv g \overline{a}_i^n$ , with  $g \neq f$

The proof is similar to that of case  $\epsilon_0 \equiv C \overline{a}_i^n$ .

- **Case**  $\epsilon_0 \equiv \mathbf{let} \ x_1 = \epsilon_{0,1} \ \mathbf{in} \ \epsilon_{0,2}$

Since  $\theta$  is a solution of  $E$ , it is also a solution of its components, namely  $E_1$  and  $E_2$ . This allows us to apply the induction hypothesis on each subexpression with  $\theta$  so that we get:

$$\theta(\Gamma) \Vdash_f \theta(\epsilon_1) : \theta(\alpha_1) \quad \theta(\Gamma) \setminus x_1 + [x_1 : \theta(\alpha_1)] \Vdash_{\overline{\varphi}} \theta(\epsilon_2) : \theta(\alpha_2)$$

Now we can apply the rule [LET<sub>HM</sub>] by using these two judgements and the fact that  $\theta(\alpha_2) = \theta(\alpha)$ , since  $\theta$  is a solution of  $\{\alpha_2 = \alpha\}$ .

- **Case**  $\epsilon_0 \equiv \mathbf{case}(!) x \text{ of } \overline{C_i} \overline{x_{ij}^{n_i}} \rightarrow \epsilon_{0,i}^n$

Again, the general solution  $\theta$  is also a solution for each particular  $E_i$ , so we apply the induction hypothesis to each  $\epsilon_{0,i}$  in order to get  $\theta(\Gamma) \setminus \overline{x_{ij}^{n_i}} + [\overline{x_{ij} : \theta(\alpha_{ij})}^{n_i}] \Vdash_{\overline{\varphi}} \theta(\epsilon_i) : \theta(\alpha_i)$ , which is equivalent to  $\theta(\Gamma) \setminus \overline{x_{ij}^{n_i}} + [\overline{x_{ij} : \theta(\alpha_{ij})}^{n_i}] \Vdash_{\overline{\varphi}} \theta(\epsilon_i) : \theta(\alpha)$  because of  $\theta$  being a solution to  $\{\alpha_i = \alpha\}$ . Now we have to prove  $\overline{\theta(\alpha_{ij})}^{n_i} \rightarrow \rho \rightarrow \theta(\alpha_x) \leq \theta(\Gamma(C_i))$  for some  $\rho$  and for each  $i \in \{1 \dots n\}$ . One of the premises of the rule [CASE<sub>Gen</sub>] tell us that there exist some  $\overline{s_{ij}^n}, \rho'$  and  $s_i$  such that  $\overline{s_{ij}^n} \rightarrow \rho' \rightarrow s_i \leq \Gamma(C_i)$ . We take advantage of the equation  $\{\overline{s_{ij}^n} \rightarrow s_i = \overline{\alpha_{ij}^{n_i}} \rightarrow \alpha_x\}$ , so that we get  $\overline{\theta(s_{ij}^n)}^{n_i} = \overline{\theta(\alpha_{ij})}^{n_i}$  for each  $j \in \{1 \dots n_i\}$  and  $\theta(s_i) = \theta(\alpha_x)$ . Moreover, we know that  $\rho'$  belongs to the domain of  $\theta$ , since the latter is defined for  $\theta(s_i)$  and the outermost region of  $s_i$  is  $\rho'$ . Therefore, we get the required result as follows:

$$\overline{\theta(\alpha_{ij})}^{n_i} \rightarrow \theta(\rho') \rightarrow \theta(\alpha_x) = \overline{\theta(s_{ij}^n)}^{n_i} \rightarrow \theta(\rho') \rightarrow \theta(s_i) \leq \overline{s_{ij}^n} \rightarrow \rho' \rightarrow s_i \leq \Gamma(C_i) = \theta(\Gamma(C_i))$$

□

Before embarking on the proof of the second step, we need an auxiliary definition which allows us to change the decorations of  $\epsilon$  of the form  $[k, \overline{s_i^n} \rightarrow s]$  by decorations of the form  $[\overline{\rho_j^m}]$ , appearing in non-recursive function applications. Of course, we can only do this when the set of additional region type parameters ( $R_{arg}$ ) is known, since otherwise we could not build the type signature  $\sigma$  of the function. By substituting  $[\varphi_k(R_{arg})]$  for each decoration  $[k, \overline{s_i^n} \rightarrow s]$  in  $\epsilon$  (as it is done indirectly in *annotateExp*), we obtain an expression  $\epsilon'$  which is also typable via the  $\Vdash_{\overline{\varphi}}$  rules.

**Example 4.12.** If  $\epsilon$  is the expression at the end of Example 4.6, we substitute  $[\varphi_1(\rho_4)] = [\rho_4]$  for the decoration  $[1, [\alpha_3]@_{\rho_1} \rightarrow [a_3]@_{\rho_6} \rightarrow [a_3]@_{\rho_4}]$  in order to obtain:

$$\begin{aligned} \epsilon' &= \mathbf{case} \text{ xs of} \\ &\quad [] \rightarrow \text{ys } @_{[\rho_4]} \\ &\quad (x : \text{xx}) \rightarrow \mathbf{let} \ x_1 = \text{append } \text{xx } \text{ys}_{[\rho_4]} \ \mathbf{in} \ (x : x_1)_{[\rho_4]} \end{aligned}$$

□

**Example 4.13.** If  $\epsilon$  is the intermediate expression obtained in Example 4.8:

$$\epsilon = \mathbf{let} \ \text{zs} = (1 : \text{xs})_{[\rho_1]} \ \mathbf{in} \ f \ \text{ys } \text{zs}_{[1, [\text{Int}]@_{\rho_2} \rightarrow [\text{Int}]@_{\rho_1}]}$$

we get  $R_{arg} = \{\rho_1, \rho_2\}$ , so  $\varphi_1(R_{arg}) = [\rho_2, \rho_1]$  and we transform  $\epsilon$  as follows:

$$\epsilon' = \mathbf{let} \ \text{zs} = (1 : \text{xs})_{[\rho_1]} \ \mathbf{in} \ f \ \text{ys } \text{zs}_{[\rho_2, \rho_1]}$$

□

Now we can prove the correctness of the second part of the algorithm, provided that the choice of  $R_{arg}$ ,  $R_{self}$  and  $R_{expl}$  satisfies the conditions enumerated in Section 4.4.2:

**Theorem 4.14.** Let us assume  $\Gamma + [f : \bar{s}_i^n \rightarrow s] \Vdash_{\bar{\varphi}} \epsilon : s'$  where every annotation in  $\epsilon$  of the form  $[k, sf]$  occur in a call to  $f$ . Let  $R_{arg}$ ,  $R_{self}$  and  $R_{expl}$  sets of RTVs such that the conditions (1), (2) and (4) hold, where:

$$\begin{aligned} R_{in} &= \{regions(s_i) \mid i \in \{1 \dots n\}\} \\ R_{out} &= \{regions(s)\} \\ R_{arg} &= \{\rho_1, \dots, \rho_m\}, \text{ for some } \rho_1, \dots, \rho_m \end{aligned}$$

and the  $\epsilon$ ,  $\bar{\varphi}_i^p$  occurring in condition (4) are those of the  $\Vdash$  derivation above. If we define  $\mathcal{R}$  as follows:

$$\mathcal{R} = [\bar{\rho}_i \mapsto r_i^m] \uplus [\rho \mapsto self \mid \rho \in R_{self}]$$

Then:

(a)  $\mathcal{R}$  is well-defined, i.e. the  $\uplus$  operator is applied to region mappings with disjoint definition domains.

(b) There exists an  $\epsilon' \in \mathbf{Exp}_1$  not containing annotations of the form  $[k, sf]$ , such that:

1.  $\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] \Vdash_{\bar{\varphi}} \epsilon' : s'$
2.  $R_{expl} = regions(\epsilon')$
3.  $annotateExp \mathcal{R} \epsilon' \bar{\varphi}_i^p \{\bar{\rho}_j^m\} = annotateExp \mathcal{R} \epsilon \bar{\varphi}_i^p \{\bar{\rho}_j^m\}$

(c) If  $e = annotateExp \mathcal{R} \epsilon \bar{\varphi}_i^p \{\bar{\rho}_j^m\}$ , then  $e$  is well-defined and the following judgment can be derived by using the  $\vdash_{Reg}$  rules:

$$\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} e : s' \quad (4.4)$$

*Proof.* (a) follows trivially from condition (2). Now we prove that the expression  $\epsilon'$  that results from replacing every annotation of the form  $[k, sf]$  by  $[\varphi_k^+(\rho_j)^m]$  satisfies the conditions stated in (b):

1. The first condition can be proven by induction on the  $\Vdash_{\bar{\varphi}}$  derivation applied to  $\epsilon$ . Most cases follow trivially from the induction hypothesis, except when the rule [APP-R<sub>HM</sub>] is applied. In this case  $\epsilon$  has an annotation of the form  $[k, sf]$ . This implies, by hypothesis, that  $\epsilon$  is a call to  $f$ , and that  $sf$  has the form  $\bar{s}_i^n \rightarrow s'$ . Our aim is to apply the [APP-NR<sub>HM</sub>] rule to its counterpart in  $\epsilon'$ , which is a call to  $f$  decorated with  $[\varphi_k^+(\rho_j)^m]$ . We have to ensure that  $\bar{s}_i^n \rightarrow \overline{\varphi_k(\rho_j)^m} \rightarrow s'$  is an instance of  $\forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s$ , but this follows from the premises of the [APP-R<sub>HM</sub>] rule:

$$\bar{s}_i^n \rightarrow \overline{\varphi_k(\rho_j)^m} \rightarrow s' = \overline{\varphi_k(s_i)^n} \rightarrow \overline{\varphi_k^+(\rho_j)^m} \rightarrow \varphi_k(s) = \varphi_k^+(\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s) \leq \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s$$

2. We know that  $regions(\epsilon) \subseteq regions(\epsilon')$ , since the annotations in  $\epsilon$  of the form  $[\rho]$  and  $[\bar{\rho}_j]$  are left untouched in  $\epsilon'$ . The only new RTVs in  $regions(\epsilon')$  are those that arise from replacing the  $[k, sf]$  annotations (which are not taken into account in the  $regions$  function) with their counterparts in  $\epsilon'$ . Therefore, we get:

$$regions(\epsilon') = regions(\epsilon) \cup \bigcup_{j=1}^p \{\varphi_j^+(\rho_i) \mid 1 \leq i \leq m\} = regions(\epsilon) \cup \bigcup_{j=1}^p \varphi_j^+(R_{arg})$$

which is, by condition (4), equal to  $R_{expl}$ .

3. The last condition follows trivially from the definition of  $annotateExp$ .

Lastly we prove **(c)**. From the results given in **(b)** we can prove the existence of an  $\epsilon'$  for which **(b).1** can be derived without using the  $[APP-R_{HM}]$  rule, since  $\epsilon'$  does not contain annotations of the form  $[k, sf]$ . Hence we prove the following: given  $\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] \Vdash_{\bar{\varphi}} \epsilon' : s$  and the sets  $R_{arg}$ ,  $R_{self}$  and  $R_{expl}$  such that the conditions **(1)**, **(2)** and **(4)** hold, where  $R_{in}$  and  $R_{out}$  are defined as above, then we can derive **(4.4)** being  $e = \text{annotateExp } \mathcal{R} \epsilon' \bar{\varphi}_i^p \{ \bar{\rho}_j^m \}$ , which we know that is equivalent to  $\text{annotateExp } \mathcal{R} \epsilon \bar{\varphi}_i^p \{ \bar{\rho}_j^m \}$ . We apply structural induction on  $\epsilon'$ . Let us distinguish cases according to the structure of  $\epsilon'$ :

- **Cases**  $\epsilon' \equiv c, x$

Let us compare the HM-rules that have been applied in the  $\Vdash_f$  derivation with the rules we want to apply in order to derive **(4.4)**:

$$\frac{}{\Gamma \Vdash_{\bar{\varphi}} c : B} [LIT_{HM}] \quad \frac{}{\Gamma \vdash_{Reg} c : B} [LIT_{Reg}]$$

$$\frac{}{\Gamma + [x : s'] \Vdash_{\bar{\varphi}} x : s'} [VAR_{HM}] \quad \frac{}{\Gamma + [x : s'] \vdash_{Reg} x : s'} [VAR_{Reg}]$$

We can trivially obtain the judgements on the right-hand side from those on the left-hand side. We can extend the environment by applying Lemma 3.48 in order to get **(4.4)**.

- **Case**  $\epsilon' \equiv x@_{[\rho]}$

$$\frac{\Gamma(x) = T@_{\rho'}}{\Gamma \Vdash_{\bar{\varphi}} x@_{[\rho]} : T@_{\rho}} [COPY_{HM}] \quad \frac{}{\Gamma + [x : T@_{\rho'}, r : \rho] \vdash_{Reg} x@_r : T@_{\rho}} [COPY_{Reg}]$$

Since  $\rho \in R_{expl}$ , by condition **(1)** we get  $\rho \in R_{arg} \cup R_{self}$ . We distinguish cases:

$\rho \in R_{arg}$  Then  $\rho$  is one of the  $\rho_i$  ( $i \in \{1 \dots m\}$ ), so that  $e \equiv x @ r_i$  and we can derive  $\Gamma + [x : T@_{\rho}, r_i : \rho_i] \vdash_{Reg} x @ r_i$ .

$\rho \in R_{self}$  We obtain  $e \equiv x @ self$  and hence  $\Gamma + [x : T@_{\rho}, self : \rho_{self}] \vdash_{Reg} x @ self$ .

In both cases we can extend the environment until we obtain **(4.4)**.

- **Case**  $\epsilon' \equiv C \bar{a}_i^q_{[\rho]}$

The following rule has been applied in the  $\Vdash_{\bar{\varphi}}$  derivation:

$$\frac{\bar{s}_i^q \rightarrow \rho \rightarrow s' \trianglelefteq \Gamma(C) \quad \forall i \in \{1 \dots q\} : \Gamma \Vdash_{\bar{\varphi}} a_i : s_i}{\Gamma \Vdash_{\bar{\varphi}} C \bar{a}_i^q_{[\rho]} : s'} [CONS_{HM}]$$

We need to apply this rule to get the desired result:

$$\frac{\bar{s}_i^q \rightarrow \rho \rightarrow s' \trianglelefteq \Gamma(C) \quad \forall i \in \{1 \dots q\} : \Gamma \vdash_{Reg} a_i : s_i}{\Gamma + [r : \rho] \vdash_{Reg} C \bar{a}_i^q @ r : s'} [CONS_{Reg}]$$

We know that  $\text{annotateExp } \mathcal{R} a_i = a_i$  for every  $a_i$ , so we can apply the induction hypothesis on each  $a_i$  in order to get:

$$\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [r_i : \rho_i^m] + [self : \rho_{self}] \vdash_{Reg} a_i : s_i \quad \text{for } i \in \{1 \dots n\}$$

Now we have to prove that the addition of  $[r : \rho]$  to this environment (as required by rule  $[CONS_{Reg}]$ ) results in a well-defined environment. Since  $\rho \in R_{expl}$ , we get  $\rho \in R_{arg} \cup R_{self}$  by condition (1). Again, we distinguish cases:

$\rho \in R_{arg}$  Then  $\rho = \rho_i$  for some  $i \in \{1 \dots m\}$ , so we get  $e \equiv C \bar{a}_i^n @ r_i$  as our final expression. Rule  $[CONS_{Reg}]$  would require us to add the binding  $[r_i : \rho_i]$ .

$\rho \in R_{self}$  This case leads to  $e \equiv C \bar{a}_i^n @ self$ , so we would have to add  $[self : \rho_{self}]$ .

In both cases the resulting environment is well-defined and we can apply  $[CONS_{Reg}]$  in order to derive (4.4).

- **Case**  $\epsilon' \equiv g \bar{a}_i^p \overline{[\rho_j^q]}$

$$\frac{\bar{s}_i^p \rightarrow \bar{\rho}_j^q \rightarrow s' \trianglelefteq \sigma \quad \forall i \in \{1 \dots p\} : \Gamma \Vdash_{\bar{\varphi}} a_i : s_i}{\Gamma + [g : \sigma] \Vdash_{\bar{\varphi}} g \bar{a}_i^p \overline{[\rho_j^q]} : s'} \quad [APP_{NR_{HM}}]$$

$$\frac{\bar{s}_i^p \rightarrow \bar{\rho}_j^q \rightarrow s' \trianglelefteq \sigma \quad \forall i \in \{1 \dots p\} : \Gamma \vdash_{Reg} a_i : s_i}{\Gamma + [g : \sigma] + [\bar{r}'_j : \bar{\rho}_j^q] \vdash_{Reg} g \bar{a}_i^p @ \bar{r}'_j : s'} \quad [APP_{Reg}]$$

By applying the induction hypothesis in every subexpression we get, for every  $i \in \{1 \dots p\}$ :

$$\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} a_i : s_i \quad (4.5)$$

If  $g$  is distinct from  $f$ , since  $g$  does not occur in any of the  $a_i$ , we can apply Lemma 3.48:

$$\Gamma + [g : \sigma] + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} a_i : s_i \quad (4.6)$$

The resulting *Core-Safe* expression is  $g \bar{a}_i^p @ \overline{\mathcal{R}(\rho_j^q)}$ . We denote by  $r'_j$  the result of each  $\mathcal{R}(\rho_j^q)$ , as shown in the rule  $[APP_{Reg}]$  above. Now we have to show, for each  $j \in \{1 \dots q\}$ , that the addition of the bindings  $[r'_j : \rho_j^q]$  does not contradict with the bindings  $[\bar{r}_i : \bar{\rho}_i^m]$ . This is very similar to the cases for copy and constructor application: we have  $\rho_j^q \in R_{expl} \subseteq R_{arg} \cup R_{self}$ . On the one hand, if  $\rho_j^q \in R_{arg}$  then  $\rho_j^q$  is one of the  $\rho_i$  for  $i \in \{1 \dots m\}$  and hence  $r'_j = r_i$ , so the binding  $[r'_j : \rho_j^q]$  is equivalent to  $[r_i : \rho_i]$ . On the other hand, if  $\rho_j^q \in R_{self}$  then  $r'_j = self$  and we have the binding  $[self : \rho_{self}]$  as a result. In any case, the binding  $[r'_j : \rho_j^q]$  is compatible with the typing environment above. So, we can apply rule  $[APP_{Reg}]$  by using the judgments (4.5) or (4.6) as premises (depending on whether  $g$  is equal to  $f$  or not) so as to get:

$$\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} g \bar{a}_i^p @ \bar{r}'_j : s'$$

- **Case**  $\epsilon' \equiv \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2$

$$\frac{\Gamma \Vdash_{\bar{\varphi}} \epsilon_1 : s_1 \quad \Gamma \setminus x_1 + [x_1 : s_1] \Vdash_{\bar{\varphi}} \epsilon_2 : s'}{\Gamma \Vdash_{\bar{\varphi}} \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2 : s'} \quad [LET_{HM}]$$

$$\frac{\Gamma \vdash_{Reg} \epsilon_1 : s_1 \quad \Gamma \setminus x_1 + [x_1 : s_1] \vdash_{Reg} \epsilon_2 : s'}{\Gamma \vdash_{Reg} \mathbf{let} \ x_1 = \epsilon_1 \ \mathbf{in} \ \epsilon_2 : s'} \quad [LET_{Reg}]$$

We apply the induction hypothesis to each subexpression  $\epsilon_1, \epsilon_2$  in order to obtain:

$$\begin{aligned} & \Gamma + [f : \forall \bar{\rho}. \bar{t}_i^n \rightarrow \bar{\rho}_j^m \rightarrow t] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} e_1 : s_1 \\ & \Gamma \setminus x_1 + [f : \forall \bar{\rho}. \bar{t}_i^n \rightarrow \bar{\rho}_j^m \rightarrow t] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] + [x_1 : s_1] \vdash_{Reg} e_2 : s' \end{aligned}$$

where  $e_1 = \text{annotateExp } \mathcal{R} \ e_1$  and  $e_2 = \text{annotateExp } \mathcal{R} \ e_2$ . We can apply the rule  $[LET_{Reg}]$  in order to prove (4.4).

- **Case**  $\epsilon' \equiv \text{case}(!) \ x \ \text{of} \ \overline{C_i} \ \overline{x_{ij}^{n_i}} \rightarrow \epsilon_i^n$

From the premises of the rule  $[CASE(!)_{HM}]$ , the judgment  $\Gamma + [\overline{x_{ij} : s_{ij}^{n_i}}] \Vdash_{\bar{\varphi}} \epsilon_i : s'$  holds.

$$\frac{\Gamma(x) = s_x \quad \forall i \in \{1..n\} : \left\{ \begin{array}{l} \overline{s_{ij}^{n_i}} \rightarrow \rho \rightarrow s_x \trianglelefteq \Gamma(C_i) \\ \Gamma \setminus \overline{x_{ij}^{n_i}} + [\overline{x_{ij} : s_{ij}^{n_i}}] \Vdash_{\bar{\varphi}} \epsilon_i : s' \end{array} \right.}{\Gamma \Vdash_{\bar{\varphi}} \text{case}(!) \ x \ \text{of} \ \overline{C_i} \ \overline{x_{ij}^{n_i}} \rightarrow \epsilon_i^n : s'} \quad [CASE(!)_{HM}]$$

$$\frac{\forall i \in \{1 \dots r\} : \Gamma(C_i) = \sigma_i \quad \forall i \in \{1 \dots r\} : \overline{s_{ij}^{n_i}} \rightarrow \rho \rightarrow s_x \trianglelefteq \sigma_i \quad \Gamma(x) = s_x \quad \forall i \in \{1 \dots r\} : \Gamma \setminus \overline{x_{ij}^{n_i}} + [\overline{x_{ij} : s_{ij}^{n_i}}] \vdash_{Reg} e_i : s'}{\Gamma \vdash_{Reg} \text{case}(!) \ x \ \text{of} \ \overline{C_i} \ \overline{x_{ij}^{n_i}} \rightarrow e_i^r : s'} \quad [CASE(!)_{Reg}]$$

By applying the induction hypothesis to each  $\epsilon_i$ , we get:

$$\Gamma \setminus \overline{x_{ij}^{n_i}} + [f : \forall \bar{\rho}. \bar{t}_i^n \rightarrow \bar{\rho}_j^m \rightarrow t] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] + [\overline{x_{ij} : s_{ij}^{n_i}}] \vdash_{Reg} e_i : s'$$

for each  $i \in \{1 \dots r\}$ , with  $e_i = \text{annotateExp } \mathcal{R} \ \epsilon_i$ . Therefore, by rule  $[CASE(!)_{Reg}]$  we can prove (4.4), since the rest of the premises of this rule are the same as their counterparts in  $[CASE(!)_{HM}]$ .  $\square$

Notice that the premises of this theorem do not include condition (3), since it is not necessary in the proof. This theorem guarantees the existence of a  $\vdash_{Reg}$  typing for the function's body. In Chapter 5 we put forward an algorithm for dealing with explicit destruction via  $\text{case}(!)$ , whose correctness proof guarantees the existence of a  $\vdash_{Dst}$  typing. Both  $\vdash_{Reg}$  and  $\vdash_{Dst}$  typings lead to a  $\vdash$  typing for the body of the function, as proven in Section 3.7. However, we need to satisfy an additional constraint in order to type the whole function definition by applying the rule  $[FUN]$  of Figure 3.9. This rule specifies that the region type  $\rho_{self}$  must not occur in the resulting type signature. However, we can achieve this by assuming the condition (3):

**Corollary 4.15.** *Under the same conditions of Theorem 4.14 and assuming that (3) in Section 4.4.2 holds, we can derive the following judgement:*

$$\Gamma + [f : \forall \bar{\rho}. \bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s] + [\bar{r}_i : \bar{\rho}_i^m] + [self : \rho_{self}] \vdash_{Reg} e : s'$$

where  $\rho_{self} \notin \text{regions}(\bar{s}_i^n \rightarrow \bar{\rho}_j^m \rightarrow s)$ .

*Proof.* It follows trivially from Theorem 4.14 and condition (3).  $\square$

Our last step in the correctness proof establishes that the *inferRegions* algorithm returns a solution which satisfies the conditions (1) to (4). The main part of this algorithm is the fixpoint iteration performed by *computeRargFP*. The following invariant determines the origin of every variable occurring in

$R_{arg}$ : either it belongs to the initial set of explicit regions, or it is the image of another variable already present in  $R_{arg}$  w.r.t some region mapping  $\varphi_i$ .

**Lemma 4.16.** *Given the call  $computeRargFP R_{in} R_{out} R_{arg} R_0 \overline{\varphi_i^p}$ , where  $R_{arg} = \emptyset$ , the following invariant holds in every subsequent iteration: for every  $\rho \in R_{arg}$  there exists a sequence of RTVs  $[\rho_0, \dots, \rho_n] \subseteq R_{arg}$  with  $n \geq 0$  and  $\rho = \rho_n$  such that:*

1.  $\rho_0 \in R_0$
2. For all  $\rho_i$  (with  $i \in \{1 \dots n\}$ ) there exists some  $\varphi_k$  such that  $\varphi_k(\rho_{i-1}) = \rho_i$

*Proof.* The invariant holds trivially in the root call, since  $R_{arg} = \emptyset$ . Now let us assume that it holds for the input argument  $R_{arg}$ . We have to prove the invariant for the  $R'_{arg}$  appearing in the recursive call. Let  $\rho \in R'_{arg}$ . By inspecting the definition in Figure 4.7 we know that  $\rho \in R'_{expl}$ . So, either  $\rho \in R_0$  (and the invariant holds considering the sequence  $[\rho]$ ) or  $\rho = \varphi_k(\rho')$  for some  $k \in \{1 \dots p\}$  and  $\rho' \in R_{arg}$ . In the latter case there exists a sequence  $[\rho_0, \dots, \rho_n = \rho']$  which satisfies that required condition, and which we can extend as follows  $[\rho_0, \dots, \rho_n = \rho', \rho]$ .  $\square$

**Example 4.17.** Given the following function fragment with a single recursive call:

$$f \ x_1 \ x_2 \ x_3 = \dots \ f \ x_2 \ x_3 \ x_1 \ \dots$$

Let us assume that  $x_1$  has type  $[\alpha]@_{\rho_1}$ ,  $x_2$  has type  $[\alpha]@_{\rho_2}$  and  $x_3$  has type  $[\alpha]@_{\rho_3}$ . Therefore  $R_{in} = \{\rho_1, \rho_2, \rho_3\}$ . We compare the types of the function and its recursive call:

$$\begin{aligned} \text{Function: } f &:: [\alpha]@_{\rho_1} \rightarrow [\alpha]@_{\rho_2} \rightarrow [\alpha]@_{\rho_3} \\ \text{Recursive application: } f &:: [\alpha]@_{\rho_2} \rightarrow [\alpha]@_{\rho_3} \rightarrow [\alpha]@_{\rho_1} \end{aligned}$$

Hence we get  $\varphi_1 = [\rho_1 \mapsto \rho_2, \rho_2 \mapsto \rho_3, \rho_3 \mapsto \rho_1]$ . If  $R_0 = \{\rho_1\}$  the fixpoint iteration takes the following steps:

	Initial call	1 <sup>st</sup> iteration	2 <sup>nd</sup> iteration	3 <sup>rd</sup> iteration
$R_{arg}$	$\emptyset$	$\{\rho_1\}$	$\{\rho_1, \rho_2\}$	$\{\rho_1, \rho_2, \rho_3\}$
$R_{expl}$	$\{\rho_1\}$	$\{\rho_1\}$	$\{\rho_1, \rho_2\}$	$\{\rho_1, \rho_2, \rho_3\}$

The variable  $\rho_1$  belongs to  $R_{arg}$  because it appears initially in  $R_0$ . Variable  $\rho_2$  is added to  $R_{arg}$  because  $\varphi(\rho_1) = \rho_2$  and  $\rho_1$  was already in  $R_{arg}$ . Variable  $\rho_3$  is added to  $R_{arg}$  because  $\varphi(\rho_2) = \rho_3$  and  $\rho_2$  was already in  $R_{arg}$ . We have the following sequences for each variable:

$$\rho_1 : [\rho_1] \quad \rho_2 : [\rho_1, \rho_2] \quad \rho_3 : [\rho_1, \rho_2, \rho_3]$$

$\square$

This invariant is useful for proving that the solution given by *inferRegions* is optimal with respect to the conditions (1) to (4) of Section 4.4.2. Any other solution satisfying these conditions may only lead to equal or worse results, i.e. a bigger  $R_{arg}$ , and hence a smaller  $R_{self}$ .

**Theorem 4.18.** *The algorithm *inferRegions* in Figure 4.7 returns three sets  $(R_{arg}, R_{self}, R_{expl})$  which satisfy the four conditions described in Section 4.4.2. Moreover, for every solution  $(R'_{arg}, R'_{self}, R'_{expl})$  satisfying these conditions it holds that  $R_{arg} \subseteq R'_{arg}$  and  $R_{expl} \subseteq R'_{expl}$ .*

*Proof.* The condition (4) follows trivially from the definition of *computeRargFP*. We can rewrite the remaining conditions by using (4.2) and (4.3):

1.  $R_{expl} \subseteq (R_{expl} \setminus (R_{in} \cup R_{out})) \cup (R_{expl} \cap (R_{in} \cup R_{out}))$
2.  $(R_{expl} \setminus (R_{in} \cup R_{out})) \cap (R_{expl} \cap (R_{in} \cup R_{out})) = \emptyset$
3.  $(R_{expl} \setminus (R_{in} \cup R_{out})) \cap (R_{in} \cup R_{out}) = \emptyset$

The three immediately follow from set algebra. Now we prove optimality: let us assume that the four conditions hold for  $(R'_{arg}, R'_{self}, R'_{expl})$ . First we prove that  $R_{arg} \subseteq R'_{arg}$ . If  $\rho \in R_{arg}$  then there exists (by Lemma 4.16) a sequence  $[\rho_0, \dots, \rho_n = \rho] \subseteq R_{arg}$  with  $\rho_0 \in regions(\epsilon)$ . Notice that, by the invariant (4.2), all these RTVs must belong to  $R_{in} \cup R_{out}$ . We show that every  $\rho_i$  belongs to  $R'_{arg}$  by induction on  $i$ .

- **Case  $i = 0$ .** In this case  $\rho_0 \in regions(\epsilon) \subseteq R'_{expl}$ . Since  $\rho_0 \in R_{in} \cup R_{out}$ , then  $\rho_0 \notin R'_{self}$  because of condition (3). Hence, by condition (1), we get  $\rho_0 \in R'_{arg}$ .
- **Case  $i > 0$ .** In this case  $\rho_i = \varphi_k(\rho_{i-1})$  for some  $k$ . By induction hypothesis  $\rho_{i-1} \in R'_{arg}$  and hence  $\rho_i \in R'_{expl}$  by condition (4). Again,  $\rho_i$  cannot belong to  $R'_{self}$  because it belongs to  $R_{in} \cup R_{out}$  and condition (3) forces them to be disjoint. Hence, by condition (1),  $\rho_i \in R'_{arg}$ .

Therefore,  $\rho = \rho_n \in R'_{arg}$  and  $R_{arg} \subseteq R'_{arg}$ . From this inclusion and the condition (4) it follows that  $R_{expl} \subseteq R'_{expl}$ . □

This theorem proves that the resulting  $R_{arg}$  is the minimum solution for the conditions (1) to (4). Given a fixed set  $R_{expl}$  of explicit variables, it follows from conditions (1) and (2) of Section 4.4.2 that  $R_{self}$  is the maximum solution.

## 4.6 Case studies

In this section we study the results of the region inference algorithm applied to several examples, ranging from simple functions on lists to a compiler running in several phases. Although the algorithm is defined at *Core-Safe* level, all the examples in this section, as well as their region-annotated versions, are written in *Full-Safe* for conciseness.

**Example 4.19** (List partition). As a first example, consider the following function definition:

$$\begin{aligned}
 & \text{partition } y \ [] = ([], []) \\
 & \text{partition } y \ (x : xs) \\
 & \quad | x \leq y = (x : ls, gs) \\
 & \quad | x > y = (ls, x : gs) \\
 & \quad \text{where } (ls, gs) = \text{partition } y \ xs
 \end{aligned}$$

The HM inference phase yields the following decorated program:

$$\begin{aligned}
 & \text{partition } y \ [] = ([]_{[\rho_2]}, []_{[\rho_3]})_{[\rho_4]} \\
 & \text{partition } y \ (x : xs) \\
 & \quad | x \leq y = ((x : ls)_{[\rho_2]}, gs)_{[\rho_4]} \\
 & \quad | x > y = (ls, (x : gs)_{[\rho_3]})_{[\rho_4]} \\
 & \quad \text{where } (ls, gs) = \text{partition } y \ xs_{[1, Int \rightarrow [Int]@_{\rho_1} \rightarrow ([Int]@_{\rho_2}, [Int]@_{\rho_3})@_{\rho_0}]}
 \end{aligned}$$

and the type  $partition :: Int \rightarrow [Int]@_{\rho_1} \rightarrow ([Int]@_{\rho_2}, [Int]@_{\rho_3})@_{\rho_4}$ . By comparing this type with the type of the recursive call we get the following mapping:

$$\varphi_1 = [\rho_1 \mapsto \rho_1, \rho_2 \mapsto \rho_2, \rho_3 \mapsto \rho_3, \rho_4 \mapsto \rho_9]$$

Notice that, in this case, we have polymorphic recursion over the RTV of the pair which is given as a result. We start the fixpoint iteration with these sets:

$$R_{in} = \{\rho_1\} \quad R_{out} = \{\rho_2, \rho_3, \rho_4\} \quad R_{arg} = \emptyset \quad R_0 = \{\rho_2, \rho_3, \rho_4\}$$

After the first iteration, we get:

$$R_{expl} = \{\rho_2, \rho_3, \rho_4\} \quad R_{arg} = \{\rho_2, \rho_3, \rho_4\}$$

In the next iteration the set  $R_{expl}$  grows, since  $\varphi_1(\rho_4) = \rho_9$ :

$$R_{expl} = \{\rho_2, \rho_3, \rho_4, \rho_9\} \quad R_{arg} = \{\rho_2, \rho_3, \rho_4\}$$

However, since  $R_{arg}$  does not change, we stop the fixpoint iteration and compute  $R_{self} = \{\rho_9\}$  by applying (4.3). Finally, we generate fresh region variables  $r_2, r_3$  and  $r_4$  corresponding to the RTVs in  $R_{arg}$  and build the following mapping:

$$\mathcal{R} = [\rho_2 \mapsto r_2, \rho_3 \mapsto r_3, \rho_4 \mapsto r_4, \rho_9 \mapsto self]$$

which leads to the following annotated function definition:

$$\begin{aligned} & partition :: Int \rightarrow [Int]@_{\rho_1} \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow \rho_4 \rightarrow ([Int]@_{\rho_2}, [Int]@_{\rho_3})@_{\rho_4} \\ & partition \ y \ [] \ @ \ r_2 \ r_3 \ r_4 = ([ ]@_{r_2}, [ ]@_{r_3})@_{r_4} \\ & partition \ y \ (x : xs) \ @ \ r_2 \ r_3 \ r_4 \\ & \quad | \ x \leq \ y = ((x : ls) \ @ \ r_2, gs) \ @ \ r_4 \\ & \quad | \ x > \ y = (ls, (x : gs) \ @ \ r_3) \ @ \ r_4 \\ & \quad \mathbf{where} \ (ls, gs) = partition \ y \ xs \ @ \ r_2 \ r_3 \ self \end{aligned}$$

In absence of polymorphic recursion, the recursive call would be  $partition \ y \ xs \ @ \ r_2 \ r_3 \ r_4$ , which implies that all the tuples returned from the internal calls to  $partition$  are built in the output region, even when these tuples are temporary and not part of the function's result.  $\square$

**Example 4.20** (Pascal's triangle). Let us consider the dynamic approach for computing binomial coefficients by using Pascal's triangle. We start from the unit list  $[1]$ , which is the 0-th row of the triangle. If  $[x_0, x_1, \dots, x_{i-1}]$  are the elements of the  $i$ -th row, the elements of the  $i + 1$ -th row are given by the list  $[1, x_0 + x_1, x_1 + x_2, \dots, x_{i-1} + x_i, x_i]$ . This is shown in Figure 4.10. The binomial coefficient  $\binom{n}{m}$  can be obtained from the  $m$ -th element in the  $n$ -th row of the triangle.

The function  $sumList$  computes the  $i + 1$ -th row of the triangle from its  $i$ -th row. We show directly the intermediate definition:

$$\begin{aligned} sumList \ (x : []) &= (x : [ ]_{[\rho_2]})_{[\rho_2]} \\ sumList \ (x : xs) &= ((x + y) : sumList \ xs_{[1, [Int]@_{\rho_1} \rightarrow [Int]@_{\rho_2}]_{[\rho_2]}})_{[\rho_2]} \quad \mathbf{where} \ (y : \_) = xs \end{aligned}$$

The fixpoint iteration produces  $R_{expl} = R_{arg} = \{\rho_2\}$  and  $R_{self} = \emptyset$ , which leads to the following

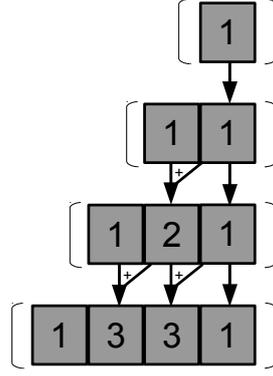


Figure 4.10: Pascal's triangle computation.

definition:

$$\begin{aligned} \text{sumList } (x : []) @ r &= (x : []) @ r \\ \text{sumList } (x : xs) @ r &= ((x + y) : \text{sumList } xs @ r) @ r \quad \textbf{where } (y : \_) = xs \end{aligned}$$

Function *pascal* iterates over the initial list in order to get the row, whose number is given as parameter. Below we show the decorated definition:

$$\begin{aligned} \text{pascal } 0 &= (1 : [])_{[\rho_1]}_{[\rho_1]} \\ \text{pascal } n &= (1 : (\text{sumList } (\text{pascal } (n - 1)))_{[1, \text{Int} \rightarrow [\text{Int} @ \rho_2]]}_{[\rho_1]}_{[\rho_1]})_{[\rho_1]} \end{aligned}$$

The type inferred for *pascal* after the HM phase is  $\text{Int} \rightarrow [\text{Int}] @ \rho_1$ . By comparing this type with that of the recursive call we get  $\varphi_1 = [\rho_1 \mapsto \rho_2]$ . Hence  $R_{in} = \emptyset$ ,  $R_{out} = \{\rho_1\}$  and  $R_0 = \{\rho_1\}$ , which gives us an initial  $R_{expl} = R_{arg} = \{\rho_1\}$ . After the second iteration  $\rho_2$  is now made explicit and it is added to  $R_{expl}$ , which now contains the region variables  $\{\rho_1, \rho_2\}$ . However,  $R_{arg}$  stays the same and hence the fixpoint has been reached. Finally, we get  $R_{self} = \{\rho_2\}$  and the program is annotated as follows:

$$\begin{aligned} \text{pascal} &:: \text{Int} \rightarrow \rho_1 \rightarrow [\text{Int}] @ \rho_1 \\ \text{pascal } 0 @ r &= (1 : [] @ r) @ r \\ \text{pascal } n @ r &= (1 : \text{sumList } (\text{pascal } (n - 1) @ self) @ r) @ r \end{aligned}$$

The resulting list from the recursive call to *pascal* will be destroyed once the calling function finishes. Hence a function call *pascal* *n* has a cost of  $\mathcal{O}(n)$  in space. Without polymorphic recursion the result of every recursive call would be built in the output region *r*, which would imply a cost of  $\mathcal{O}(n^2)$  in space (Figure 4.11).

The function definition in charge of computing the binomial coefficient  $\binom{n}{m}$  is defined as follows:

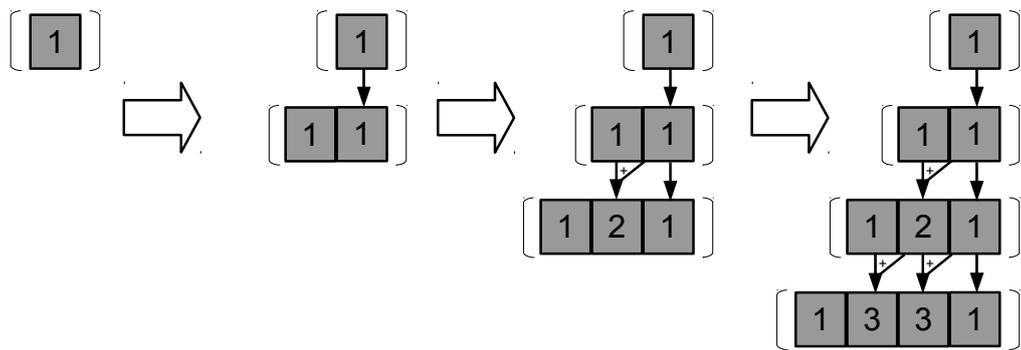
$$\text{combNumber } n \ m = \text{pascal } n \ !! \ m$$

The *!!* operator returns the *m*-th element of the list returned by *pascal*, which is inferred as residing in *self*, since is not part of the function's result. Hence the annotated program is:

$$\text{combNumber } n \ m = (\text{pascal } n @ self) !! m$$

□

Without polymorphic recursion:



With polymorphic recursion:

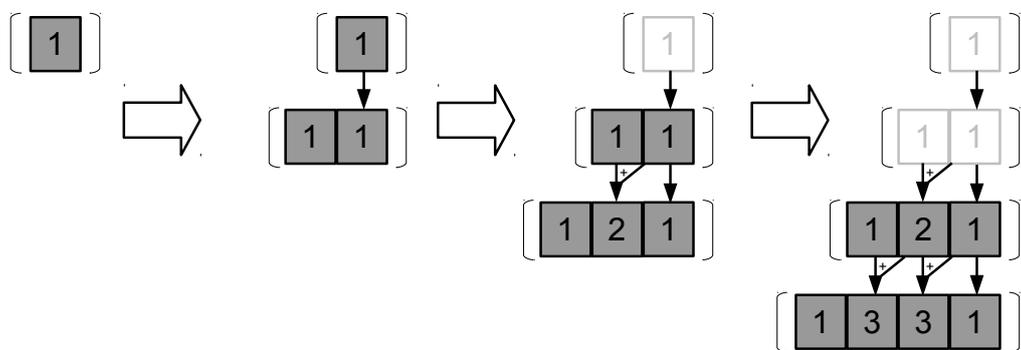


Figure 4.11: Lists being created by the function *pascal* with and without polymorphic recursion. Each column corresponds to the end of a recursive call. White cells are destroyed.

```

data BSTree  $\alpha$  @  $\rho = \text{Empty} @ \rho \mid \text{Node} (\text{BSTree } \alpha @ \rho) \alpha (\text{BSTree } \alpha @ \rho) @ \rho

\textit{insertT} :: \text{Int} \rightarrow \text{BSTree Int} @ \rho_1 \rightarrow \rho_1 \rightarrow \text{BSTree Int} @ \rho_1
\textit{insertT} y \text{Empty} @ r_1 = \text{Node} (\text{Empty} @ r_1) y (\text{Empty} @ r_1) @ r_1
\textit{insertT} y (\text{Node } l x r) @ r_1
    \mid x == y = \text{Node } l x r @ r_1
    \mid y < x  = \text{Node} (\textit{insertT} y l @ r_1) x r @ r_1
    \mid y > x  = \text{Node } l x (\textit{insertT} y r @ r_1) @ r_1

\textit{mkTree} :: [\text{Int}]@ $\rho_1 \rightarrow \rho_2 \rightarrow \text{BSTree Int} @ \rho_2
\textit{mkTree} [] @ r_2 = \text{Empty} @ r_2
\textit{mkTree} (x : xs) @ r_2 = \textit{insertT} x (\textit{mkTree} xs @ r_2) @ r_2

\textit{inorder} :: \text{BSTree } \alpha @ \rho_1 \rightarrow \rho_2 \rightarrow [\alpha]@ $\rho_2
\textit{inorder} \text{Empty} @ r_2 = [] @ r_2
\textit{inorder} (\text{Node } l x r) @ r_2
    = \textit{append} (\textit{inorder} l @ \textit{self}) (\textit{append} ((x : [])@ $\textit{self}$ )@ $\textit{self}$ ) (\textit{inorder} r @ r_2) @ r_2 @ r_2

\textit{treeSort} :: [\text{Int}]@ $\rho_1 \rightarrow \rho_2 \rightarrow [\text{Int}]@ $\rho_2
\textit{treeSort} xs @ r_2 = \textit{inorder} (\textit{mkTree} xs @ \textit{self}) @ r_2$$$$$ 
```

Figure 4.12: Region inference for the treesort algorithm.

**Example 4.21** (Treesort ). Let us consider an implementation of the treesort algorithm. Figure 4.12 shows the resulting region-annotated program. Firstly, we define a data type *BSTree* for binary search trees and define a function *insertT* implementing the usual insertion in an ordered tree. From the type signature of *insertT* we deduce that both input and output trees must live in the same region at runtime, since we reuse children of the input tree when constructing the output. The function *mkTree* builds a binary search tree by successive insertions of the elements of a list, whereas *inorder* transforms a tree into an ordered list by means of an inorder traversal. Notice that, in the latter, we use the definition of *append* in which the list passed as second parameter is shared with the result, unlike the definition in Example 4.4. That is why the second recursive call to *inorder* is called with the output region parameter; its result will be shared with the output. The result of the first recursive call is done with the working region, because its result will be copied when being concatenated with the global result. Finally, the function *treeSort* calls these two functions above. The intermediate tree is not needed in the result, so it is built in the working region.  $\square$

Region-based memory management is a powerful approach when dealing with programs that work in several phases and produce intermediate data structures between them. Compilers are a typical example of these kind of programs.

**Example 4.22.** Given a simple imperative **while** language, we want to translate it into bytecode for a *P*-machine. Both source and target languages are defined in Figure 4.13. Notice that some data types are polymorphic on a type variable  $\alpha$ . This variable is used as a placeholder for decorations, as we will see below. The translation performs in four phases:

- **Type checking:** In this language we have two types, integers and booleans. This process decorates the AST with type information and checks whether the arguments given to the operators are of a suitable type.

$$\textit{typeCheck} :: \text{Stm } \alpha \rightarrow \text{Stm} (\text{Maybe Type})$$

<b>data</b> <i>BinArithOp</i>	= <i>Add</i>   <i>Sub</i>   <i>Mul</i>   <i>Div</i>	– binary arithmetic operators
<b>data</b> <i>UnArithOp</i>	= <i>Neg</i>	– unary arithmetic operators
<b>data</b> <i>RelOp</i>	= <i>Less</i>   <i>Eq</i>	– relational operators
<b>data</b> <i>BinBoolOp</i>	= <i>And</i>   <i>Or</i>	– binary logical operators
<b>data</b> <i>UnBoolOp</i>	= <i>Not</i>	– unary logical operators
<b>data</b> <i>Exp</i> $\alpha$	= <i>Const Int</i> $\alpha$	– constant
	<i>Var Int</i> $\alpha$	– variable
	<i>AppBinArithOp BinArithOp (Exp</i> $\alpha$ ) ( <i>Exp</i> $\alpha$ ) $\alpha$	– operator application
	<i>AppUnArithOp UnArithOp (Exp</i> $\alpha$ ) $\alpha$	
	<i>AppRelOp RelOp (Exp</i> $\alpha$ ) ( <i>Exp</i> $\alpha$ ) $\alpha$	
	<i>AppBinBoolOp BinBoolOp (Exp</i> $\alpha$ ) ( <i>Exp</i> $\alpha$ ) $\alpha$	
	<i>AppUnBoolOp UnBoolOp (Exp</i> $\alpha$ ) $\alpha$	
<b>data</b> <i>Stm</i> $\alpha$	= <i>Skip</i> $\alpha$	– does nothing
	<i>Assign Int (Exp</i> $\alpha$ ) $\alpha$	– variable assignment
	<i>Seq (Stm</i> $\alpha$ ) ( <i>Stm</i> $\alpha$ ) $\alpha$	– sequencing
	<i>If (Exp</i> $\alpha$ ) ( <i>Stm</i> $\alpha$ ) ( <i>Stm</i> $\alpha$ ) $\alpha$	– conditional
	<i>While (Exp</i> $\alpha$ ) ( <i>Stm</i> $\alpha$ ) $\alpha$	– loop
<b>data</b> <i>PInst</i> $\alpha$	= <i>Ldc Int</i>   <i>Load</i>   <i>Store</i>   <i>Jmp</i> $\alpha$   <i>JFalse</i> $\alpha$	– bytecode instructions
	<i>AddP</i>   <i>SubP</i>   <i>MulP</i>   <i>DivP</i>   <i>NegP</i>	
	<i>LtP</i>   <i>GtP</i>   <i>EqP</i>   <i>AndP</i>   <i>OrP</i>   <i>NotP</i>	

Figure 4.13: Source and target language definitions for Example 4.22.

- **Constant folding:** It computes at compile time those operations which involve only constants.

$$\text{constantFold} :: \text{Stm } \alpha \rightarrow \text{Stm } \alpha$$

- **Translation into bytecode:** This phase generates the bytecode with symbolic labels. Branching instructions refer to these labels.

$$\text{translate} :: \text{Stm } \alpha \rightarrow [\text{PInst Label}]$$

- **Patching:** It translates symbolic labels into natural numbers, as well as the arguments of the branching instructions.

$$\text{patch} :: [\text{PInst Label}] \rightarrow [\text{PInst Int}]$$

For the sake of simplicity we do not show the code of these functions here. In order to take advantage of regions, we need some functions *runToZ*, which executes the phase Z and the preceding ones.

$$\begin{aligned} \text{runToConstantFold } stm &= \text{let } stm' = \text{typeCheck } stm \text{ in } \text{constantFold } stm' \\ \text{runToTranslate } stm &= \text{let } stm' = \text{runToConstantFold } stm \text{ in } \text{translate } stm' \\ \text{runToPatch } stm &= \text{let } insts = \text{runToTranslate } stm \text{ in } \text{patch } insts \end{aligned}$$

We show in Figure 4.14 the number of extra region parameters inferred for data types and function definitions. A more qualitative analysis of the results is depicted in Figure 4.15. Rows represent the working regions corresponding to each phase, while the columns stand for the execution of each phase. The pictograms represent data structures being built in a given region during the execution of a phase.

Data type	Region types
<i>BinArithOp</i>	1
<i>UnArithOp</i>	1
<i>RelOp</i>	1
<i>BinBoolOp</i>	1
<i>UnBoolOp</i>	1
<i>Exp</i>	6
<i>Stm</i>	19
<i>PInst</i>	1

Function	Region parameters
<i>typeCheck</i>	10
<i>constantFold</i>	4
<i>translate</i>	8
<i>patch</i>	2

Figure 4.14: Quantitative results for the compiler example.

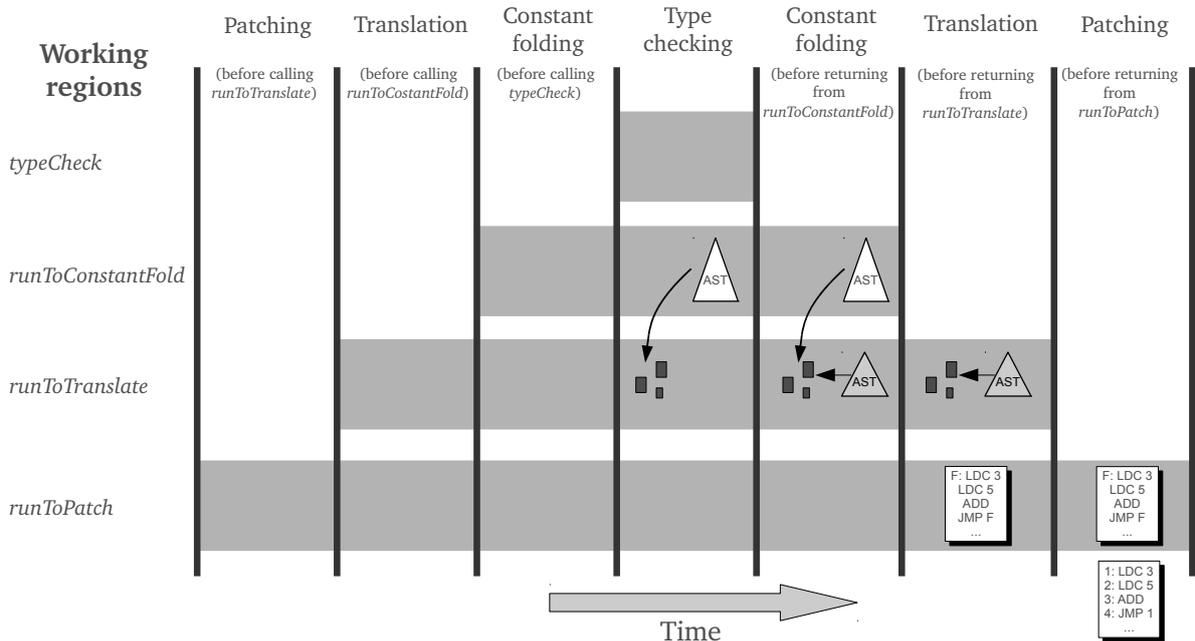


Figure 4.15: Evolution of the heap in each phase. Each gray rectangle represents the working region of a phase, its lifetime and the data structures that are stored into it. Each column stands for the state of the heap at a given time.

The type checking phase creates an AST with typing information. The result is distributed between the working regions of *constantFold* and *translate*, so that those in the latter region can be shared in subsequent phases. The constant folding phase generates another AST which reuses those parts lying in the *translate* working region. The initial AST with types is disposed after the execution of *constantFold*. The next phase creates the list of bytecode instructions in the *patch* working region. After this, the AST is no longer needed and hence its memory is reclaimed. Finally, the patching phase creates the final list of bytecode instructions without symbolic labels. The caller function decides where to locate this list by passing the suitable arguments to *patch*. Notice that, during the whole process, we only have two intermediate data structures simultaneously in the heap: the input and the output of each phase.  $\square$

## 4.7 Conclusions and related work

We have introduced an algorithm for annotating expressions with region variables. The algorithm is correct and optimal with respect to the  $\vdash_{Reg}$  typing rules. It has a time cost in  $\mathcal{O}(n^2)$  in the worst case, and  $\mathcal{O}(n)$  in the average case. The completeness of the algorithm can only be guaranteed when the function being analysed has a most general type, according to the  $\vdash_{Reg}$  type system. If such a type does not exist, the algorithm fails. The lack of principal types is due to copy expressions. For instance, the function definition  $f\ xs\ @\ r = xs\ @\ r$  accepts the types  $[Int]@_{\rho_1} \rightarrow \rho_2 \rightarrow [Int]@_{\rho_2}$  and  $BSTree\ \alpha@_{\rho_1} \rightarrow \rho_2 \rightarrow BSTree\ \alpha@_{\rho_2}$ , but both of them are instances of the type  $\alpha \rightarrow \rho_2 \rightarrow \alpha$ , which is not accepted by the  $\vdash_{Reg}$  type system, since the  $[COPY_{Reg}]$  rule demands an algebraic type in the DS being copied. This lack of principal types can be solved by incorporating a restricted form of polymorphic variable in which the outermost RTV is fixed. For instance, the type  $\alpha@_{\rho}$  stands for an algebraic type whose outermost region is  $\rho$ . In this case, the  $f$  function would have  $\alpha@_{\rho_1} \rightarrow \rho_2 \rightarrow \alpha@_{\rho_2}$  as its most general type. As a short-term future work we plan to study the integration of this extension.

The pioneer work on region inference is that of Tofte, Talpin, and their colleagues on the MLKit compiler [112] (in what follows, TT). They address a more general problem than we do, since their language is higher-order. The TT algorithm has two phases, respectively called  $\mathcal{S}$  and  $\mathcal{R}$ . The  $\mathcal{S}$  algorithm just generates fresh region variables for values and introduces the lexical scope of the regions by using a **letregion** construct. The  $\mathcal{R}$  algorithm is responsible for assigning types to recursive functions. It deals with region-polymorphic recursion, and computes a fixed point. Both algorithms use higher-order unification based on directed graphs and on a *UNION-FIND* data structure. If  $n$  is the size of the HM-typed term being inferred, the  $\mathcal{S}$  algorithm runs in  $\mathcal{O}(n^3)$  time, and the  $\mathcal{R}$  algorithm in  $\mathcal{O}(n^4)$  time in the worst case. The total cost is in  $\mathcal{O}(n^4)$ . The meaning of a typed expression **letregion**  $\rho$  in  $e : \mu$  is that region  $\rho$  does not occur free in type  $\mu$ , so it can be deallocated upon the evaluation of  $e$ . Our algorithm has some resemblances with this part of the inference, in the sense that we decide to unify with  $\rho_{self}$  all the region variables not occurring in the result type of a function. They do not claim their algorithm to be optimal, but, in fact, they create as many regions as possible, trying to make local *all* the regions not needed in the final value. Incompleteness in TT comes as a consequence of polymorphic recursion. In our case, it is a consequence of the copy expressions.

The implementation of MLKit in 1998 added a third phase, called *storage mode analysis*, which introduced a *region resetting* action previously to some allocations. In order to take profit of this analysis, the programmer has to introduce *copy* expressions in specific parts of the text. A further work [2] added a constraint and control flow-based analysis after the TT inference. These resulted in modifying the text by delaying as much as possible some region allocations, and by bringing forward some deallocations, without compromising pointer safety. Again, the programmer needs to introduce the *copy* function in

appropriate places, in order to avoid memory leaks. From the programmer’s point-of-view, it is far from trivial there to place these *copy* functions. In *Safe*, the programmer can avoid such leaks via explicit deallocation, or by augmenting the number of nested calls in order to make the program more “region-friendly”, as it was done in the compiler of Example 4.22.

One problem reported in [114] is that most of the regions inferred in the first versions of the algorithm contained a single value, so region management produced a big overhead at runtime. Later, they added a new analysis to collapse all these regions into a single one local to the invocation (allocated in the stack). In our case, having a single local region *self* per function invocation does not seem to us to be a big drawback if function bodies are small enough. We believe that region-polymorphic recursion has a much bigger impact in avoiding memory leaks than multiplicity of local regions. So, we claim that the results of our algorithm are comparable to those of TT for first-order programs.

A radical deviation from these approaches is [54], which introduces a type system in which region life-times are not necessarily nested. The compiler annotates the program with region variables and supports operations for allocation, releasing, aliasing and renaming. A reference-counting analysis is used in order to decide when a released region should be deallocated. The language is first-order. The inference algorithm [75] can be defined as a global abstract interpretation of the program by following the control flow of the functions in a backwards direction. Although the authors do not give either asymptotic costs or actual benchmarks, it can be deduced that this cost could grow more than quadratically with the program text size in the worst case, as a global fixed point must be computed and a region variable may disappear at each iteration. This lack of modularity could make the approach impractical for large programs.

The main virtue of our design is its simplicity. This comes at the cost of obtaining less granularity, since the above mentioned systems allow to create several nested regions in the same function call. Also, in the case of TT and its derivatives, they support higher-order functions. As a consequence, the inference algorithms are more complex and costly. The techniques shown in this chapter could also be applied to any first-order functional language featuring Hindley-Milner types, since it can be implemented as a slight extension of Hindley-Milner inference. In the near future we plan to extend *Safe* to support higher-order functions and mutually recursive data structures (see Chapter 8). We expect high difficulties in other aspects of the language such as extending dangling pointers safety analyses or memory bounds inference, but not so many to extend the region inference algorithm presented here. However, it is still open whether a higher-order variant of this algorithm could achieve a cost better than the  $\mathcal{O}(n^4)$  got by Tofte and Talpin.

## Chapter 5

# Safe types inference

### 5.1 Introduction

In this chapter we continue addressing the problem of type reconstruction for the type system of Chapter 3. While the algorithm of the last chapter computes a  $\vdash_{Reg}$  derivation for a given program, in this chapter we aim to design an algorithm for finding  $\vdash_{Dst}$  derivations. We are no longer concerned about the regions in which a DS lives, but only about whether this DS is safe, condemned, or in-danger. Unlike the region inference algorithm, which annotates the source program with region variables, the algorithm of this chapter leaves the source program untouched: it only checks whether the destructive pattern matching, specified by the programmer, is done in a safe way. A program being accepted by both algorithms is guaranteed to be pointer safe, as explained in Chapter 3.

The inference of safe, condemned, and in-danger marks (*mark inference*, in the following) for finding a  $\vdash_{Dst}$  typing derivation is not a trivial task: the  $\vdash_{Dst}$  rules are not syntax-directed, and there may be several ways in which a rule can be applied. For instance, in a **let** expression, there are many ways to split the typing environment into the two environments typing its corresponding sub-expressions. The idea of the mark inference algorithm is to do a bottom-up traversal of the abstract syntax tree of the expression being analysed. As this traversal is done, those variables in the scope of the expression being analysed are classified into the set of safe, condemned, and in-danger variables. If more than one mark is applicable to a given variable, the algorithm gives precedence to safe types.

This paper extends the work of [81], whose contribution is a mark inference algorithm for the type system of [84]. Since in Chapter 3 we have introduced an improved version of the latter, in this Chapter we improve the mark inference algorithm of [81] in order to deal with the type system of this thesis. The main advantages of the algorithm of this chapter with respect to that of [81] are the following:

- It is more efficient. Given a function definition of  $n$  parameters, if  $s$  denotes the size of its abstract syntax tree, the worst-case time complexity of the algorithm in this chapter is  $\mathcal{O}(ns^2)$ , whereas the algorithm of [81] runs in  $\mathcal{O}(ns^3)$  time.
- It is complete. The  $\vdash_{Dst}$  rules admit minimal mark signatures for function definitions, and the algorithm is able to find these minimal signatures.
- It is simpler. It consists in a single bottom-up traversal of the abstract syntax tree, whereas the algorithm of [81] combines bottom-up and top-down traversals, which makes its correctness proof considerably more difficult.

In Section 5.2 we explain the inference algorithm in detail. We start from the mark inference of isolated expressions. Then we continue with the inference of function definitions, and, finally, we describe the inference of *Core-Safe* programs. Section 5.3 applies the algorithm to several case studies. In Section 5.4 we prove the algorithm correct and complete with respect to the  $\vdash_{Dst}$  rules. Finally, Section 5.5 concludes.

## 5.2 Mark inference algorithm

The inference algorithm is modular; each function definition is inferred separately. For each function of  $n$  parameters a mark signature  $\overline{m}_i^n \rightarrow s$  is inferred. Since the body of a function definition may also contain function applications, there exists an environment  $\Sigma$  containing the mark signatures of the already inferred functions. The inference of a program starts with an empty environment, and adds the mark signature of the function definitions as they are processed by the algorithm.

The inference of a function definition consists in a bottom-up traversal of its abstract syntax tree. For each sub-expression  $e$ , the algorithm tries to find an environment  $\Gamma$  such that the judgement  $\Gamma \vdash_{Dst} e : s$  is derivable. The  $\Gamma$  environment must contain the marks of every free variable of  $e$ , so the algorithm assigns a mark to the variables it comes across and stores that mark in the mark environment of the sub-expression being processed. There are several situations in which more than one mark may be applicable. For instance, if the expression being inferred is a variable  $x$ , the environment could contain the binding  $[x : s]$ ,  $[x : d]$ , or  $[x : r]$ , since  $x$  is typable under all of these environments. In a similar way, the discriminant of a **case!** may get a  $d$  or an  $r$  mark in the context of that expression. In these cases, the algorithm assigns the lowest mark with respect the following order:  $s \leq d \leq r$ . Thus, in the case of the variable  $x$ , it would get a safe mark, whereas in the case of the destructive **case!**, the discriminant would get a condemned mark. The rationale behind this decision is that the lowest mark is the less restrictive when considering the expression  $e$  as a sub-expression in a higher context. Let us explain this with an example:

**Example 5.1.** Given the following *Core-Safe* fragment:

$$\mathbf{let } y = x \mathbf{ in case } x \mathbf{ of } \dots$$

The  $x$  variable in the bound expression may get a safe, condemned, or in-danger mark. However, if it got a condemned or in-danger mark, the **let** expression would not be typeable with the  $[LET_{Dst}]$  rule, since there would be a variable with an unsafe mark in the bound expression of a **let** occurring free in its main expression.  $\square$

An additional advantage of assigning the lowest possible mark at a given context, is the fact that the  $[WEAK_{Dst}]$  rule allows the algorithm to rise that mark in a higher context, when necessary, but there is no rule in the type system for lowering marks.

### 5.2.1 Inference rules for expressions

The main part of the algorithm is a set of syntax-directed rules that allows us to derive judgements of the form  $e \vdash (R, D, S)$ , where  $R$ ,  $D$  and  $S$  are sets of variables. These sets specify which variables get an in-danger ( $r$ ), condemned ( $d$ ) and safe ( $s$ ) mark in the *minimal*  $\Gamma$  environment typing (w.r.t. the  $\vdash_{Dst}$  rules) the expression  $e$ . Every free variable in  $e$  occurs in one of these three sets, and their union is a subset of the variables in  $scope(e)$ . The inference rules define the  $(R, D, S)$  sets associated with an expression in

$$\begin{array}{c}
\frac{}{c \vdash (\emptyset, \emptyset, \emptyset)} [LIT_I] \quad \frac{}{x \vdash (\emptyset, \emptyset, \{x\})} [VAR_I] \quad \frac{}{x @ r \vdash (\emptyset, \emptyset, \{x\})} [COPY_I] \\
\\
\frac{}{C \bar{a}_i^n @ r \vdash (\emptyset, \emptyset, \{\bar{a}_i^n\})} [CONS_I] \\
\\
\begin{array}{l}
\Sigma(f) = \bar{m}_i^n \rightarrow s \\
S = \{a_i \mid m_i = s\} \\
R = \bigcup_{m_i=d} \text{sharerec}(a_i, f \bar{a}_i^n @ \bar{r}_j^m) \setminus \{a_i\} \\
\forall i \in \{1..n\}. D_i = \begin{cases} \{a_i\} & \text{if } m_i = d \\ \emptyset & \text{otherwise} \end{cases} \\
D = \bigcup_{i=1}^n D_i
\end{array}
\begin{array}{l}
R \cap S = \emptyset \\
R \cap D = \emptyset \\
S \cap D = \emptyset \\
\forall i, j \in \{1..n\}. i \neq j \Rightarrow D_i \cap D_j = \emptyset \\
\forall x \in D. \text{isTree}(x)
\end{array} \\
\hline
f \bar{a}_i^n @ \bar{r}_j^m \vdash (R, D, S) \quad [APP_I] \\
\\
\begin{array}{l}
e_1 \vdash (R_1, D_1, S_1) \\
e_2 \vdash (R_2, D_2, S_2)
\end{array}
\begin{array}{l}
(R, D, S) = (R_1, D_1, S_1) \sqcup ((R_2, D_2, S_2) \setminus \{x_1\}) \\
(R_1 \cup D_1) \cap \text{fv}(e_2) = \emptyset
\end{array} \\
\hline
\text{let } x_1 = e_1 \text{ in } e_2 \vdash (R, D, S) \quad [LET_I] \\
\\
\begin{array}{l}
\forall i \in \{1..n\}. P_i = \{\bar{x}_{ij}^n\} \\
(R, D, S) = (\bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i)) \sqcup (\emptyset, \emptyset, \{x\})
\end{array}
\begin{array}{l}
\forall i \in \{1..n\}. e_i \vdash (R_i, D_i, S_i)
\end{array} \\
\hline
\text{case } x \text{ of } \overline{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n \vdash (R, D, S) \quad [CASE_I] \\
\\
\begin{array}{l}
\forall i \in \{1..n\}. e_i \vdash (R_i, D_i, S_i) \\
\forall i \in \{1..n\}. P_i = \{\bar{x}_{ij}^n\} \\
\forall i \in \{1..n\}. \text{Rec}_i = \{x_{ij} \mid j \in \text{RecPos}(C_i)\} \\
\forall i \in \{1..n\}. (P_i \setminus \text{Rec}_i) \cap (D_i \cup R_i) = \emptyset \\
\forall i \in \{1..n\}. \text{Rec}_i \cap R_i = \emptyset
\end{array}
\begin{array}{l}
(R, D, S) = \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i) \\
(R', D', S') = ((R, D, S) \setminus \{x\}) \sqcup (R_{SHR}, \emptyset, \emptyset) \\
R_{SHR} = \text{sharerec}(x, \text{case! } x \text{ of } \overline{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n) \setminus \{x\} \\
\forall i \in \{1..n\}. (R_{SHR} \cup \{x\}) \cap \text{fv}(e_i) = \emptyset
\end{array} \\
\hline
\text{case! } x \text{ of } \overline{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n \vdash (R', D' \cup \{x\}, S') \quad [CASE!_I]
\end{array}$$

Figure 5.1: Inference rules for expressions.

terms of the triples obtained for each sub-expression. Thus these rules can be implemented by means of a bottom-up traversal of the abstract syntax tree of the expression. We assume the existence of a mark environment  $\Sigma : \mathbf{Fun} \rightarrow \mathbf{MarkSig}$  that it is implicitly propagated through the  $\vdash$  rules, and contains the mark signatures of the functions being called from the corresponding expression. In case of ambiguity, we shall make this signature explicit with the notation  $e \vdash_{\Sigma} (R, D, S)$ .

Figure 5.1 shows the inference rules. Rules  $[\mathbf{VAR}_I]$ ,  $[\mathbf{COPY}_I]$ , and  $[\mathbf{CONS}_I]$  assign an s mark to the variables in the corresponding expression. By abuse of notation, we assume that the set  $\{\bar{a}_i^n\}$  only contains the atoms  $a_i$  that are not literals. In the  $[\mathbf{APP}_I]$  rule, the mark signature of the function being called determines the marks of the arguments. Those occurring in safe positions get a safe mark, and those occurring in condemned positions get a condemned mark. The disjointness condition between the different  $D_i$  sets, together with the condition  $S \cap D = \emptyset$  mimic the well-definedness property of the  $\oplus$  operator in the typing rules: the same variable can occur in two positions only if both are safe. The conditions  $R \cap D = \emptyset$  and  $R \cap S = \emptyset$  correspond to the disjoint union of the  $\Gamma_R$  environment with the  $\bigoplus_{i=1}^n [a_i : m_i]$  environment.

In the  $[\mathbf{LET}_I]$ , the triples of both auxiliary and main expressions  $e_1$  and  $e_2$  are inferred. The notation  $(R, D, S) \setminus M$  abbreviates the triple  $(R \setminus M, D \setminus M, S \setminus M)$ . The  $\sqcup$  operator, which is analogous to the  $\sqcup$  operator on typing environments, is defined as follows:

$$(R_1, D_1, S_1) \sqcup (R_2, D_2, S_2) = (R_1 \cup R_2, (D_1 \cup D_2) \setminus (R_1 \cup R_2), (S_1 \cup S_2) \setminus (D_1 \cup D_2 \cup R_1 \cup R_2))$$

It is easy to prove that the  $\sqcup$  operator is commutative and associative. The generalization of this operator to  $n$  triples can be expressed as follows:

$$\bigsqcup_{i=1}^n (R_i, D_i, S_i) = \left( \bigcup_{i=1}^n R_i, \bigcup_{i=1}^n D_i \setminus \left( \bigcup_{i=1}^n R_i \right), \bigcup_{i=1}^n S_i \setminus \left( \bigcup_{i=1}^n R_i \cup \bigcup_{i=1}^n D_i \right) \right) \quad (5.1)$$

The side condition  $(R_1 \cup D_1) \cap fv(e_2) = \emptyset$  is equivalent to the side condition of  $[\mathbf{LET}_{Dst}]$ . It ensures that no variable is mentioned in the main expression if a part of it has been destroyed previously.

In the  $[\mathbf{CASE}_I]$  and  $[\mathbf{CASE!}_I]$  rules we infer the triple of each alternative separately. This implies that the same variable may get different marks in different **case** branches. The  $\sqcup$  operator ensures that this variable gets in the whole **case**(!) expression the lowest (most general) mark common to all the  $e_i$ . In terms of typing rules, it represents the least common environment (excluding pattern variables) typing every branch. The least upper bound with  $(\emptyset, \emptyset, \{x\})$  obeys the fact that every free variable must belong to one of the sets of condemned, in-danger and safe marks. Thus it is assigned a safe (s) mark if it does not occur in any triple  $(R_i, D_i, S_i)$ . The definition of  $R_{SHR}$  in  $[\mathbf{CASE!}_I]$ , and its associated side condition, are analogous to the  $R$  set occurring in  $[\mathbf{CASE!}_{Dst}]$ . The remaining side conditions in  $[\mathbf{CASE!}_I]$  mimic the *inh* predicate. On the one hand,  $(P_i \setminus Rec_i) \cap (D_i \cup R_i) = \emptyset$  demands those pattern variables occurring in non-recursive positions to have a safe mark, or no mark at all, in every branch. On the other hand,  $Rec_i \cap R_i = \emptyset$  specifies that, in every branch, recursive pattern variables must have a safe or condemned mark, or remain unmarked in every branch.

## 5.2.2 Inference of function definitions

When inferring non-recursive function definitions we only have to apply the rules of Figure 5.1 to the body of the function once. We assume that functions being called from the current one have already been inferred, and their mark signatures are already available in  $\Sigma$ . Once we obtain the triple  $(R, D, S)$

$$\begin{aligned}
inferMarksDef \Sigma (f \bar{x}_i^n @ \bar{r}_j^m = e_f) &= inferMarksFP (\Sigma \uplus [f \mapsto (\bar{s}^n \rightarrow s)]) (f \bar{x}_i^n @ \bar{r}_j^m = e_f) \\
inferMarksFP \Sigma (f \bar{x}_i^n @ \bar{r}_j^m = e_f) & \\
\quad | \quad R \neq \emptyset &= error \\
\quad | \quad R = \emptyset \wedge \Sigma(f) = \bar{m}_i^n \rightarrow s &= \Sigma \\
\quad | \quad R = \emptyset \wedge \Sigma(f) \neq \bar{m}_i^n \rightarrow s &= inferMarksFP ((\Sigma \setminus f) \uplus [f \mapsto (\bar{m}_i^n \rightarrow s)]) (f \bar{x}_i^n @ \bar{r}_j^m = e_f) \\
\text{where } e_f \vdash_{\Sigma} (R, D, S) & \\
\quad \forall i \in \{1..n\}. m_i &= \begin{cases} d & \text{if } x_i \in D \\ s & \text{if } x_i \notin D \end{cases}
\end{aligned}$$

Figure 5.2: Inference algorithm applied to function definitions.

for the function, we can reconstruct easily the mark signature of the function as follows: if there is a parameter in the  $R$  set, the function is not typeable. If a parameter belongs to the  $D$  set (resp. the  $S$  set), it gets a condemned mark  $d$  (resp. safe mark  $s$ ). If it does not belong to any of the three sets, the parameter may have a safe or a condemned mark. Since our goal is to get the lowest type, it is sensible to assign a safe mark  $s$  to these parameters.

The inference of recursive function definitions is slightly more involved, because we do not know in advance which mark signature should be assigned to the function being inferred. Since we aim for the most general mark signature, we start assuming that every parameter of the function has a safe mark and apply the rules of Figure 5.1. If every parameter turns out to be safe, our assumption was correct and we return the corresponding updated signature. If some parameter gets a condemned mark, our assumption was wrong and we have to update the signature  $\Sigma$  according to the triple  $(R, D, S)$  obtained for the function's body. Parameters belonging to  $D$  now get a condemned mark, and those in  $S$  (or not occurring in any of the sets) get a safe mark. With these new assumptions (reflected in the updated signature  $\Sigma$ ) we apply the inference rules again. We iterate this process until a fixpoint is reached, that is, the obtained triple  $(R, D, S)$  corresponds to the assumption under which it is inferred.

These ideas are formalized in the algorithm of Figure 5.2. The *inferMarks* function is given a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  and a signature environment  $\Sigma$  containing the mark signatures of all the functions being called from  $f$  (except  $f$  itself). This function does the first call to *inferMarksFP* with the initial assumption, in which every parameter has a safe mark. The *inferMarksFP* function does the fixed point iteration. In every iteration, it applies the  $\vdash$  rules of Figure 5.1 and obtains a triple  $(R, D, S)$ . If there is an element in  $R$ , then it is one of the parameters, since  $R$  only contains variables in scope. In this case, the inference algorithm returns an error, since only safe and condemned types are allowed in function signatures (see Section 3.3 for a discussion on this). If  $R$  is empty, we build a mark signature  $\bar{m}_i^n \rightarrow s$  which assigns a  $d$  mark to those parameters in  $D$  and an  $s$  mark to those parameters not occurring in  $D$ . If the new mark signature matches the one already existing in  $\Sigma$ , a fixed point has been reached and we return  $\Sigma$ . If it does not match, the function updates the signature with the new information and starts another iteration.

**Example 5.2.** Let us consider the *reverseD* function that reverses the list given as input. It uses an auxiliary function *revAuxD* that destroys its first parameter, while accumulating the result in its second

$$\Sigma = [\text{revAuxD} \mapsto (s \rightarrow s \rightarrow s)]$$

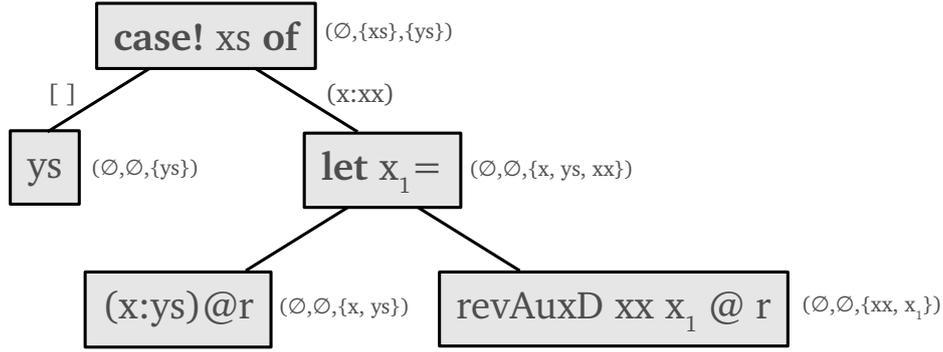


Figure 5.3: Sets inferred after the first iteration of *revAuxD*. The  $(R, D, S)$  triple computed for each subexpression is shown to the right of its gray box.

$$\Sigma = [\text{revAuxD} \mapsto (d \rightarrow s \rightarrow s)]$$

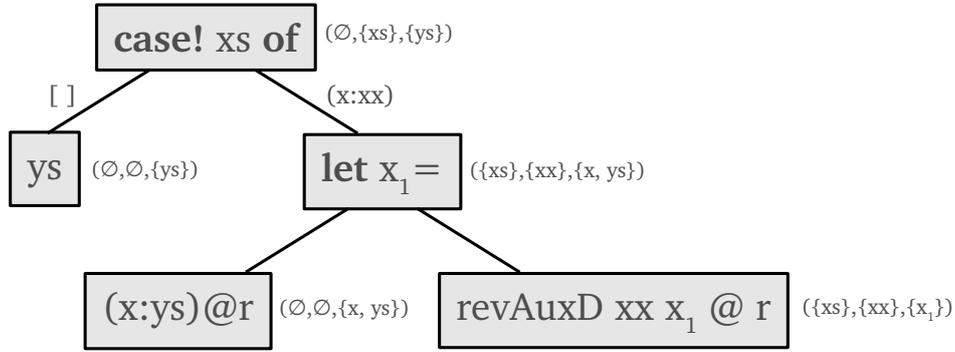


Figure 5.4: Sets inferred after the second iteration of *revAuxD*.

parameter.

$$\begin{aligned} \text{revAuxD } xs \text{ } ys \text{ } @ r & \\ &= \text{case! } xs \text{ of} \\ &\quad [] \rightarrow ys \\ &\quad (x : xx) \rightarrow \text{let } x_1 = (x : ys)@r \text{ in } \text{revAuxD } xx \text{ } x_1 \text{ } @ r \end{aligned}$$

$$\text{reverseD } xs \text{ } @ r = \text{let } x_1 = [] @ r \text{ in } \text{revAuxD } xs \text{ } x_1 \text{ } @ r$$

We start applying the inference rules to the body of *revAuxD*. We assume that the mark environment  $\Sigma$  contains the following signature for *revAuxD*:  $s \rightarrow s \rightarrow s$ . The results inferred for each sub-expression are shown in Figure 5.3. Notice that *xx* and  $x_1$  are assigned safe marks, as a consequence of the initial environment  $\Sigma$  assigning a safe type to every parameter. At the end of the iteration we obtain the triple  $(\emptyset, \{xs\}, \{ys\})$  for the whole body of *revAuxD*, which corresponds to the signature  $d \rightarrow s \rightarrow s$ . Since this signature is different from the one initially assumed, we update the signature environment with the new signature, and apply again the inference rules. The result is shown in Figure 5.4. A difference with respect to the first iteration is the triple of sets inferred in the recursive call:  $(\{xs\}, \{xx\}, \{x_1\})$ . The *xx* variable is condemned, since it occurs in a condemned position. Since *xs* always points to this

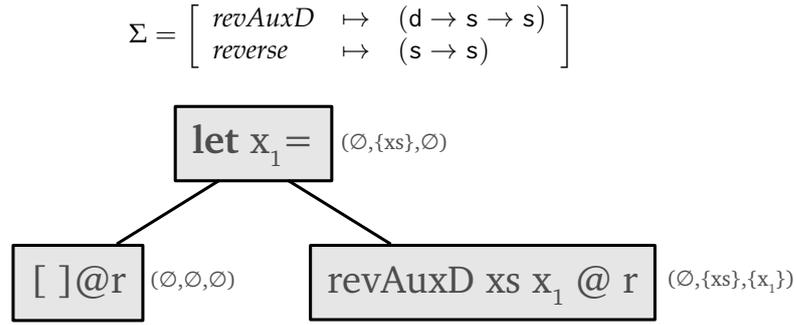


Figure 5.5: Sets inferred after the first iteration of *reverse*.

$$\text{inferMarksProg } (\overline{\text{data}}_i; \overline{\text{def}}_i; e) = \text{if } e \vdash_{\Sigma'} (\emptyset, \emptyset, \emptyset) \text{ then OK else error}$$

$$\text{where } \Sigma' = \text{foldl inferMarksDef } [] \overline{\text{def}}_i$$

Figure 5.6: Inference algorithm applied to *Core-Safe* programs.

variable, it must occur in  $\text{sharerec}(xx, \text{revAuxD } xx \ x_1 \ @ \ r)$ , and is assigned an in-danger mark. However, this mark is turned into a condemned mark as a consequence of the **case!** in the outermost expression. As a result we obtain, again, the mark signature  $d \rightarrow s \rightarrow s$ , so we have reached a fixed point. The algorithm stores the binding  $[\text{revAuxD} \mapsto (d \rightarrow s \rightarrow s)]$  in the signature environment  $\Sigma$  and proceeds with the following definition in the program.

Figure 5.5 shows the  $(R, D, S)$  tuples inferred by the algorithm for each sub-expression of *reverse*, assuming  $(s \rightarrow s)$  as its mark signature. From the inference rules we get that  $xs$  gets a condemned mark, which results in the signature  $(d \rightarrow s)$ . In the second iteration it turns out we have reached a fixed point. The last iteration is unnecessary, as, in this case, the signature of  $\Sigma(\text{reverse})$  is not relevant to the resulting  $(R, D, S)$  tuples. This happens, in general, when inferring the mark signatures of non-recursive function definitions.  $\square$

### 5.2.3 Inference of a *Core-Safe* program

The *inferMarksProg* function shown in Figure 5.6 receives a *Core-Safe* program, and determines whether it is well-typed with respect to the  $\vdash_{Dst}$  rules. Firstly, it builds a signature environment  $\Sigma'$  in an incremental way, as function definitions are analysed by the *inferMarksDef* function. After this, it checks whether the judgement  $e \vdash (\emptyset, \emptyset, \emptyset)$  is derivable under this environment  $\Sigma'$ . The three components of the resulting triple must be empty sets, since in the context of the main expression of a program there are no variables in scope.

## 5.3 Case studies

In this section we show how the inference algorithm behaves with several case studies. For the sake of simplicity, we show only the *Full-Safe* code of the programs being analysed, and we show only the results of the algorithm, without any mention to the derivation of the  $\vdash$  judgements.

**Example 5.3.** A *quadtree* is a tree data structure in which each node has four children. It has two different

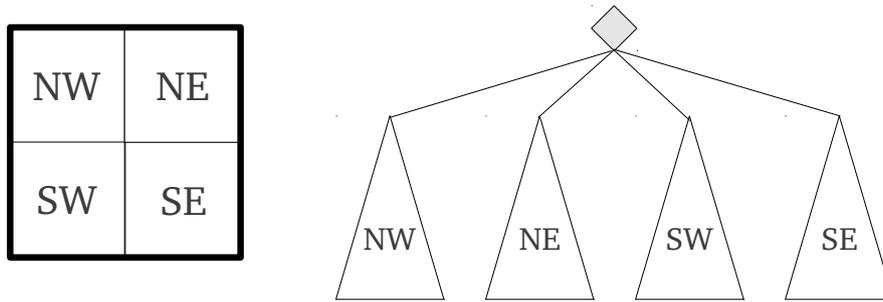


Figure 5.7: Quadtree representation of a bitmap. The whole area is divided into four quadrants. Each quadrant is represented by its own quadtree.

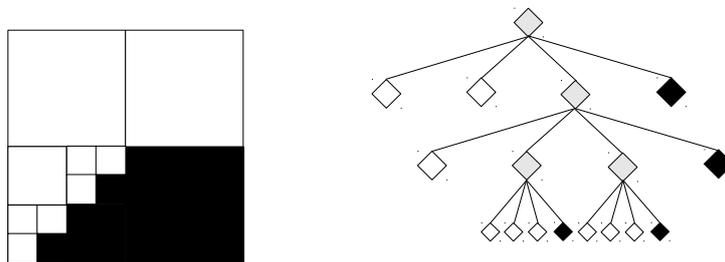


Figure 5.8: Bitmap and its representation as a quadtree.

kinds of leaves: black leaves and white leaves. It is defined as follows:

```
data QuadTree = Black | White | Node QuadTree QuadTree QuadTree QuadTree
```

In some situations, quadtrees provide an efficient representation of bitmaps by following a divide-and-conquer strategy. If the bitmap contains only black pixels, it is represented by *Black*. If it contains only white pixels, it is represented by *White*. Otherwise, we divide the bitmap into four quadrants, each of which is represented by its corresponding quadtree. The bitmap is then represented by the *Node* constructor applied to the bitmaps of these quadrants (Figure 5.7). Figure 5.8 shows an example of a bitmap and its quadtree representation.

The *rotateD* function performs a ninety-degree counterclockwise rotation of the bitmap given as parameter (see Figure 5.9). The original bitmap is destroyed, so this function does not need additional heap space:

```
rotateD Black! = Black
rotateD White! = White
rotateD (Node nw ne sw se)! =
  Node (rotateD ne) (rotateD se) (rotateD nw) (rotateD sw)
```

This function is successfully typed by the algorithm, which infers the following mark signature for *rotateD*:  $d \rightarrow s$ . The *flipHD* function, which mirrors the input image horizontally, works in a similar

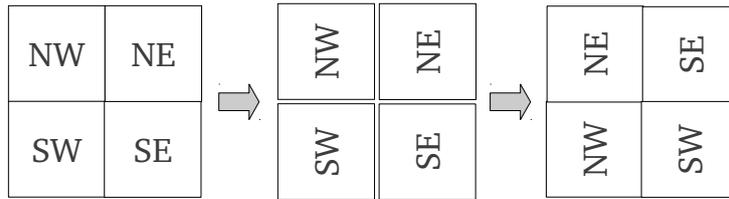


Figure 5.9: Rotation of a bitmap. Firstly, each piece is rotated individually. Then, a global rotation of the four pieces is done.

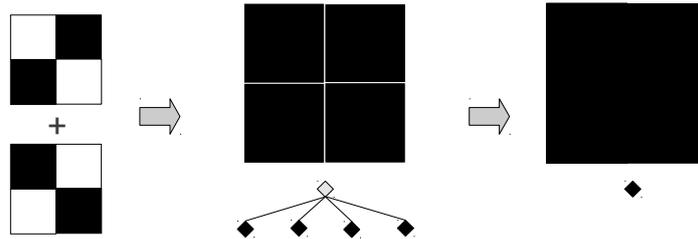


Figure 5.10: When overlying two bitmaps, a simplification of the quadtree may be needed.

fashion:

$$\begin{aligned}
 \text{flipHD Black!} &= \text{Black} \\
 \text{flipHD White!} &= \text{White} \\
 \text{flipHD (Node nw ne sw se)!} &= \\
 &\text{Node (flipHD ne) (flipHD nw) (flipHD se) (flipHD sw)}
 \end{aligned}$$

The inference algorithm infers the signature  $d \rightarrow s$  for this function. Finally, we define a function *overlayD* which merges the black pixels of the two bitmaps given as parameters. Both input pictures are destroyed.

$$\begin{aligned}
 \text{overlayD White! } q! &= q \\
 \text{overlayD Black! } q! &= \text{Black} \\
 \text{overlayD (Node nw ne sw se)! White!} &= \text{Node nw ne sw se} \\
 \text{overlayD (Node nw ne sw se)! Black!} &= \text{Black} \\
 \text{overlayD (Node nw}_1 \text{ ne}_1 \text{ sw}_1 \text{ se}_1 \text{)! (Node nw}_2 \text{ ne}_2 \text{ sw}_2 \text{ se}_2 \text{)!} \\
 &= \text{Node (overlayD nw}_1 \text{ nw}_2) \text{ (overlayD ne}_1 \text{ ne}_2) \text{ (overlayD sw}_1 \text{ sw}_2) \text{ (overlayD se}_1 \text{ se}_2)
 \end{aligned}$$

However, this solution is not as optimal as one would expect, since it might give place to internal nodes whose leaves are all of the same colour (see Figure 5.10). In these cases, a single node of that colour needs less heap space. We can replace the call to the constructor by a wrapper that performs all the necessary simplifications:

$$\begin{aligned}
 \text{overlayD (Node nw}_1 \text{ ne}_1 \text{ sw}_1 \text{ se}_1 \text{)! (Node nw}_2 \text{ ne}_2 \text{ sw}_2 \text{ se}_2 \text{)!} \\
 = \text{buildNode (overlayD nw}_1 \text{ nw}_2) \text{ (overlayD ne}_1 \text{ ne}_2) \text{ (overlayD sw}_1 \text{ sw}_2) \text{ (overlayD se}_1 \text{ se}_2)
 \end{aligned}$$

So, let us define *buildNode*. A first attempt on implementing this function is as follows:

```

buildNode Black! Black! Black! Black! = Black
buildNode White! White! White! White! = White
buildNode nw! ne! sw! se! = Node nw ne sw se

```

This function is rejected by the inference algorithm, since it is not correct. When the execution reaches the right-hand side of the third equation it accesses the *nw*, *ne*, *sw*, *se* variables, which have been already destroyed by the pattern matching. In principle we would have to access to the children of each variable, and rebuild the corresponding DSs in the result, as done, for example, in the third equation of *overlayD*. However, that would result in a considerable number of equations, since we would have to distinguish each possible combination of constructors for the four parameters. The following definition, though not very elegant, circumvents this problem:

```

buildNode nw ne sw se
| isBlack nw  $\wedge$  isBlack ne  $\wedge$  isBlack sw  $\wedge$  isBlack se =
    let  $x_1$  = destroyBlackNodes nw ne sw se in Black
| isWhite nw  $\wedge$  isWhite ne  $\wedge$  isWhite sw  $\wedge$  isWhite se
    let  $x_1$  = destroyWhiteNodes nw ne sw se in White
| otherwise = Node nw ne sw se

```

where the *isXXX* and *destroyXXXNodes* functions are defined as follows:

```

isBlack Black = True
isBlack _ = False

isWhite White = True
isWhite _ = False

destroyBlackNodes Black! Black! Black! Black! = 0

destroyWhiteNodes White! White! White! White! = 0

```

The *destroyXXXNodes* functions produce only the side-effect of destroying the leaves passed as parameters, but their return value is meaningless, so it is not used in *buildNode*. The inference algorithm accepts all these functions, and returns the following mark signatures:

```

isBlack :: s  $\rightarrow$  s
isWhite :: s  $\rightarrow$  s
destroyBlackNodes :: d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  s
destroyWhiteNodes :: d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  s
buildNode :: d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  d  $\rightarrow$  s
overlayD :: d  $\rightarrow$  d  $\rightarrow$  s

```

□

**Example 5.4.** Let us recall the implementation of priority queues shown in Example 3.13. All functions are accepted by the inference algorithm, which returns the following signatures:

$$\begin{array}{ll}
\mathit{cons} & :: d \rightarrow s \rightarrow d \rightarrow s \\
\mathit{join} & :: d \rightarrow d \rightarrow s \\
\mathit{emptyPQueue} & :: s \\
\mathit{addPQueue} & :: d \rightarrow s \rightarrow s \\
\mathit{minPQueue} & :: s \rightarrow s \\
\mathit{delMinPQueue} & :: d \rightarrow s
\end{array}$$

The first two functions require two iterations of the  $\vdash$  rules in order to reach a fixed point. The remaining ones are non-recursive, and hence require only one iteration.  $\square$

**Example 5.5.** Figure 5.11 shows an implementation of the *mergesort* algorithm for sorting a list of integers. The *splitD* function is given a natural number  $n$ , and a list  $xs$ . It returns a pair with the  $n$  first elements of the list in the first component, and the remaining ones in the second component. It also destroys the list given as input. The *mergeD* function coalesces two sorted lists into a single ordered list. It is destructive on its two parameters. As a consequence, the execution of both functions can be done in constant space. Both functions are accepted by the algorithm, which results in the following signatures:

$$\begin{array}{ll}
\mathit{splitD} & :: s \rightarrow d \rightarrow s \\
\mathit{mergeD} & :: d \rightarrow d \rightarrow s
\end{array}$$

Notice that, in the first equation of *splitD*, the  $xs$  parameter occurs with a (!) mark, but its destruction does not take place until the pattern matching is done (second and third equations of *splitD*). That is why  $xs$  can be safely returned in the first equation. The same applies to *mergeD*. The *msortD* function, which sorts the input list in constant space, is not accepted by the algorithm. This is due to the information provided by the sharing analysis, which is not accurate enough to infer that the partial sorted lists (i.e. the results of *msortD*  $xs_1$  and *msortD*  $xs_2$ ) cannot share with each other. As a consequence, the algorithm fails in the function application of *mergeD*, since two mutually sharing lists cannot be passed as two different condemned parameters of *mergeD*. The current implementation of the algorithm allows the programmer to adjust the sharing information manually, in order to deal with these situations. Obviously, the safety of the program can no longer be guaranteed if the programmer modifies the information of the sharing analysis in a wrong way. In our case, it is easy to see that *msortD* builds its result from scratch, and consequently, cannot share its recursive spine with that of the input list. If we remove that over-approximating information, the function is accepted by the algorithm, and we get the following signature:

$$\mathit{msortD} :: d \rightarrow s$$

Another possibility is to modify the definition of *mergeD* so that it does not destroy any of the input lists. In this case the algorithm also results in the mark signature above. However, the memory costs of the algorithm are no longer constant.  $\square$

Figure 5.12 shows the mark signatures returned by the algorithm for some of the examples in last chapters. Some functions have been slightly modified, in order to include destructive pattern matching, instead of non-destructive, or to call the destructive version of a function, instead the non-destructive one. The last column indicates whether it was necessary to modify manually the results of the sharing analysis, in order to make the inference algorithm accept the function. Almost all the examples are accepted by the inference algorithm as is, except the *joinAVL* function, which required a manual refinement of the information given by the sharing analysis.

```

splitD 0 xs!      = ([], xs)
splitD n []!      = ([], [])
splitD n (x : xs)! = (x : xs1, xs2)
                where (xs1, xs2) = splitD (n - 1) xs

mergeD []!      ys!      = ys
mergeD (x : xs)! []!      = x : xs
mergeD (x : xs)! (y : ys)!
  | x ≤ y = x : mergeD xs (y : ys)
  | x > y = y : mergeD (x : xs) ys

msortD []! = []
msortD (x : xs)! = case! xs of
  [] → x : []
  (y : ys) → mergeD (msortD xs1) (msortD xs2)
  where (xs1, xs2) = splitD (n 'div' 2) (x : (y : ys))
        n = length (x : (y : ys))

```

Figure 5.11: Implementation of *mergesort* algorithm.

Function name	Mark signature	Adjust sharing?
appendD (Ex. 2.10, pg. 27)	$d \rightarrow s \rightarrow s$	No
insertD (Ex. 2.11, pg. 28)	$s \rightarrow d \rightarrow s$	No
inssort (Ex. 2.11, pg. 28)	$d \rightarrow s$	No
insertT (Ex. 4.21, pg. 156)	$s \rightarrow d \rightarrow s$	No
mkTree (Ex. 4.21, pg. 156)	$d \rightarrow s$	No
inorder (Ex. 4.21, pg. 156)	$d \rightarrow s$	No
treesort (Ex. 4.21, pg. 156)	$d \rightarrow s$	No
sJoinAVL (Ex. 3.14, pg. 94)	$s \rightarrow s \rightarrow s \rightarrow s$	No
lJoinAVL (Ex. 3.14, pg. 94)	$s \rightarrow s \rightarrow d \rightarrow s$	No
rJoinAVL (Ex. 3.14, pg. 94)	$d \rightarrow s \rightarrow s \rightarrow s$	No
joinAVL (Ex. 3.14, pg. 94)	$d \rightarrow s \rightarrow d \rightarrow s$	Yes
insertAVL (Ex. 3.14, pg. 94)	$s \rightarrow d \rightarrow s$	No
deleteAVL (Ex. 3.14, pg. 94)	$s \rightarrow d \rightarrow s$	No
partition (Ex. 4.19, pg. 152)	$s \rightarrow d \rightarrow s$	No
sumList (Ex. 4.20, pg. 153)	$d \rightarrow s$	No
pascal (Ex. 4.20, pg. 153)	$s \rightarrow s$	No

Figure 5.12: Results of the inference algorithm for some of the examples in this thesis.

## 5.4 Correctness, completeness, and efficiency

In this section we prove that every function accepted by the mark inference algorithm is typeable w.r.t. the  $\vdash_{Dst}$  rules, and that every typeable function w.r.t. the latter rules is accepted by the mark inference algorithm. In case there are several valid signatures for a given function, we explain which one is returned by the algorithm.

### 5.4.1 Correctness

Before getting in the correctness proof of the  $\vdash$  rules, let us introduce some of their invariants. In particular, we prove that a variable cannot get more than a mark in the context of a given expression. That is, the sets returned by the  $\vdash$  rules are pairwise disjoint. In addition, the  $D$  and  $S$  sets may only contain free variables in  $e$ .

**Lemma 5.6.** *Given an expression  $e$  such that  $e \vdash (R, D, S)$ :*

1.  $R, D$  and  $S$  are pairwise disjoint.
2.  $D \cup S \subseteq fv(e)$ .

*Proof.* By induction on the structure of  $e$ . Cases  $e \equiv c$ ,  $e \equiv x$ ,  $e \equiv x!$  and  $e \equiv C \bar{a}_i^n @ \bar{r}_j^m$  are straightforward. In the case of a function application, the lemma is a direct consequence of the side conditions of rule  $[APP_I]$ . With respect to the remaining cases, from (5.1) it directly follows that the  $\sqcup$  operator always returns mutually disjoint sets, so the first conclusion holds. With respect to the second one, we can prove from (5.1) that, if  $(R, D, S) = (R_1, D_1, S_1) \sqcup (R_2, D_2, S_2)$  then  $D \cup S = D_1 \cup S_1 \cup D_2 \cup S_2$ . Together with the fact that  $[LET_I]$  removes the bound variable from  $(R_2, D_2, S_2)$ , and that  $[CASE_I]$ ,  $[CASE!_I]$  removes the pattern variables from each  $(R_i, D_i, S_i)$ , the second conclusion follows trivially from the induction hypothesis applied to each sub-expression.  $\square$

With regard to the correctness of the inference rules, we are interested in proving the following fact:

*If an expression is accepted by the inference algorithm, it is  $\vdash_{Dst}$ -typeable.*

However, we can take advantage of the information provided by the  $\vdash$  rules (namely, the  $(R, D, S)$  sets) and be more specific on which mark environment  $\Gamma$  can be used to type the expression w.r.t. the  $\vdash_{Dst}$  rules: the variables in the  $R$  set (resp.  $D, S$ ) should occur with a  $r$  (resp.  $d, s$ ) mark in that environment. Hence, we aim to prove the following stronger property:

*If  $e \vdash (R, D, S)$  is derived by the rules of Figure 5.1, then  $e$  is  $\vdash_{Dst}$ -typeable under an environment assigning an in-danger mark to the variables in  $R$ , a condemned mark to those in  $D$ , and a safe mark to those in  $S$ .*

Before proving this, we need an auxiliary result with respect to the  $\sqcup$  operator. The definition of this operator, when applied to type environments, has been introduced in Section 3.3. The result of  $\Gamma_1 \sqcup \Gamma_2$ , when defined, is another type environment which combines the types of  $\Gamma_1$  and  $\Gamma_2$ , but giving precedence to in-danger marks over the remaining ones, and condemned marks over safe marks. In this chapter we have given a definition of  $\sqcup$  when applied to tuples  $(R, D, S)$  (see (5.1)). It is not surprising that this operator behaves in a similar way to its counterpart defined on type environments: the  $R$  set takes precedence over  $D$  and  $S$ , and the  $D$  takes precedence over  $S$ . This connection between the two meanings of  $\sqcup$  is formalized as follows.

**Lemma 5.7.** *If  $(R, D, S) = (R_1, D_1, S_1) \sqcup (R_2, D_2, S_2)$ , and the following environments are well-defined:*

$$\begin{aligned}\Gamma_1 &= [x : r \mid x \in R_1] + [x : d \mid x \in D_1] + [x : s \mid x \in S_1] \\ \Gamma_2 &= [x : r \mid x \in R_2] + [x : d \mid x \in D_2] + [x : s \mid x \in S_2] \\ \Gamma &= [x : r \mid x \in R] + [x : d \mid x \in D] + [x : s \mid x \in S]\end{aligned}$$

then  $\Gamma = \Gamma_1 \sqcup \Gamma_2$ .

*Proof.* First we prove that  $\text{dom } \Gamma = \text{dom } (\Gamma_1 \sqcup \Gamma_2)$ . This follows from  $R \cup D \cup S$  being equal to  $R_1 \cup R_2 \cup S_1 \cup S_2 \cup D_1 \cup D_2$ , which can be easily proven from the definition of the  $\sqcup$  operator when applied to triples  $(R, D, S)$ .

Now we prove that, for all  $x \in \text{dom } \Gamma$ ,  $(\Gamma_1 \sqcup \Gamma_2)(x) = r$  if and only if  $x \in R$ .

$$\begin{aligned}(\Gamma_1 \sqcup \Gamma_2)(x) &= r \\ \Leftrightarrow \Gamma_1(x) = r \vee \Gamma_2(x) = r \\ \Leftrightarrow x \in R_1 \vee x \in R_2 \\ \Leftrightarrow x \in R_1 \cup R_2 \\ \Leftrightarrow x \in R\end{aligned}$$

Now we prove  $(\Gamma_1 \sqcup \Gamma_2)(x) = d$  if and only if  $x \in D$ .

$$\begin{aligned}(\Gamma_1 \sqcup \Gamma_2)(x) &= d \\ \Leftrightarrow (\Gamma_1(x) = d \wedge (\Gamma_2(x) \neq r \vee x \notin \text{dom } \Gamma_2)) \vee (\Gamma_2(x) = d \wedge (\Gamma_1(x) \neq r \vee x \notin \text{dom } \Gamma_1)) \\ \Leftrightarrow (x \in D_1 \wedge (x \notin R_2 \vee x \notin R_2 \cup D_2 \cup S_2)) \vee (x \in D_2 \wedge (x \notin R_1 \vee x \notin R_1 \cup D_1 \cup S_1)) \\ \Leftrightarrow (x \in D_1 \wedge \neg(x \in R_2 \wedge x \in R_2 \cup D_2 \cup S_2)) \vee (x \in D_2 \wedge \neg(x \in R_1 \wedge x \in R_1 \cup D_1 \cup S_1)) \\ \Leftrightarrow (x \in D_1 \wedge x \notin (R_2 \cap (R_2 \cup D_2 \cup S_2))) \vee (x \in D_2 \wedge x \notin (R_1 \cap (R_1 \cup D_1 \cup S_1))) \\ \Leftrightarrow (x \in D_1 \wedge x \notin R_2) \vee (x \in D_2 \wedge x \notin R_1) \\ \Leftrightarrow (x \in D_1 \wedge x \notin R_1 \wedge x \notin R_2) \vee (x \in D_2 \wedge x \notin R_1 \wedge x \notin R_2) \\ \Leftrightarrow (x \in D_1 \vee x \in D_2) \wedge x \notin R_1 \wedge x \notin R_2 \\ \Leftrightarrow x \in D_1 \cup D_2 \wedge x \notin R_1 \cup R_2 \\ \Leftrightarrow x \in D\end{aligned}$$

Finally we prove  $(\Gamma_1 \sqcup \Gamma_2)(x) = s$  if and only if  $x \in S$ .

$$\begin{aligned}(\Gamma_1 \sqcup \Gamma_2)(x) &= s \\ \Leftrightarrow (\Gamma_1 \sqcup \Gamma_2)(x) \neq r \wedge (\Gamma_1 \sqcup \Gamma_2)(x) \neq d \\ \Leftrightarrow \Gamma(x) \neq r \wedge \Gamma(x) \neq d \\ \Leftrightarrow x \notin R \wedge x \notin D \\ \Leftrightarrow x \in S\end{aligned}$$

The last step follows because  $\Gamma_1 \sqcup \Gamma_2$  and  $\Gamma$  have the same domain: since  $x \in \text{dom } (\Gamma_1 \sqcup \Gamma_2)$ , we get  $x \in \text{dom } \Gamma = R \cup D \cup S$ , so if a variable does not belong to  $R \cup D$ , it must occur in  $S$ .  $\square$

Now we can prove the correctness of the  $\vdash$  rules when applied to an expression  $e$ . Recall that, although not explicitly indicated in most cases, these rules are parametric on an environment  $\Sigma$  which contains the mark signatures of the functions being called from  $e$ . This environment is part of the mark environment  $\Gamma$ , under which  $e$  is typed.

**Theorem 5.8.** *Let us assume that the algorithm infers  $e \vdash_{\Sigma} (R, D, S)$  under some signature  $\Sigma$ . If we define  $\Gamma$  as follows:*

$$\Gamma = \Sigma + [x : r \mid x \in R] + [x : d \mid x \in D] + [x : s \mid x \in S]$$

*then  $\Gamma$  is well-defined (i.e. is a function) and  $\Gamma \vdash_{Dst} e : s$ .*

*Proof.* Well-definedness follows from the fact that the sets  $R, D$  and  $S$  are disjoint (Lemma 5.6). Now we prove that  $\Gamma \vdash_{Dst} e : s$  by induction on the structure of  $e$ . We distinguish cases:

- **Case  $e \equiv c$**

Since  $(R, D, S) = (\emptyset, \emptyset, \emptyset)$ , we get  $\Gamma = \Sigma$ . Rule  $[LIT_{Dst}]$  allows us to derive  $\emptyset \vdash_{Dst} c : s$ . By using the  $[EXT_{Dst}]$  rule we obtain  $\Gamma \vdash_{Dst} e : s$ .

- **Cases  $e \in \{x, x @ r\}$**

In these cases  $\Gamma = \Sigma + [x : s]$ . By applying the  $[VAR_{Dst}]$  or  $[COPY_{Dst}]$  rules we obtain  $[x : s] \vdash_{Dst} e : s$ , from which we get  $\Sigma + [x : s] \vdash_{Dst} e : s$  by using  $[EXT_{Dst}]$ .

- **Case  $e \equiv C \bar{a}_i^n @ r$**

We obtain  $\Gamma = \Sigma \cup [\bar{a}_i : \bar{s}^n]$ . The environment  $\bigoplus_{i=1}^n [a_i : s]$  is well-defined, as it is the union of bindings with  $s$  marks. So, by applying  $[CONS_{Dst}]$  and  $[EXT_{Dst}]$  we get  $\Gamma \vdash e : s$ .

- **Case  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$**

Assume that  $\Sigma(f) = \bar{m}_i^n \rightarrow s$ . Each  $m_i$  belongs to the set  $\{d, s\}$ . We get, for every  $i \in \{1..n\}$  (assuming  $a_i \in \mathbf{Var}$ )

$$\Gamma(a_i) = s \Leftrightarrow a_i \in S \Leftrightarrow m_i = s$$

Analogously:

$$\Gamma(a_i) = d \Leftrightarrow a_i \in D_i \Leftrightarrow m_i = d$$

First we prove that  $\bigoplus_{i=1}^n [a_i : m_i]$  is well-defined. If it were not, we would get two different  $i, j \in \{1..n\}$  such that  $a_i = a_j$ , and  $m_i = d$ .

- If  $m_j = d$ , then  $a_j = a_i \in D_i$  and  $a_i = a_j \in D_j$ , so  $D_i$  and  $D_j$  would not be disjoint, which contradicts the assumptions of  $[APP_I]$ .
- If  $m_j = s$ , then  $a_i = a_j \in S$ , but  $a_j = a_i \in D$ . This contradicts the  $S \cap D = \emptyset$  assumption of  $[APP_I]$ .

If we define  $\Gamma_R = [y : r \mid y \in R]$ , the environment  $\Gamma_R + \bigoplus_{i=1}^n [a_i : m_i]$  is also well-defined, because of the conditions  $R \cap D = \emptyset$  and  $R \cap S = \emptyset$ . We can apply rule  $[APP_{Dst}]$  so as to get  $\Gamma_R + [f : \bar{m}_i \rightarrow s] + \bigoplus_{i=1}^n [a_i : m_i] \vdash_{Dst} e : s$ . Since  $\Gamma$  is a superset of this environment, we get  $\Gamma \vdash_{Dst} e : s$  by  $[EXT_{Dst}]$ .

- **Case  $e \equiv \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2$**

From the premises in  $[LET_I]$  we get  $e_1 \vdash (R_1, D_1, S_1)$ , and  $e_2 \vdash (R_2, D_2, S_2)$ . By induction hypothesis,  $\Gamma_1 \vdash_{Dst} e_1 : s$ , and  $\Gamma'_2 \vdash_{Dst} e_2 : s$ , where  $\Gamma_1$  and  $\Gamma'_2$  are defined as follows:

$$\Gamma_1 = \Sigma + [x : r \mid x \in R_1] + [x : d \mid x \in D_1] + [x : s \mid x \in S_1]$$

$$\Gamma'_2 = \Sigma + [x : r \mid x \in R_2] + [x : d \mid x \in D_2] + [x : s \mid x \in S_2]$$

We define  $m_1$  and  $\Gamma_2$  as follows:

$$\Gamma_2 = \begin{cases} \Gamma'_2 \setminus x_1 & \text{if } x_1 \in \text{dom } \Gamma'_2 \\ \Gamma'_2 & \text{if } x_1 \notin \text{dom } \Gamma'_2 \end{cases} \quad m_1 = \begin{cases} \Gamma'_2(x_1) & \text{if } x_1 \in \text{dom } \Gamma'_2 \\ s & \text{if } x_1 \notin \text{dom } \Gamma'_2 \end{cases}$$

In both cases we can derive  $\Gamma_2 + [x_1 : m_1] \vdash_{Dst} e_2 : s$  (we have to use  $[EXT_{Dst}]$  if  $x_1 \notin \text{dom } \Gamma'_2$ ). From Lemma 5.7,

$$\Gamma \setminus (\text{dom } \Sigma) = \Gamma_1 \setminus (\text{dom } \Sigma) \sqcup \Gamma_2 \setminus (\text{dom } \Sigma)$$

and, hence,

$$\Gamma = \Gamma_1 \sqcup \Gamma_2$$

So we can apply  $[LET_{Dst}]$  if its side condition holds. From the side condition of  $[LET_I]$  we get:

$$(R_1 \cup D_1) \cap fv(e_2) = \emptyset$$

which is equivalent to

$$\begin{aligned} & (\{x \in \text{dom } \Gamma_1 \mid \Gamma_1(x) = r\} \cup \{x \in \text{dom } \Gamma_1 \mid \Gamma_1(x) = d\}) \cap fv(e_2) = \emptyset \\ \Leftrightarrow & (\{x \in \text{dom } \Gamma_1 \mid \Gamma_1(x) \in \{d, r\}\}) \cap fv(e_2) = \emptyset \\ \Leftrightarrow & \forall x \in \text{dom } \Gamma_1. \Gamma_1(x) \notin \{d, r\} \vee x \notin fv(e_2) \\ \Leftrightarrow & \forall x \in \text{dom } \Gamma_1. \Gamma_1(x) \in \{d, r\} \Rightarrow x \notin fv(e_2) \end{aligned}$$

which proves the result.

- **Case  $e \equiv \text{case } x \text{ of } \overline{C_i} \overline{x_{ij}^{n_i}} \rightarrow e_i^n$**

For each sub-expression  $e_i$  we get  $e_i \vdash (D_i, R_i, S_i)$ , which implies, by induction hypothesis  $\Gamma'_i \vdash_{Dst} e_i : s$ , where  $\Gamma'_i$  is defined as follows:

$$\Gamma'_i = \Sigma + [z : r \mid z \in R_i] + [z : d \mid z \in D_i] + [z : s \mid z \in S_i]$$

For each  $i$  we define  $\Gamma_i = \Gamma'_i \setminus \{\overline{x_{ij}^{n_i}}\}$ . From the judgement  $\Gamma'_i \vdash_{Dst} e_i : s$  (and by using rule  $[EXT_{Dst}]$  if necessary) we can infer  $\Gamma_i + [\overline{x_{ij} : \overline{m_{ij}^{n_i}}}] \vdash_{Dst} e_i : s$  for some  $\{\overline{m_{ij}^{n_i}}\}$ . Let  $(R', D', S') = \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i)$ . If we define  $\Gamma'$  as follows:

$$\Gamma' = \Sigma + [z : r \mid z \in R'] + [z : d \mid z \in D'] + [z : s \mid z \in S']$$

Then, by Lemma 5.7,

$$\Gamma' \setminus (\text{dom } \Sigma) = \bigsqcup_{i=1}^n \Gamma_i \setminus (\text{dom } \Sigma)$$

which implies

$$\Gamma' = \bigsqcup_{i=1}^n \Gamma_i$$

Then, for every  $i \in \{1..n\}$  and  $z \in \text{dom } \Gamma_i$ ,  $\Gamma_i(z) \leq \Gamma'(z)$ . This allows us to apply  $[WEAK_{Dst}]$  and/or  $[EXT_{Dst}]$  in order to obtain  $\Gamma' + [\overline{x_{ij} : \overline{m_{ij}^{n_i}}}] \vdash_{Dst} e_i : s$  for each  $i \in \{1..n\}$ . Now we distinguish cases:

- $x \in R' \cup D' \cup S'$

Then  $R = R'$ ,  $D = D'$  and  $S = S'$ . Therefore  $\Gamma = \Gamma'$  and  $\Gamma + [\overline{x_{ij} : m_j^{n_i}}] \vdash_{Dst} e_i : s$  for each  $i \in \{1..n\}$ . We apply  $[CASE_{Dst}]$  in order to obtain  $\Gamma \vdash_{Dst} e : s$ .

–  $x \notin R' \cup D' \cup S'$

In this case  $R = R'$ ,  $D = D'$ , but  $S = S' \cup \{x\}$ , so  $\Gamma = \Gamma' + [x : s]$ . By rule  $[EXT_{Dst}]$ ,  $\Gamma + [\overline{x_{ij} : m_j^{n_i}}] \vdash_{Dst} e : s$  for each  $i \in \{1..n\}$ . Again, by rule  $[CASE_{Dst}]$ , we obtain  $\Gamma \vdash_{Dst} e : s$ .

• **Case  $e \equiv \text{case! } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i^n}$**

As in the previous case, we get  $\Gamma'_i \vdash_{Dst} e_i : s$  for each  $i \in \{1..n\}$  by induction hypothesis, where  $\Gamma'_i$  is defined as follows:

$$\Gamma'_i = \Sigma + [z : r \mid z \in R_i] + [z : d \mid z \in D_i] + [z : s \mid z \in S_i]$$

Let us think about the pattern variables  $\overline{x_{ij}^{n_i}}$ . By the conditions  $Rec_i \cap R_i = \emptyset$  and  $(P_i \setminus Rec_i) \cap (D_i \cup R_i) = \emptyset$  occurring in  $[CASE!_I]$  we know that, for every  $j \in \{1..n_i\}$ :

- If  $j \in RecPos(C_i)$ , then  $x_{ij} \in Rec_i$  and  $x_{ij}$  occurs in  $\Gamma'_i$  with an s or d mark, or does not occur in  $\Gamma'_i$  at all. In the latter case, we can apply rule  $[EXT_{Dst}]$  to make it occur with a d mark. If  $\Gamma'_i(x_{ij}) = s$ , we can apply rule  $[WEAK_{Dst}]$  to upgrade it to a d mark.
- If  $j \notin RecPos(C_i)$ , then  $x_{ij} \in P_i \setminus Rec_i$ , and  $x_{ij}$  occurs in  $\Gamma'_i$  with an s mark, or does not occur in  $\Gamma'_i$  at all. Again, in the latter case we can apply  $[EXT_{Dst}]$  and add  $x_{ij}$  to the environment with an s mark.

Given the above, we can manage to obtain a judgement  $\Gamma'_i \setminus \{\overline{x_{ij}^{n_i}}\} + [\overline{x_{ij} : m_j^{n_i}}] \vdash e_i : s$ , where the  $\overline{m_j^{n_i}}$  are defined as follows:

$$\forall i \in \{1..n\}. \forall j \in \{1..n_i\}. m_j^{n_i} = \begin{cases} d & \text{if } j \in RecPos(C_i) \\ s & \text{if } j \notin RecPos(C_i) \end{cases} \quad (5.2)$$

Let  $(R', D', S') = \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i)$ . By defining  $\Gamma'$  in the following way:

$$\Gamma' = \Sigma + [z : r \mid z \in R'] + [z : d \mid z \in D'] + [z : s \mid z \in S']$$

we can prove that  $\Gamma' = \bigsqcup_{i=1}^n \Gamma'_i \setminus \{x_{ij}\}$  and  $\Gamma' + [\overline{x_{ij} : m_j^{n_i}}] \vdash_{Dst} e_i : s$ . This is done in the same way as in the non-destructive **case**.

If we define  $(R'', D'', S'') \stackrel{\text{def}}{=} ((R', D', S') \setminus \{x\}) \sqcup (R_{SHR}, \emptyset, \emptyset)$ , where  $R_{SHR}$  is defined as in the  $[CASE!_I]$  rule, we get, by Lemma 5.7,

$$\Gamma'' \setminus (\text{dom } \Sigma) = \Gamma_R \sqcup (\Gamma' \setminus (\text{dom } \Sigma \cup \{x\})) \quad (5.3)$$

where:

$$\Gamma'' = \Sigma + [z : r \mid z \in R''] + [z : d \mid z \in D''] + [z : s \mid z \in S'']$$

$$\Gamma_R = [z : r \mid z \in R_{SHR}]$$

By incorporating  $\Sigma$  into both sides of (5.3) we obtain  $\Gamma'' = \Gamma_R \sqcup (\Gamma' \setminus x)$ . However, notice that

$R'' = R, S'' = S$  and  $D'' \cup \{x\} = D$ . Therefore:

$$\begin{aligned}\Gamma'' + [x : d] &= \Sigma + [z : r \mid z \in R''] + [z : d \mid z \in D''] + [z : s \mid z \in S''] + [x : d] \\ &= \Sigma + [z : r \mid z \in R] + [z : d \mid z \in D] + [z : s \mid z \in S] \\ &= \Gamma\end{aligned}$$

So  $\Gamma = \Gamma'' + [x : d] = \Gamma_R \sqcup (\Gamma' \setminus x) + [x : d]$ . We can apply  $[CASE!_{Dst}]$  in order to get  $\Gamma \vdash_{Dst} e : s$  if we are able to prove that the side conditions of this rule hold. The condition of  $\forall z \in R \cup \{x\}. \forall i \in \{1..n\}. z \notin fv(e_i)$  occurring in  $[CASE!_{Dst}]$  follows from its counterpart  $\forall i \in \{1..n\}. (R_{SHR} \cup \{x\}) \cap fv(e_i) = \emptyset$  in  $[CASE!_I]$ . With respect to the *inh* conditions on the  $\overline{m'_{ij}}^{n_i}$ , they follow immediately from (5.2). □

The correctness of the *inferMarksDef* function for inferring the mark signatures of function definitions follows immediately from this theorem. Before proving this, we need to define what is a typeable function, since the  $\vdash_{Dst}$  rules only apply to expressions. By inspecting the  $[FUN]$  rule of Figure 3.9 we can derive the following definition:

**Definition 5.9.** Given a fixed signature environment  $\Sigma$  not containing  $f$  in its domain, a mark signature  $\overline{m}_i^n \rightarrow s$  is said to be correct for the function definition  $f \overline{x}_i^{n'} @ \overline{r}_j^m = e_f$  if and only if these three conditions hold:

1.  $n = n'$ .
2.  $\forall i \in \{1..n\}. m_i \in \{s, d\}$ .
3.  $\Sigma + [f : \overline{m}_i^n \rightarrow s] + [\overline{x}_i : \overline{m}_i^n] \vdash_{Dst} e_f : s$ .

Given a function definition  $f \overline{x}_i^{n'} @ \overline{r}_j^m = e_f$ , let us assume that *inferMarksFP* has reached a fixed point. This means that, if we apply the inference rules to the body of the expression under a signature  $\Sigma = \Sigma' + [f : \overline{m}_i^n \rightarrow s]$  for some  $\overline{m}_i^n$ , and we obtain  $e_f \vdash (R_f, D_f, S_f)$ , the following judgement is derivable by Theorem 5.8:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [x : r \mid x \in R_f] + [x : d \mid x \in D_f] + [x : s \mid x \in S_f] \vdash e_f : s$$

The set  $R_f$  must be empty. Otherwise *inferMarksFP* would have returned an error, and it would have not reached a fixed point. Hence:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [x : d \mid x \in D_f] + [x : s \mid x \in S_f] \vdash e_f : s$$

We know that  $D_f \cup S_f \subseteq fv(e_f) \subseteq \{\overline{x}_i^{n'}\}$ , but there could be parameters  $\overline{x}_i^{n'}$  that do not appear in any of these sets, so let us apply the  $[EXT_{Dst}]$  rule to the previous judgement in order to add those variables to the type environment with a safe mark:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [x : d \mid x \in D_f] + [x : s \mid x \in S_f] + [x : s \mid x \in \{\overline{x}_i^{n'}\} \setminus (D_f \cup S_f)] \vdash e_f : s$$

This is equivalent to the following:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [x : d \mid x \in D_f] + [x : s \mid x \in \{\overline{x}_i^{n'}\} \setminus D_f] \vdash e_f : s$$

For each  $i \in \{1..n\}$ , let us define:

$$m'_i = \begin{cases} d & \text{if } x_i \in D_f \\ s & \text{if } x_i \notin D_f \end{cases}$$

Our typing judgement can be transformed as follows:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [\overline{x}_i : \overline{m}'_i^n] \vdash e_f : s$$

Since we have assumed a fixed point for *inferMarksFP*, we know that  $m'_i = m_i$  for each  $i \in \{1..n\}$ . Therefore:

$$\Sigma' + [f : \overline{m}_i^n \rightarrow s] + [\overline{x}_i : \overline{m}_i^n] \vdash e_f : s$$

Now the marks of the parameters  $\overline{x}_i^n$  match the marks in the signature of  $f$ . If we combine this judgement with its counterpart in the  $\vdash_{Reg}$  rules, the rule [FUN] can be applied, and hence the function is well-typed.

## 5.4.2 Termination

The discussion at the end of the last section only makes sense if the algorithm reaches a fixed point. If it does not, it could be due to three situations:

1. The application of the inference rules results in an error, since no rule can be applied.
2. The application of the inference rules to the body of a function definition results in a triple  $(R, D, S)$  in which  $R \neq \emptyset$ .
3. The *inferMarksFP* function does not terminate.

In this section we prove that the third situation never holds. Assume a function definition  $f \overline{x}_i^n @ \overline{r}_j^m = e_f$ . The key idea is to define an order between mark signatures, and to prove that each iteration of the  $\vdash$  rules to  $e_f$  under an environment  $\Sigma$  results in a signature which is greater or equal than  $\Sigma(f)$ . So, if *inferMarksFP* did not terminate, there would exist an infinitely strictly-increasing chain of mark signatures. However, this cannot happen, since the set of mark signatures associated with a function of  $n$  parameters has  $2^n$  elements, and hence is finite.

So, let us start defining an order between signatures:

**Definition 5.10.** Given a function of  $n$  parameters, we define the following order between signatures:

$$\overline{m}_i^n \rightarrow s \leq \overline{m}'_i^n \rightarrow s \stackrel{def}{\iff} \forall i \in \{1..n\}. m_i \leq m'_i$$

And we extend this order to  $\Sigma$  in the standard way.

From this definition it follows that the set of mark signatures for a given function with this order has the structure of a lattice. Figure 5.13 shows the lattices corresponding to a function of two and three parameters, respectively.

We define a similar order for  $(R, D, S)$  tuples. A tuple  $(R, D, S)$  is said to be lower than  $(R', D', S')$  if the former contains less unsafe variables (i.e. variables in  $R$  and  $D$ ).

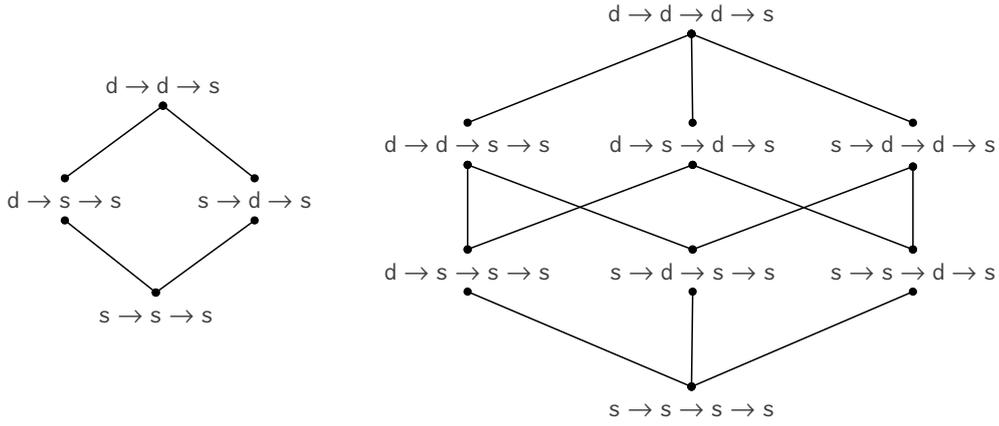


Figure 5.13: Lattice of mark signatures for functions with two and three parameters, respectively.

**Definition 5.11.** We define the following pre-order between triples  $(R, D, S)$ :

$$(R, D, S) \sqsubseteq (R', D', S') \Leftrightarrow R \cup D \subseteq R' \cup D'$$

**Lemma 5.12.** The  $\sqcup$  operator is  $\sqsubseteq$ -monotonic w.r.t. all its parameters.

*Proof.* Let us assume that  $(R_1, D_1, S_1) \sqsubseteq (R'_1, D'_1, S'_1)$  and that  $(R_2, D_2, S_2) \sqsubseteq (R'_2, D'_2, S'_2)$ . If we define,

$$\begin{aligned} (R, D, S) &= (R_1, D_1, S_1) \sqcup (R_2, D_2, S_2) \\ (R', D', S') &= (R'_1, D'_1, S'_1) \sqcup (R'_2, D'_2, S'_2) \end{aligned}$$

we get:

$$\begin{aligned} R \cup D &= (R_1 \cup R_2) \cup ((D_1 \cup D_2) \setminus (R_1 \cup R_2)) \\ &= R_1 \cup R_2 \cup D_1 \cup D_2 \\ &\subseteq R'_1 \cup R'_2 \cup D'_1 \cup D'_2 \\ &= (R'_1 \cup R'_2) \cup ((D'_1 \cup D'_2) \setminus (R'_1 \cup R'_2)) \\ &= R' \cup D' \end{aligned}$$

Hence,  $(R, D, S) \sqsubseteq (R', D', S')$ . □

Given a fixed expression  $e$ , we can consider the application of the  $\vdash$  rules as a function which, given a signature environment  $\Sigma$ , returns the result of applying the  $\vdash$  rules on  $e$  under that environment. The following theorem proves that this function is monotone with respect to the  $\leq$  order defined on signature environments.

**Theorem 5.13.** Let  $\Sigma$  and  $\Sigma'$  be two signature environments without in-danger marks. If  $\Sigma \leq \Sigma'$ ,  $e \vdash_{\Sigma} (R, D, S)$ , and  $e \vdash_{\Sigma'} (R', D', S')$ , then  $(R, D, S) \sqsubseteq (R', D', S')$

*Proof.* By induction on the structure of  $e$ . We distinguish cases:

- **Cases**  $e \in \{c, x, x @ r, C \bar{a}_i^n @ r\}$

Trivial, since  $R \cup D = \emptyset = R' \cup D'$ .

- **Case**  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$

Let  $\Sigma(f) = \bar{m}_i \rightarrow s$  and  $\Sigma'(f) = \bar{m}'_i \rightarrow s$ . By assumption,  $m_i = d$  implies  $m'_i = d$ . Therefore:

$$\begin{aligned} R \cup D &= \bigcup_{m_i=d} \text{sharerec}(a_i, e) \setminus \{a_i\} \cup \bigcup_{i=1}^n D_i \\ &= \bigcup_{m_i=d} \text{sharerec}(a_i, e) \setminus \{a_i\} \cup \{a_i \mid m_i = d\} \\ &\subseteq \bigcup_{m'_i=d} \text{sharerec}(a_i, e) \setminus \{a_i\} \cup \{a_i \mid m'_i = d\} \\ &= R' \cup D' \end{aligned}$$

- **Case**  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$

Let us assume:

$$\begin{aligned} e_1 \vdash_{\Sigma} (R_1, D_1, S_1) \quad e_1 \vdash_{\Sigma'} (R'_1, D'_1, S'_1) \\ e_2 \vdash_{\Sigma} (R_2, D_2, S_2) \quad e_2 \vdash_{\Sigma'} (R'_2, D'_2, S'_2) \end{aligned}$$

By induction hypothesis,  $(R_1, D_1, S_1) \sqsubseteq (R'_1, D'_1, S'_1)$  and  $(R_2, D_2, S_2) \sqsubseteq (R'_2, D'_2, S'_2)$ . Since  $\sqcup$  is monotonic with respect to both operands, we get:

$$(R, D, S) = (R_1, D_1, S_1) \sqcup ((R_2, D_2, S_2) \setminus \{x_1\}) \sqsubseteq (R'_1, D'_1, S'_1) \sqcup ((R'_2, D'_2, S'_2) \setminus \{x_1\}) = (R', D', S')$$

- **Case**  $e \equiv \text{case } x \text{ of } \overline{C_i \bar{x}_{ij}^{m_i} \rightarrow e_i^n}$

Assume, for every  $i \in \{1..n\}$ ,  $e_i \vdash_{\Sigma} (R_i, D_i, S_i)$  and  $e_i \vdash_{\Sigma'} (R'_i, D'_i, S'_i)$ . By induction hypothesis,  $(R_i, D_i, S_i) \sqsubseteq (R'_i, D'_i, S'_i)$  and by monotonicity of  $\sqcup$  we get:

$$(R, D, S) = \left( \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i) \right) \sqcup (\emptyset, \emptyset, \{x\}) \sqsubseteq \left( \bigsqcup_{i=1}^n ((R'_i, D'_i, S'_i) \setminus P_i) \right) \sqcup (\emptyset, \emptyset, \{x\}) = (R', D', S')$$

- **Case**  $e \equiv \text{case! } x \text{ of } \overline{C_i \bar{x}_{ij}^{m_i} \rightarrow e_i^n}$

Again, we assume, for every  $i \in \{1..n\}$ ,  $e_i \vdash_{\Sigma} (R_i, D_i, S_i)$  and  $e_i \vdash_{\Sigma'} (R'_i, D'_i, S'_i)$ . Let us define:

$$(R_a, D_a, S_a) = \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus P_i) \quad (R'_a, D'_a, S'_a) = \bigsqcup_{i=1}^n ((R'_i, D'_i, S'_i) \setminus P_i)$$

$$(R_b, D_b, S_b) = ((R_a, D_a, S_a) \setminus \{x\}) \sqcup (R_{SHR}, \emptyset, \emptyset) \quad (R'_b, D'_b, S'_b) = ((R'_a, D'_a, S'_a) \setminus \{x\}) \sqcup (R_{SHR}, \emptyset, \emptyset)$$

where  $R_{SHR}$  is as defined in [CASE!]. By monotonicity of  $\sqcup$  operator, we get:

$$\forall i. (R_i, D_i, S_i) \sqsubseteq (R'_i, D'_i, S'_i) \Rightarrow (R_a, D_a, S_a) \sqsubseteq (R'_a, D'_a, S'_a) \Rightarrow (R_b, D_b, S_b) \sqsubseteq (R'_b, D'_b, S'_b)$$

which implies  $R_b \cup D_b \subseteq R'_b \cup D'_b$ . Therefore:

$$R \cup D = R_b \cup D_b \cup \{x\} \subseteq R'_b \cup D'_b \cup \{x\} = R' \cup D'$$

□

The previous lemma is the key result to prove the termination of *inferMarksDef*. Assume that the latter function is called with a signature environment  $\Sigma$  and a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ , and that *inferMarksFP* does not terminate. The first call to *inferMarksFP* is done with a signature  $\Sigma_0$  in which  $\Sigma_0(f) = \bar{s}^n \rightarrow s$ . Let us assume that  $e_f \vdash_{\Sigma_0} (D_0, R_0, S_0)$  for some  $D_0, R_0$ , and  $S_0$ . The second call is done with a signature  $\Sigma_1$  such that  $\Sigma_0(f) < \Sigma_1(f)$ , so  $\Sigma_0 \leq \Sigma_1$ . The inequality is strict, since *inferMarksFP* would have terminated otherwise. Let us assume that  $e_f \vdash_{\Sigma_1} (D_1, R_1, S_1)$ . By Theorem 5.13 we know that  $D_0 \cup R_0 \subseteq D_1 \cup R_1$ . Since both  $R_0$  and  $R_1$  are empty (otherwise the algorithm would stop with an error), we get  $D_0 \subseteq D_1$ , which results in the following call to *inferMarksFP* under an environment  $\Sigma_2$  such that  $\Sigma_1(f) < \Sigma_2(f)$ . By repeating this process, we get an infinite ascending chain:

$$\Sigma_0(f) < \Sigma_1(f) < \Sigma_2(f) < \dots$$

This leads to a contradiction, since the set of possible signatures for  $f$  is finite.

### 5.4.3 Completeness

In this section we prove that, if a function is typeable w.r.t. the  $\vdash_{Dst}$  rules, the inference algorithm is able to return a mark signature for that function. Throughout this section we will write our type environments as  $\Sigma + \Gamma$ , where  $\Sigma$  only contains bindings from function and constructor names to mark signatures, and  $\Gamma$  associates variables with marks. We also assume that the mark signatures in  $\Sigma$  do not contain in-danger marks.

The first thing to point out is that there may be more than one correct mark signature for a given function. In principle, we want to prove that our algorithm returns one of these correct signatures. However, we should be more precise and specify *which* signature is given by the algorithm. Recall that the set of mark signatures for a given function is a lattice with respect to the order of Definition 5.10. The completeness proof is done in two steps:

1. Among all the correct mark signatures for a given function, there is a *minimal* correct mark signature.
2. Our algorithm computes this minimal correct mark signature.

The search for the minimal correct mark signature is closely related to the search of the minimal mark environment which types a given expression, so let us prove that such a minimal mark environment exists. Firstly we prove that the set of mark environments with the  $\leq$  order defined in Section 3.7 has the structure of a distributive complete lattice:

**Proposition 5.14.** *The set of mark environments  $\Gamma$  such that  $\text{dom } \Gamma \subseteq \mathbf{Var}$  is a distributive complete lattice with respect to the  $\leq$  order.*

*Proof.* The  $\sqcup$  and  $\sqcap$  operators satisfy the following properties,

$$\text{dom } (\Gamma_1 \sqcup \Gamma_2) = \text{dom } \Gamma_1 \cup \text{dom } \Gamma_2 \tag{5.4}$$

$$\text{dom } (\Gamma_1 \sqcap \Gamma_2) = \text{dom } \Gamma_1 \cap \text{dom } \Gamma_2 \tag{5.5}$$

$\Gamma_1(x)$	$\Gamma_2(x)$	$\Gamma_3(x)$	$((\Gamma_1 \sqcup \Gamma_2) \sqcap \Gamma_3)(x)$	$((\Gamma_1 \sqcap \Gamma_3) \sqcup (\Gamma_2 \sqcap \Gamma_3))(x)$
–	–	s	s	s
–	–	r	$\Gamma_1 \sqcup \Gamma_2$	$\Gamma_1 \sqcup \Gamma_2$
r	–	d	d	d
–	r	d	d	d
s	s	d	s	s
s	d	d	d	d
d	s	d	d	d
d	d	d	d	d

Figure 5.14: This table covers all the possibilities for the values of  $\Gamma_1(x)$ ,  $\Gamma_2(x)$  and  $\Gamma_3(x)$ . The dash symbol (–) stands for any mark. In every case, the desired equality holds.

and they are defined as follows:

$$\forall x \in \text{dom } \Gamma_1 \cup \text{dom } \Gamma_2. (\Gamma_1 \sqcup \Gamma_2)(x) = \begin{cases} \Gamma_1(x) & x \notin \text{dom } \Gamma_2 \\ \Gamma_2(x) & x \notin \text{dom } \Gamma_1 \\ \Gamma_1(x) \sqcup \Gamma_2(x) & \text{otherwise} \end{cases}$$

$$\forall x \in \text{dom } \Gamma_1(x) \cap \text{dom } \Gamma_2. (\Gamma_1 \sqcap \Gamma_2)(x) = \Gamma_1(x) \sqcap \Gamma_2(x)$$

Given a (possibly infinite) family  $\mathcal{G}$  of type environments, the result of  $\sqcap \mathcal{G}$  is defined as follows:

$$\text{dom } \sqcap \mathcal{G} = \bigcap_{\Gamma \in \mathcal{G}} \text{dom } \Gamma \quad \forall x \in \text{dom } \sqcap \mathcal{G}. (\sqcap \mathcal{G})(x) = \prod_{\Gamma \in \mathcal{G}} \Gamma(x)$$

and similarly with  $\sqcup \mathcal{G}$ . Since the ordered set  $(\{r, d, s\}, \leq)$  is a complete lattice, so is the set of mark environments. Finally, we prove distributivity. Let us prove that, for every  $\Gamma_1, \Gamma_2$  and  $\Gamma_3$ :

$$(\Gamma_1 \sqcup \Gamma_2) \sqcap \Gamma_3 = (\Gamma_1 \sqcap \Gamma_2) \sqcup (\Gamma_1 \sqcap \Gamma_3) \quad (5.6)$$

The fact that both sides of the equality have the same domain follows from (5.4), (5.5), and set algebra. Let us assume a variable  $x$  belonging to that domain. It must hold that  $x \in \text{dom } \Gamma_3$ , and  $x \in \text{dom } \Gamma_1 \cup \text{dom } \Gamma_2$ . If  $x \notin \text{dom } \Gamma_1$  we get:

$$((\Gamma_1 \sqcup \Gamma_2) \sqcap \Gamma_3)(x) = (\Gamma_2 \sqcap \Gamma_3)(x) = ((\Gamma_1 \sqcap \Gamma_2) \sqcup (\Gamma_2 \sqcap \Gamma_3))(x)$$

since  $x$  does not belong to the domain of  $\Gamma_1 \sqcap \Gamma_2$ . If  $x \notin \text{dom } \Gamma_2$  we proceed in a similar way. Finally, if  $x$  belongs to the domain of the three environments, we distinguish cases as in Figure 5.14. The equality holds in all these cases, so we have proved (5.6). The fact  $(\Gamma_1 \sqcap \Gamma_2) \sqcup \Gamma_3 = (\Gamma_1 \sqcup \Gamma_3) \sqcap (\Gamma_2 \sqcup \Gamma_3)$  follows by duality.  $\square$

Given a fixed  $\Sigma$ , the next lemma proves that, if there exists two mark environments typing a given expression, the greatest lower bound of them (which exists, by Proposition 5.14) also types that expression.

**Lemma 5.15.** *If  $\Sigma + \Gamma_1 \vdash_{Dst} e : s$  and  $\Sigma + \Gamma_2 \vdash_{Dst} e : s$ , then  $\Sigma + (\Gamma_1 \sqcap \Gamma_2) \vdash_{Dst} e : s$ .*

*Proof.* By induction on the structure of  $e$ . Without loss of generality, let us assume that the last rule in each of the  $\vdash_{Dst}$  derivations is neither  $[EXT_{Dst}]$  nor  $[WEAK_{Dst}]$ . Otherwise we can prove the existence of

some  $\Sigma' \subseteq \Sigma$ ,  $\Gamma'_1 \leq \Gamma_1$ , and  $\Gamma'_2 \leq \Gamma_2$  such that  $\Sigma' + \Gamma'_1 \vdash_{Dst} e : s$  and  $\Sigma' + \Gamma'_2 \vdash_{Dst} e : s$ , and this condition holds in both derivations. Lemma 3.49 is useful for ensuring that  $\Sigma'$  is the same in both judgements. Once we have proved that  $\Sigma' + (\Gamma'_1 \sqcap \Gamma'_2) \vdash_{Dst} e : s$  we know that  $\Gamma'_1 \sqcap \Gamma'_2 = \Gamma_1 \sqcap \Gamma_2$  and, by the  $[EXT_{Dst}]$  rule,  $\Sigma + (\Gamma_1 \sqcap \Gamma_2) \vdash_{Dst} e : s$ . Therefore, we can safely rule out the possibility of having  $[EXT_{Dst}]$  or  $[WEAK_{Dst}]$  as the last rule applied, and we distinguish between the remaining cases:

- **Cases**  $[LIT_{Dst}]$ ,  $[VAR_{Dst}]$ ,  $[COPY_{Dst}]$ , and  $[CONS_{Dst}]$ .

All these cases are trivial, since they imply that  $\Sigma = \emptyset$  and  $\Gamma_1 = \Gamma_2$ , so  $\Sigma + (\Gamma_1 \sqcap \Gamma_2) = \Gamma_1 = \Gamma_2$ .

- **Case**  $[APP_{Dst}]$ .

Let  $e \equiv f \bar{x}_i @ \bar{r}_j$ . The value of  $\Sigma(f)$  determines the contents of  $\Gamma_1$  and  $\Gamma_2$ . Since  $\Sigma$  is the same in both judgements, we get  $\Gamma_1 = \Gamma_2$ , and the lemma holds trivially.

- **Case**  $[LET_{Dst}]$ .

Let  $e \equiv \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2$ , and the following derivations:

$$\frac{\frac{\Sigma_{11} + \Gamma_{11} \vdash_{Dst} e_1 : s \quad \Sigma_{12} + \Gamma_{12} + [x_1 : m_1] \vdash_{Dst} e_2 : s}{\Sigma + (\underbrace{\Gamma_{11} \sqcup \Gamma_{12}}_{\Gamma_1}) \vdash_{Dst} e : s}}{\frac{\Sigma_{21} + \Gamma_{21} \vdash_{Dst} e_1 : s \quad \Sigma_{22} + \Gamma_{22} + [x_1 : m_2] \vdash_{Dst} e_2 : s}{\Sigma + (\underbrace{\Gamma_{21} \sqcup \Gamma_{22}}_{\Gamma_2}) \vdash_{Dst} e : s}}$$

since each  $\Sigma_{ij}$  is a subset of  $\Sigma$  for each  $i, j \in \{1, 2\}$ , we can apply  $[EXT_{Dst}]$  in each of the premises so as to obtain  $\Sigma + \Gamma_{ij}$  in each judgement, possibly followed by  $[x_1 : m_i]$  (where  $i, j \in \{1, 2\}$ ). By induction hypothesis, we get:

$$\Sigma + (\Gamma_{11} \sqcap \Gamma_{21}) \vdash_{Dst} e_1 : s \quad \Sigma + (\Gamma_{12} \sqcap \Gamma_{22}) + [x_1 : m_1 \sqcap m_2] \vdash_{Dst} e_2 : s$$

We can apply  $[LET_{Dst}]$ , since  $(\Gamma_{11} \sqcap \Gamma_{21})(x) \in \{d, r\}$  implies  $\Gamma_{11}(x) \in \{d, r\}$ , or  $\Gamma_{21} \in \{d, r\}$ , which implies  $x \notin \mathit{fv}(e_2)$ . Hence we get,

$$(\Sigma + (\Gamma_{11} \sqcap \Gamma_{21})) \sqcup (\Sigma + (\Gamma_{12} \sqcap \Gamma_{22})) \vdash_{Dst} e : s$$

which is equivalent to:

$$\Sigma + ((\Gamma_{11} \sqcap \Gamma_{21}) \sqcup (\Gamma_{12} \sqcap \Gamma_{22})) \vdash_{Dst} e : s$$

If we prove that  $(\Gamma_{11} \sqcap \Gamma_{21}) \sqcup (\Gamma_{12} \sqcap \Gamma_{22}) \leq (\Gamma_{11} \sqcup \Gamma_{12}) \sqcap (\Gamma_{21} \sqcap \Gamma_{22}) = \Gamma_1 \sqcap \Gamma_2$  we will be able to apply the  $[WEAK_{Dst}]$  rule to get the desired result. Thus, we have to prove this inequality. Let us define:

$$\begin{aligned} \Gamma_A &= (\Gamma_{11} \sqcap \Gamma_{21}) \sqcup (\Gamma_{12} \sqcap \Gamma_{22}) \\ \Gamma_B &= (\Gamma_{11} \sqcup \Gamma_{12}) \sqcap (\Gamma_{21} \sqcap \Gamma_{22}) \end{aligned}$$

Firstly we prove that  $\mathit{dom} \ \Gamma_A \subseteq \mathit{dom} \ \Gamma_B$ . Let  $x \in \mathit{dom} \ \Gamma_A$ . We distinguish cases:

- $x \in \mathit{dom} \ \Gamma_{11} \wedge x \in \mathit{dom} \ \Gamma_{21}$ . In this case we get  $x \in \mathit{dom} \ (\Gamma_{11} \sqcup \Gamma_{12})$ , and  $x \in \mathit{dom} \ (\Gamma_{21} \sqcap \Gamma_{22})$ . Therefore,  $x \in \mathit{dom} \ \Gamma_B$ .

- $x \in \text{dom } \Gamma_{12} \wedge x \in \text{dom } \Gamma_{22}$ . Again, we get  $x \in \text{dom } (\Gamma_{11} \sqcup \Gamma_{12})$ , and  $x \in \text{dom } (\Gamma_{21} \sqcup \Gamma_{22})$ . Therefore,  $x \in \text{dom } \Gamma_B$ .

Now we prove that, for every  $x \in \text{dom } \Gamma_A$ , we get  $\Gamma_A(x) \leq \Gamma_B(x)$ . If  $\Gamma_A(x) = s$  we are done. If  $\Gamma_A(x) = r$  there are two possibilities:

- $(\Gamma_{11} \sqcap \Gamma_{21})(x) = r$ , which implies  $\Gamma_{11}(x) = r$ , and  $\Gamma_{21} = r$ . Hence we get  $(\Gamma_{11} \sqcup \Gamma_{12})(x) = r$ ,  $(\Gamma_{21} \sqcap \Gamma_{22})(x) = r$ , and, consequently  $\Gamma_B(x) = r$ .
- $(\Gamma_{12} \sqcap \Gamma_{22})(x) = r$ , which implies  $\Gamma_{12}(x) = r$ , and  $\Gamma_{22} = r$ . Similarly as above,  $\Gamma_B(x) = r$ .

Exactly in the same way we can prove that  $\Gamma_A(x) = d$  implies  $\Gamma_B \in \{d, r\}$ . The required result follow from this.

- **Case**  $[CASE_{Dst}]$

Let  $e \equiv \mathbf{case } x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n}$ . From the premises of the  $[CASE_{Dst}]$  rule we get,

$$\Sigma + (\Gamma_1 + [\overline{x_{ij} : m_{ij}^{n_i}}]) \vdash_{Dst} e_i : s \wedge \Sigma + (\Gamma_2 + [\overline{x_{ij} : m'_{ij}{}^{n_i}}]) \vdash_{Dst} e_i : s \quad \text{for each } i \in \{1..n\}$$

and, by induction hypothesis,

$$(\Sigma + (\Gamma_1 + [\overline{x_{ij} : m_{ij}^{n_i}}])) \sqcap (\Sigma + (\Gamma_2 + [\overline{x_{ij} : m'_{ij}{}^{n_i}}])) \vdash_{Dst} e_i : s$$

which is equivalent to:

$$\Sigma + ((\Gamma_1 \sqcap \Gamma_2) + [\overline{x_{ij} : m_{ij} \sqcap m'_{ij}{}^{n_i}}]) \vdash_{Dst} e_i : s$$

for each  $i \in \{1..n\}$ . The required result follow from applying  $[CASE_{Dst}]$  to these judgements.

- **Case**  $[CASE!_{Dst}]$

Let  $e \equiv \mathbf{case}(!) x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n}$ . From  $[CASE!_{Dst}]$  we get,

$$\Sigma + (\Gamma_1 + [\overline{x_{ij} : m_{ij}^{n_i}}]) \vdash_{Dst} e_i : s \wedge \Sigma + (\Gamma_2 + [\overline{x_{ij} : m'_{ij}{}^{n_i}}]) \vdash_{Dst} e_i : s \quad \text{for each } i \in \{1..n\}$$

The *inh* predicates in each judgement force  $m_{ij} = m'_{ij}$  for each  $i \in \{1..n\}, j \in \{1..n_i\}$ . Moreover, the  $R$  set occurring in the premises of  $[CASE!_{Dst}]$  is the same in both judgements, and so is the  $\Gamma_R$  environment. The induction hypothesis gives the following judgement:

$$(\Sigma + (\Gamma_1 + [\overline{x_{ij} : m_{ij}^{n_i}}])) \sqcap (\Sigma + (\Gamma_2 + [\overline{x_{ij} : m'_{ij}{}^{n_i}}])) \vdash_{Dst} e_i : s$$

which is equivalent to:

$$\Sigma + ((\Gamma_1 \sqcap \Gamma_2) + [\overline{x_{ij} : m_{ij}^{n_i}}]) \vdash_{Dst} e_i : s$$

The remaining conditions in  $[CASE!_{Dst}]$  follow from their counterparts in the initial judgements. Hence we get:

$$\Gamma_R \sqcup ((\Sigma + (\Gamma_1 \sqcap \Gamma_2)) \setminus x) + [x : d] \vdash_{Dst} e : s$$

The type environment can be transformed as follows:

$$\begin{aligned}
\Gamma_R \sqcup ((\Sigma + (\Gamma_1 \sqcap \Gamma_2)) \setminus x) + [x : d] &= \Sigma + (\Gamma_R \sqcup ((\Gamma_1 \sqcap \Gamma_2) \setminus x) + [x : d]) \\
&= \Sigma + (\Gamma_R \sqcup ((\Gamma_1 \setminus x) \sqcap (\Gamma_2 \setminus x))) + [x : d] \\
&= \Sigma + ((\Gamma_R \sqcup (\Gamma_1 \setminus x)) \sqcap (\Gamma_R \sqcup (\Gamma_2 \setminus x))) + [x : d] \\
&= \Sigma + ((\Gamma_R \sqcup (\Gamma_1 \setminus x) + [x : d]) \sqcap (\Gamma_R \sqcup (\Gamma_2 \setminus x) + [x : d]))
\end{aligned}$$

This proves the lemma. □

The following lemma allows us to strengthen the  $\Sigma$  signature in a typing derivation. The intuitive idea is that, if an expression is typeable under a mark environment  $\Sigma$ , we can replace some of the mark signatures in  $\Sigma$  by some other “less destructive” variants, and the expression is still typeable under the modified environment.

**Lemma 5.16.** *If  $\Sigma + \Gamma \vdash_{Dst} e : s$  and  $\Sigma'$  is an environment which contains the mark signatures of the functions being called in  $e$ , and  $\Sigma' \leq \Sigma$ , then  $\Sigma' + \Gamma \vdash_{Dst} e : s$ .*

*Proof.* By induction on the size of the typing derivation. Let us distinguish cases on the last rule applied.

- **Case**  $[EXT_{Dst}]$

We can derive  $\Sigma'' + \Gamma'' \vdash_{Dst} e : s$ , for some  $\Sigma'' \subseteq \Sigma$  and  $\Gamma'' \subseteq \Gamma$ . Let us define  $\Sigma'_1$  and  $\Sigma'_2$  as follows:

$$\begin{aligned}
\Sigma'_1 &= \Sigma' \setminus \Sigma'' \\
\Sigma'_2 &= \Sigma' \sqcap \Sigma''
\end{aligned}$$

It is obvious that  $\text{dom } \Sigma' = \text{dom } \Sigma'_1 \uplus \text{dom } \Sigma'_2$ . Since  $\Sigma'_2 \leq \Sigma''$ , and both  $\Sigma'$  and  $\Sigma''$  contain the signatures of the functions being applied in  $e$ , then  $\Sigma'_2$  also contain those signatures. We can apply the induction hypothesis in order to get:

$$\Sigma'_2 + \Gamma'' \vdash_{Dst} e : s$$

By applying  $[EXT_{Dst}]$  we get,

$$\Sigma'_1 \uplus \Sigma'_2 + \Gamma \vdash_{Dst} e : s$$

which proves the result.

- **Case**  $[WEAK_{Dst}]$

It follows trivially from the induction hypothesis.

- **Cases**  $[LIT_{Dst}]$ ,  $[VAR_{Dst}]$ ,  $[COPY_{Dst}]$ , and  $[CONS_{Dst}]$ .

We get  $\Sigma = []$ , which implies  $\Sigma' = []$ , and the result holds trivially.

- **Case**  $[APP_{Dst}]$

Let  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$ . In this case,  $\Sigma = [f : \bar{m}_i^n \rightarrow s]$  and  $\Sigma' = [f : \bar{m}'_i^n \rightarrow s]$  for some  $\bar{m}_i^n, \bar{m}'_i^n$  such that  $m'_i \leq m_i$  for all  $i \in \{1..n\}$ . Let us define:

$$R = \bigcup_{i=1}^n \{\text{sharerec}(a_i, e) \mid m_i = d\} \quad R' = \bigcup_{i=1}^n \{\text{sharerec}(a_i, e) \mid m'_i = d\}$$

Since  $m'_i = d$  implies  $m_i = d$  for every  $i \in \{1..n\}$  (since the  $m_i$  must belong to the set  $\{s, d\}$ ), we get  $R' \subseteq R$ . For the same reason, the condition  $\bigwedge_{m'_i=d} isTree(a_i)$  follows from  $\bigwedge_{m_i=d} isTree(a_i)$ , and the well-definedness of  $\bigoplus_{i=1}^n [a_i : m'_i]$  also follows from that of  $\bigoplus_{i=1}^n [a_i : m_i]$ . We apply  $[APP_{Dst}]$ :

$$[f : \overline{m'_i}^n \rightarrow s] + [y : r \mid y \in R'] + \bigoplus_{i=1}^n [a_i : m_i] \vdash_{Dst} e : s$$

Moreover, since  $R' \subseteq R$ , and none of the  $\overline{a_i}$  variables is in  $R$ , we can apply the  $[EXT_{Dst}]$  rule:

$$\underbrace{[f : \overline{m'_i}^n \rightarrow s]}_{\Sigma'} + \underbrace{[y : r \mid y \in R] + \bigoplus_{i=1}^n [a_i : m_i]}_{\Gamma} \vdash_{Dst} e : s$$

and the lemma holds.

- **Case**  $[LET_{Dst}]$

By the premises of that rule we get, assuming  $e \equiv \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2$ :

$$\Sigma_1 + \Gamma_1 \vdash_{Dst} e_1 : s \quad \Sigma_2 + \Gamma_2 + [x_1 : m_1] \vdash_{Dst} e_2 : s$$

where  $\Sigma \geq \Sigma_1, \Sigma \geq \Sigma_2$ , and  $\Gamma = \Gamma_1 \sqcup \Gamma_2$ . Let us define  $\Sigma'_1 = \Sigma_1 \sqcap \Sigma'$  and  $\Sigma'_2 = \Sigma_2 \sqcap \Sigma'$ . Hence  $\Sigma' = \Sigma'_1 \cup \Sigma'_2$ . Again,  $\Sigma'$  and  $\Sigma_1$  contain the signatures of the functions being applied in  $e_1$ , and, hence, so does  $\Sigma'_1$ . Similarly with  $\Sigma'_2$ . By induction hypothesis:

$$\Sigma'_1 + \Gamma_1 \vdash_{Dst} e_1 : s \quad \Sigma'_2 + \Gamma_2 + [x_1 : m_1] \vdash_{Dst} e_2 : s$$

By applying  $[LET_{Dst}]$  we get  $(\Sigma'_1 + \Sigma'_2) + (\Gamma_1 \sqcup \Gamma_2) \vdash_{Dst} e : s$ , from which the desired result follows.

- **Cases**  $[CASE_{Dst}]$  **and**  $[CASE!_{Dst}]$

Both follow trivially from the induction hypothesis. □

Finally, we use the results of these lemmas in order to show that the set of correct mark signatures for a given function is closed w.r.t. the greatest lower bound operator:

**Lemma 5.17.** *If  $\Phi$  is the set of correct signatures for a function  $f$ ,  $\sqcap \Phi$  is also a correct signature for  $f$ .*

*Proof.* Let us assume  $\sqcap \Phi = \overline{m'_i}^n \rightarrow s$  for some  $\overline{m'_i}^n$ . By definition of correct signature, we get:

$$\Sigma + [f : \overline{m'_i}^n \rightarrow s] + [\overline{x_i} : \overline{m'_i}^n] \vdash_{Dst} e_f : s \quad \forall (\overline{m'_i}^n \rightarrow s) \in \Phi$$

By Lemma 5.16:

$$\Sigma + [f : \overline{m'_i}^n \rightarrow s] + [\overline{x_i} : \overline{m'_i}^n] \vdash_{Dst} e_f : s \quad \forall (\overline{m'_i}^n \rightarrow s) \in \Phi$$

Since the set  $\Phi$  is finite (because so is the set of signatures for  $f$ ), we can apply Lemma 5.15 repeatedly so as to get:

$$\Sigma + [f : \overline{m'_i}^n \rightarrow s] + [\overline{x_i} : \sqcap_{\overline{m'_j}^n \rightarrow s \in \Phi} \overline{m'_j}^n] \vdash_{Dst} e_f : s$$

Finally, since  $m'_i = \prod_{\overline{m}_j \rightarrow s \in \Phi} m_i$ , we get that  $\overline{m}_i \rightarrow s$  is a correct signature for  $f$ :

$$\Sigma + [f : \overline{m}_i^n \rightarrow s] + [x_i : \overline{m}_i^n] \vdash_{Dst} e_f : s$$

□

The  $\prod \Phi$  of the lemma above is, in fact, the *minimal* correct signature for the function definition. We have proved that, if a function is typeable, there exists a minimal mark signature for it. Our next step is to show that the inference algorithm shown in this chapter returns that minimal mark signature. The main part of the proof is a completeness result of the  $\vdash$  rules with respect to the  $\vdash_{Dst}$  type system. It states that, if an expression is typeable under some environment  $\Sigma + \Gamma$ , the  $\vdash$  rules are able to find some  $(R, D, S)$  sets such that the environment they define is lower than  $\Gamma$ .

**Theorem 5.18.** *If  $\Sigma + \Gamma \vdash_{Dst} e : s$ , then we can derive  $e \vdash_{\Sigma} (R, D, S)$  for some sets  $R, D$ , and  $S$ . Moreover, it holds that:*

$$[x : r \mid x \in R] + [x : d \mid x \in D] + [x : s \mid x \in S] \leq \Gamma$$

*Proof.* By induction on the structure of  $e$ . Let us denote by  $\Gamma[R, D, S]$  the environment occurring in the left-hand side of the equation above. Without loss of generality, we assume that the last  $\vdash_{Dst}$  rule applied is neither  $[EXT_{Dst}]$  nor  $[WEAK_{Dst}]$ . Otherwise, we can assume the existence of a  $\Sigma' \subseteq \Sigma$  and a  $\Gamma' \leq \Gamma$  such that  $\Sigma' + \Gamma' \vdash_{Dst} e : s$ . If we are able to derive  $e \vdash_{\Sigma'} (R, D, S)$ , it also holds that  $e \vdash_{\Sigma} (R, D, S)$ , and if  $\Gamma[R, D, S] \leq \Gamma'$ , then  $\Gamma[R, D, S] \leq \Gamma$ , as we want to prove. So, let us distinguish cases on the structure of the expression  $e$ :

- **Cases**  $e \equiv c, e \equiv x, e \equiv x @ r$  and  $e \equiv C \overline{a}_i^n @ r$ .

The result holds trivially in all these cases. For instance:

$$C \overline{a}_i^n @ r \vdash (\emptyset, \emptyset, \{\overline{a}_i^n\}) \quad \text{and} \quad \Gamma[\emptyset, \emptyset, \{\overline{a}_i^n\}] = [\overline{a}_i : \overline{s}^n] = \Gamma$$

- **Case**  $e \equiv f \overline{a}_i^n @ \overline{r}_j^m$ .

Firstly, let us prove that the side conditions in rule  $[APP_I]$  hold. Each one is proven by contradiction:

$$R \cap S = \emptyset$$

Assume that  $x \in R \cap S$ . Since  $x \in S$ , there exists an  $i \in \{1..n\}$  such that  $x = a_i$  and  $m_i = s$ , where  $\Sigma(f) = \overline{m}_i^n \rightarrow s$ . Hence,  $x \in \text{dom } \bigoplus_{i=1}^n [a_i : m_i]$ . On the other hand, and since  $x \in R$ ,  $x \in \text{dom } \Gamma_R$ , where  $\Gamma_R$  is defined as  $[y : r \mid y \in R]$ . This leads to a contradiction as, according to the  $[APP_{Dst}]$  rule,  $\Gamma_R$  and  $\bigoplus_{i=1}^n [a_i : m_i]$  have disjoint domains.

$$R \cap D = \emptyset$$

The proof is analogous to that of the previous condition.

$$S \cap D = \emptyset$$

If  $x \in S$  and  $x \in D_i$  for some  $i \in \{1..n\}$ , then  $x = a_i = a_j$  for some  $i, j \in \{1..n\}$  such that  $m_i = s$  and  $m_j = d$  (hence  $i \neq j$ ), but this contradicts the fact that  $\bigoplus_{i=1}^n [a_i : m_i]$  is well-defined, as the  $[APP_{Dst}]$  rule demands.

$$\forall i, j. i \neq j \Rightarrow D_i \cap D_j = \emptyset$$

Let  $x \in D_i$  and  $x \in D_j$  for some  $i \neq j$ . This implies that  $x = a_i = a_j$  for some  $i, j \in \{1..n\}$  such that  $m_i = m_j = d$ . Hence  $\bigoplus_{i=1}^n [a_i : m_i]$  in  $[APP_{Dst}]$  is not well-defined.

$\forall x \in D. isTree(x)$

If there is some  $x \in D$  such that  $isTree(x)$  does not hold, then there exists some  $i \in \{1..n\}$  such that  $x = a_i$  and  $m_i = d$ , which contradicts the  $\bigwedge_{m_i=d} isTree(a_i)$  property.

Besides this,  $\Sigma = [f : \overline{m_i^n} \rightarrow s]$  and:

$$\begin{aligned} \Gamma[R, D, S] &= [y : r \mid y \in R] + [a_i : s \mid m_i = s] + [a_i : d \mid m_i = d] \\ &= \Gamma_R + \bigoplus_{i=1}^n [a_i : m_i] \\ &= \Gamma \end{aligned}$$

• **Case  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$**

Assume  $\Sigma_1 + \Gamma_1 \vdash_{Dst} e_1 : s$ ,  $\Sigma_2 + \Gamma_2 + [x_1 : m_1] \vdash_{Dst} e_2 : s$ . By induction hypothesis we get  $e_1 \vdash_{\Sigma_1} (R_1, D_1, S_1)$  and  $e_2 \vdash_{\Sigma_2} (R_2, D_2, S_2)$ , which imply their corresponding  $\vdash_{\Sigma}$  judgements. We prove by contradiction that the side condition in  $[LET_I]$  holds. Assume  $x \in R_1 \cup D_1$  such that  $x \in fv(e_2)$ . Since  $\Gamma[R_1, D_1, S_1] \leq \Gamma_1$  holds by induction hypothesis, we get that  $\Gamma_1(x) \in \{r, d\}$ , and, by  $[LET_{Dst}]$ ,  $x \notin fv(e_2)$ , which contradicts our assumption. We can apply the  $[LET_I]$  rule, so  $e \vdash (R, D, S)$ .

Now we prove that  $\Gamma[R, D, S] \leq \Gamma_1 \sqcup \Gamma_2$ . Firstly,  $\Gamma[R_2, D_2, S_2] \leq \Gamma_2 + [x_1 : m_1]$  by induction hypothesis. From this, it follows that  $\Gamma[(R_2, D_2, S_2) \setminus x_1] \leq \Gamma_2$ . On the other hand, we get, for every pair of triples  $(R, D, S), (R', D', S')$ :

$$\begin{aligned} \Gamma[R, D, S] \sqcup \Gamma[R', D', S'] &= ([x : r \mid x \in R] + [x : d \mid x \in D] + [x : s \mid x \in S]) \sqcup \\ &\quad ([x : r \mid x \in R'] + [x : d \mid x \in D'] + [x : s \mid x \in S']) \\ &= [x : r \mid x \in R \cup R'] + [x : d \mid x \in (D \cup D') \setminus (R \cup R')] + \\ &\quad [x : s \mid x \in (S \cup S') \setminus (D \cup D' \cup R \cup R')] \\ &= \Gamma[(R, D, S) \sqcup (R', D', S')] \end{aligned}$$

Finally, we obtain:

$$\begin{aligned} \Gamma[R, D, S] &= \Gamma[(R_1, D_1, S_1) \sqcup ((R_2, D_2, S_2) \setminus \{x_1\})] \\ &= \Gamma[R_1, D_1, S_1] \sqcup \Gamma[(R_2, D_2, S_2) \setminus \{x_1\}] \\ &\leq \Gamma_1 \sqcup \Gamma_2 \end{aligned}$$

The last step justified by the fact that  $\sqcup$  is monotone w.r.t.  $\leq$ . The proof of this fact is straightforward and it shall be omitted here.

• **Case  $e \equiv \text{case } x \text{ of } \overline{C_i \overline{x_{ij}} \rightarrow e_i^n}$**

We know that, for each  $i \in \{1..n\}$ ,  $\Sigma + \Gamma' + [\overline{x_{ij} : \overline{m_{ij}}}] \vdash_{Dst} e_i : s$  for some  $\Gamma'$  and  $\overline{m_{ij}}$  such that  $\Gamma = \Gamma' \sqcup [x : s]$ . By induction hypothesis,  $e_i \vdash (R_i, D_i, S_i)$ , and  $\Gamma[R_i, D_i, S_i] \leq \Gamma' + [\overline{x_{ij} : \overline{m_{ij}}}]$ . The latter implies that  $\Gamma[(R_i, D_i, S_i) \setminus \{\overline{x_{ij}}\}] \leq \Gamma'$ . Hence we can apply  $[CASE_I]$  to get  $e \vdash (R, D, S)$ ,

where  $(R, D, S) = (\bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus \{\overline{x_{ij}}\}) \sqcup (\emptyset, \emptyset, \{x\})$ , and we get:

$$\begin{aligned}
\Gamma[R, D, S] &= \Gamma \left[ \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus \{\overline{x_{ij}}\}) \sqcup (\emptyset, \emptyset, \{x\}) \right] \\
&= \bigsqcup_{i=1}^n \Gamma[(R_i, D_i, S_i) \setminus \{\overline{x_{ij}}\}] \sqcup \Gamma[\emptyset, \emptyset, \{x\}] \\
&\leq \bigsqcup_{i=1}^n \Gamma' \sqcup [x : s] \\
&= \Gamma' \sqcup [x : s] \\
&= \Gamma
\end{aligned}$$

and the theorem holds.

- **Case  $e \equiv \text{case! } x \text{ of } \overline{C_i \overline{x_{ij}} \rightarrow e_i}^n$**

Again, let us assume that  $\Sigma + \Gamma' + [\overline{x_{ij} : m_{ij}}] \vdash_{Dst} e_i : s$ , for every  $i \in \{1..n\}$ , and for some  $\Gamma'$  such that  $\Gamma = \Gamma_R \sqcup (\Gamma' \setminus x) + [x : d]$ , where  $\Gamma_R$  is as defined in  $[CASE!_{Dst}]$ . By induction hypothesis, for every  $i \in \{1..n\}$ ,  $e_i \vdash (R_i, D_i, S_i)$  holds for some  $(R_i, D_i, S_i)$  such that  $\Gamma[R_i, D_i, S_i] \leq \Gamma' + [\overline{x_{ij} : m_{ij}}]$ . This implies, for every  $i \in \{1..n\}$ ,  $\Gamma[(R_i, D_i, S_i) \setminus \{\overline{x_{ij}}\}] \leq \Gamma'$ . Let us define

$$(R', D', S') = \left( \bigsqcup_{i=1}^n ((R_i, D_i, S_i) \setminus \{\overline{x_{ij}}, x\}) \right) \sqcup (R_{SHR}, \emptyset, \emptyset)$$

where  $R_{SHR}$  is as defined in  $[CASE!_I]$ . We obtain:

$$\Gamma[R', D', S'] = \left( \bigsqcup_{i=1}^n \Gamma[(R_i, D_i, S_i) \setminus \{\overline{x_{ij}}, x\}] \right) \sqcup \Gamma[R_{SHR}, \emptyset, \emptyset] \leq \Gamma_R \sqcup \Gamma' \setminus \{x\}$$

If we are able to apply  $[CASE!_I]$ , we get  $e \vdash (R, D, S)$ , where  $(R, D, S) = (R', D' \cup \{x\}, S')$ , and hence, since  $(R', D', S')$  does not contain the discriminant of the **case**:

$$\Gamma[R, D, S] = \Gamma[R', D', S'] + [x : d] \leq \Gamma_R \sqcup (\Gamma' \setminus \{x\}) + [x : d] = \Gamma$$

So, let us prove that the side conditions in  $[CASE!_I]$  hold. Again, we prove each of these by contradiction:

$$\forall i \in \{1..n\}. (P_i \setminus Rec_i) \cap (D_i \cup R_i) = \emptyset$$

Assume that  $x_{ij} \in P_i \setminus Rec_i$  (that is,  $j \notin RecPos(C_i)$ ) and that  $x_{ij} \in D_i \cup R_i$ . Since  $\Gamma[R_i, D_i, S_i] \leq \Gamma' + [x_{ij} : m_{ij}]$ , we get that  $m_{ij} \in \{d, r\}$ . But the condition  $inh_{j, C_i}(m_{ij})$  in  $[CASE!_{Dst}]$  demands  $m_{ij} = s$ .

$$\forall i \in \{1..n\}. Rec_i \cap R_i = \emptyset$$

Assume that  $x_{ij} \in Rec_i$  (that is,  $j \in RecPos(C_i)$ ). Similarly as above,  $x_{ij} \in R_i$  implies  $m_{ij} = r$ , which contradicts the  $inh_{j, C_i}(m_{ij})$  condition in  $[CASE!_{Dst}]$ , which demands  $m_{ij} = d$ .

$$\forall i \in \{1..n\}. (R_{SHR} \cup \{x\}) \cap fv(e_i) = \emptyset$$

Assume that  $z \in R_{SHR} \cup \{x\}$  and  $z \in fv(e_i)$  for some  $i \in \{1..n\}$ . This implies  $z \in R \cup \{x\}$  (being  $R$  as defined in  $[CASE!_{Dst}]$ ), since  $R = R_{SHR}$ . By the latter rule,  $z \notin fv(e_i)$ , which leads to a contradiction.

□

Given a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ , our algorithm performs several iterations of the  $\vdash$  rules until it reaches a mark signature that coincides with the one initially assumed for  $f$ . An iteration can be formalized as a function  $\varphi_f$  which, given a mark signature  $\bar{m}_i^n \rightarrow s$  returns the mark signature which would be given to the next iteration.

$$\varphi_f(\bar{m}_i \rightarrow s) = \bar{m}'_i \rightarrow s, \text{ where: } e_f \vdash_{\Sigma \uplus [f:\bar{m}_i \rightarrow s]} (\emptyset, D, S) \quad (5.7)$$

$$\forall i \in \{1..n\}. m'_i = \begin{cases} d & x_i \in D \\ s & x_i \notin D \end{cases}$$

The *inferMarksFP* algorithm stops when the mark signature is given for  $f$  is a fixed point of  $\varphi_f$ . A consequence of the previous theorem is that the minimal correct signature of a function is a fixed point of  $\varphi_f$ .

**Corollary 5.19.** *If  $\Phi$  is the set of correct signatures for  $f \bar{x}_i @ \bar{r}_j = e_f$ , then  $\bigcap \Phi \in \text{Fix}(\varphi_f)$ .*

*Proof.* Let  $\bigcap \Phi = \bar{m}_i \rightarrow s$  be the minimal signature. By Lemma 5.17 this signature is correct, so we get:

$$\underbrace{\Sigma + [f : \bar{m}_i \rightarrow s]}_{\Sigma'} + [\bar{x}_i : \bar{m}_i] \vdash_{Dst} e_f : s$$

where none of the  $m_i$  is an  $r$  mark. By Theorem 5.18 there exists some  $(R, D, S)$  such that  $e \vdash_{\Sigma'} (R, D, S)$  and  $[x : r \mid x \in R] + [x : d \mid x \in D] + [x : s \mid x \in S] \leq [\bar{x}_i : \bar{m}_i]$ . Since none of the  $m_i$  is an in-danger mark,  $R = \emptyset$ . Moreover, all the variables in the environment in the left-hand side must be one of the  $x_i$ . By Theorem 5.8:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [x_i : d \mid x_i \in D] + [x_i : s \mid x_i \in S] \vdash_{Dst} e_f : s$$

By applying  $[EXT_{Dst}]$  rule, we get:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [x_i : d \mid x_i \in D] + [x_i : s \mid x_i \notin D] \vdash_{Dst} e_f : s \quad (5.8)$$

Let us define, for each  $i \in \{1..n\}$ :

$$m'_i = \begin{cases} d & x_i \in D \\ s & x_i \notin D \end{cases}$$

so that  $\varphi_f(\bar{m}_i^n \rightarrow s) = \bar{m}'_i \rightarrow s$ . We can transform (5.8) into the following:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [\bar{x}_i : m'_i] \vdash_{Dst} e_f : s$$

Let us prove that  $m'_i \leq m_i$  for each  $i \in \{1..n\}$ . If  $m'_i = s$  we are done. If  $m'_i = d$ , then  $x_i \in D$ , and hence  $m_i = d$ , since  $[x : d \mid x \in D] + [x : s \mid x \in S] \leq [\bar{x}_i : \bar{m}_i]$ . Hence,  $\bar{m}'_i \rightarrow s \leq \bar{m}_i \rightarrow s$ , and we can apply Lemma 5.16 to the previous judgement:

$$\Sigma + [f : \bar{m}'_i \rightarrow s] + [\bar{x}_i : m'_i] \vdash_{Dst} e_f : s$$

Hence  $\bar{m}'_i \rightarrow s$  is a correct signature for  $f$  that is lower than  $\bar{m}_i \rightarrow s$ . However, recall that  $\bar{m}_i \rightarrow s$  is the greatest lower bound, so  $m'_i = m_i$  for every  $i \in \{1..n\}$ , and hence  $\varphi_f(m_i \rightarrow s) = m_i \rightarrow s$ . □

Reciprocally, every fixed point of  $\varphi_f$  is a correct mark signature.

**Lemma 5.20.** *If  $\Phi$  is the set of correct signatures for  $f \bar{x}_i @ \bar{r}_j = e_f$ , then  $\text{Fix}(\varphi_f) \subseteq \Phi$ .*

*Proof.* If  $\varphi_f(\bar{m}_i \rightarrow s) = \bar{m}_i \rightarrow s$ , we get  $e_f \vdash_{\Sigma \cup [f:\bar{m}_i \rightarrow s]} (\emptyset, D, S)$  for some  $D$  and  $S$ . By Theorem 5.8, we get:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [x : d \mid x \in D] + [x : s \mid x \in S] \vdash_{Dst} e_f : s$$

Since  $D \cup S \subseteq \text{fv}(e_f) \subseteq \{\bar{x}_i\}$ , we can apply the  $[EXT_{Dst}]$  rule in order to get:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [x_i : d \mid x_i \in D] + [x_i : s \mid x_i \notin D] \vdash_{Dst} e_f : s$$

By the definition of  $m'_i$  in (5.7) we can transform the judgement above into the following:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [x_i : d \mid m_i = d] + [x_i : s \mid m_i = s] \vdash_{Dst} e_f : s$$

which is equivalent to:

$$\Sigma + [f : \bar{m}_i \rightarrow s] + [\bar{x}_i : \bar{m}_i] \vdash_{Dst} e_f : s$$

Therefore,  $\bar{m}_i \rightarrow s$  is a correct signature for  $f$ . □

From the last two results we can prove that the inference algorithm returns the minimal correct type signature. It is straightforward to see that *inferMarksDef* returns the least fixed point of  $\varphi_f$ , since it performs the computation of a Kleene's ascending chain, starting from the bottommost element  $\bar{s} \rightarrow s$ . On the one hand, the least fixed point of  $\varphi_f$  (denoted  $\text{lfp}(\varphi_f)$ ) is a correct signature by Lemma 5.20, so  $\text{lfp}(\varphi_f) \geq \sqcap \Phi$ , where  $\Phi$  denotes the set of correct signatures for  $f$ . On the other hand,  $\sqcap \Phi$  is a fixed point, so  $\text{lfp}(\varphi_f) \leq \sqcap \Phi$ . As a consequence,  $\text{lfp}(\varphi_f) = \sqcap \Phi$ , which proves that the *inferMarksDef* computes the minimal mark signature for  $f$ .

#### 5.4.4 Efficiency

Assume we execute the *inferMarksDef* function on a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . If  $s$  denotes the size of the AST corresponding to  $e_f$ , a single traversal of the tree has  $\mathcal{O}(s)$  complexity if we assume constant-time set operations. However, the complexity of set operations  $\cup, \cap, \setminus$  is linear with respect the sizes of the sets to which they are applied. In our case, these sets contain variables occurring in  $e_f$ . Since the worst-case scenario assumes a number of variables in the AST proportional to  $s$ , a single traversal of the tree has a worst-case  $\mathcal{O}(s^2)$  time complexity. Besides this, the algorithm performs, in the worst case, as many iterations as the number of parameters  $n$ . As a consequence, the worst-case time complexity of the algorithm for inferring the mark signature of a function definition is in  $\mathcal{O}(ns^2)$ . Nevertheless, all the examples shown in this chapter only require two iterations: The first one computes the fixed point, whereas the second one checks that the result is a fixed point. Thus we expect that the average-case complexity of our algorithm is  $\mathcal{O}(s^2)$ .

### 5.5 Conclusions and related work

The mark inference algorithm introduced in this chapter represents a notable improvement with respect to its predecessor, described in [81]. The main reason behind this fact is the definition of a total order on the different marks, which allows the algorithm to assign the strongest mark to a variable at a given

context, and to weaken that mark in a higher context, if necessary. This results in a simpler algorithm, in which a single bottom-up traversal is done in each iteration. Previous work required some interleaving of bottom-up and top-down traversals with two different set of rules. As a consequence, we get a simpler correctness proof. Moreover, the distinction between strongest and weakest marks allows the  $\vdash_{Dst}$  rules to have a minimal typing environment under which a given expression is typeable (Lemma 5.17). This allows us to prove the completeness of our algorithm: if an expression is typeable, the algorithm finds the minimal environment typing that expression.

The examples shown in Section 3.5 have been typed by assuming the most precise analysis satisfying the correctness conditions of Section 3.6.3. On the contrary, in this chapter we have used the sharing analysis of [98]. As we have shown in Section 5.3, the sharing information has to be adjusted in some cases. This motivates the need for a more precise analysis, whose study is subject of future work.

In Section 3.8 we showed the relation between our type system and other approaches related to functional languages. In particular, a detailed comparison was done with Aspinall and Hofmann’s type system [11] based on usage aspects (UAPL in the following). An usage aspect inference algorithm for this system is due to Konečný [68]. It is worth highlighting the similarities and differences of his inference algorithm with ours.

- The set of typing rules is, in both cases, non-deterministic. Our inference algorithm is given by a set of syntax-directed rules, whereas UAPL features a deterministic type-checking strategy in which some rules are given more (or less) priority than the remaining ones.
- In UAPL, the (RAISE) rule takes the highest priority. This rule upgrades the usage aspects of the variables in the type environment from 2 (not destroyed, but shares with result) to 3 (not destroyed, does not share with result) in those cases when the result is of a basic (i.e. heap-free) type. This is not necessary in our type system, since the sharing analysis of *Safe* is aware of the fact that values of basic types do not take space in the heap, so they cannot share anything with the remaining variables. Moreover, all basic types have a safe mark in our system.
- The (WEAK) and (DROP) rules of UAPL, which are similar to the [EXT] and [WEAK] rules of *Safe*, are given the lowest priority, and are only used to make the premises in the same rule match each other. In *Safe*, the  $\sqcup$  operator on triples has a similar goal.
- The inference of the signature of recursive function definitions follows a similar pattern in both cases. It starts by assuming the strongest signature for the function being inferred, and performs a fixed point iteration. This is not very surprising, as this approach (known as the Kleene’s ascending chain) is commonly used when the set of function signatures has the structure of a complete partial order. In particular, the sharing analysis of [98] also works in this way.

Although the intended semantics of safe, condemned, and in-danger types is substantially different from the usage aspects of UAPL, the fact of defining a total order between the marks/usage aspects makes the inference algorithms of UAPL and *Safe* very similar. A contribution of our algorithm with respect to UAPL’s is the explicit formalization of the algorithm itself, together with a comprehensive proof of its correctness and completeness.



## Chapter 6

# Certified absence of dangling pointers

### 6.1 Introduction

Let us briefly recap the results of last chapters. On the one hand, we have described in Chapter 4 an algorithm for annotating a *Safe* program with region variables, such that the resulting program is well-typed w.r.t. the  $\vdash_{Reg}$  rules. On the other hand, we have developed in Chapter 5 an algorithm that checks whether the destruction of cells, managed by the programmer, does not lead to an ill-checked program regarding the  $\vdash_{Dst}$  rules. If a *Safe* program is accepted by both algorithms, we can ensure, by the correctness proof given in Chapter 3, that this program is pointer-safe. The aim of this chapter is to formalize this property and its corresponding proof for our specific program, so both things can be mechanically checked. This is an example of a *proof-carrying code* scheme [88], in which the compiler generates, besides the target program, a proof (*certificate*) of the presence of a property which the code consumer expects to hold. Both the program and the certificate are sent to the code consumer, which checks the proof against the program before executing the latter.

We have used the Isabelle proof assistant for encoding all these properties and their proofs. Isabelle provides a meta-logic in which Higher-order logic (HOL) is specified. Both things are commonly referred to with the name *Isabelle/HOL*. This tool provides a framework in which we can specify definitions, theorems, and their proofs. Isabelle/HOL checks whether the theorems actually follow from the given proofs. In this chapter we explain how to generate Isabelle/HOL scripts encoding the above mentioned pointer-safety properties. It is not surprising that most of the proofs and definitions relevant to this chapter are those occurring in Section 3.6, where the pointer-safety of well-typed programs was established. However, the proof rules of this chapter are based on the original *Safe*'s type system described in [84], from which the type system of this thesis stems. As a consequence, there are some slight differences in the certificate w.r.t. the proof shown in this thesis. It is subject of future work to adapt the proofs of this chapter so as to fit those of Section 3.6.

This chapter is based on the work described in [36]. Unlike previous chapters, it is more oriented towards implementation issues. In particular we explain in detail the automatic generation of Isabelle/HOL scripts, since this is the main contribution of the author of this thesis to [36]. More details about the formalization itself can be found in [37], or in de Dios' PhD thesis [35] (the latter in Spanish). However, we briefly describe in Section 6.2 the properties to be proved, and the information that the certificate is expected to contain for a specific program. Section 6.3 deals with the generation of the certificate itself. In Section 6.4 we measure, by means of some case studies, the length of the obtained proofs w.r.t. the original *Core-Safe* code. Finally, Section 6.5 concludes.

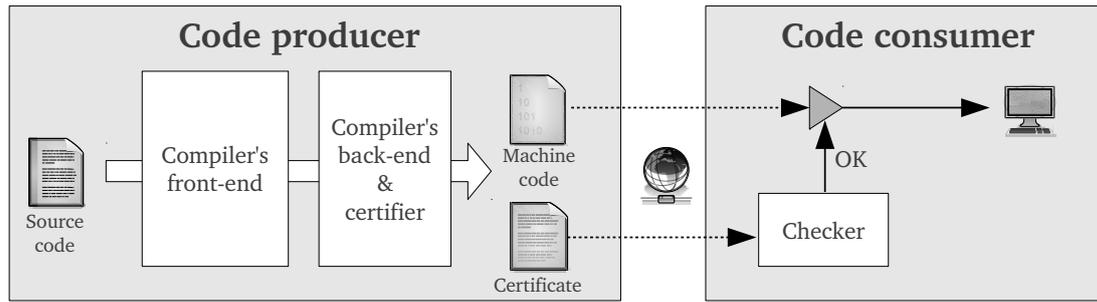


Figure 6.1: Standard PCC paradigm. The certificate refers to the low-level machine code, which is sent to the consumer.

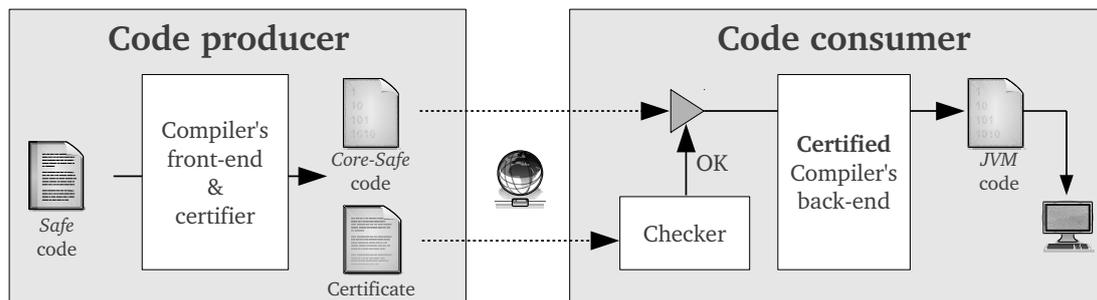


Figure 6.2: Our PCC paradigm.

## 6.2 Preliminaries

The standard approach to PCC is depicted in Figure 6.1. In this scenario, the compiler attaches a certificate to the resulting machine code, and the proof is performed at machine level. That is, the properties being certified are checked against the machine code. The code consumer receives both low-level code and certificate and checks the validity of the latter. If it is correct, it allows the execution of the code on the target machine. A variation of this scheme advances the generation of the certificate before the translation into machine code, but after the front-end phase, in which the properties of interest are known to hold. In this case, the certificate refers to the intermediate representation resulting from the front-end, and the back-end translates both this intermediate representation and the certificate into a lower-level representation, which is sent to the code consumer. The main drawback of these approaches is that the involved certificates are of considerable size, which results in longer checking times.

Our approach consists in postponing the machine code generation after the certificate is checked (Figure 6.2). The compiler generates a certificate proving the pointer-safety properties of a *Core-Safe* program. After the consumer has checked its validity against the latter, the program is translated into Java bytecode and executed. Since the certificates obtained in this way refer to the properties of a higher-level language (*Core-Safe*) they are smaller and easier to check than their lower-level counterparts. They also save the effort of translating the certificate into a lower-level variant. However, this approach relies on the correctness of the compiler's back-end that translates *Core-Safe* code into Java byte-code. There are two possibilities:

1. Assume the correctness of this translation phase. In this case, the back-end is said to be part of the *trusted code base*.

## 2. Mechanically prove and certify these translation phases.

In [39] and [38] the authors prove in Isabelle/HOL, respectively, the correctness of the translation from *Core-Safe* into SVM code (see Section 2.6) and its further translation into Java bytecode. The definitions of these translations are specified in Isabelle/ML (a functional, ML-alike language), and a Haskell implementation is automatically extracted from these definitions. Thus, if the Isabelle/HOL definitions are proven correct, so is their corresponding implementation. A drawback of this approach to PCC is that the consumer receives the *Core-Safe* code, which it may be easier to reverse-engineer than its bytecode equivalent. Besides this, part of the compiling process (translation into bytecode) is delegated to the code consumer. This is a non-standard approach in compilers.

Once we know the level at which the certificate is going to be targeted, the next step is to define the contents of the certificate (i.e. *what* to prove). Given a *Core-Safe* program *prog*, the certificate is expected to prove the following fact:

$$prog \text{ does not access dangling pointers} \tag{6.1}$$

However, it would be a very poor solution if each certificate contained the full proof of this fact, since several parts of this proof could also be applied to every *Core-Safe* program. Since we have devised a generic way (namely, a type system) to prove a program pointer-safe, it seems sensible to split the proof of (6.1) into two different facts:

$$\text{Well-typed programs do not access dangling pointers} \tag{6.2}$$

$$prog \text{ is a well-typed program} \tag{6.3}$$

It is trivial to see that (6.1) follows from these facts. The first one is generic, whereas the second one is program-specific. The idea of the certificate generation is to ask the compiler to deliver the static information inferred during the type inference phase (Chapters 4 and 5) and to use this information for proving (6.3). This can be done with a set of *proof rules*. There exists a proof rule for every syntactical construct of the language. Given an expression of the program, its certificate must ensure that the premises of a proof rule hold before applying that rule. These premises are known as *proof obligations*. The fact (6.2) is proved by relating the static information (types) with the dynamic properties a program is expected to satisfy at runtime (absence of dangling pointers). For every proof rule there exists a lemma establishing this relation. These lemmas are proved once for all, and can be stored in the code consumer side, since they are common to all *Core-Safe* programs. The program-specific part of the certificate consists in establishing the validity of the necessary proof obligations for the given program, and applying the corresponding proof rules. Since there are general theorems relating these proof rules with the safety properties of the language, the specific program can be proved pointer-safe by applying these theorems. This strategy results in small certificates and short checking times, as the total amount of work is linear w.r.t. the size of the program. The heaviest part of the proof (namely, the database of proved proof rules) has been done in advance and is reused by each certified program.

Given the above, our first step is to present these proof rules with their corresponding safety properties. There is some resemblance with the typing rules of Chapter 3, but this resemblance becomes more apparent when comparing the proof rules with those in [84]. In the same way we have considered the inference of regions and marks separately, the certificates also distinguishes the part of pointer safety regarding implicit region deallocation from that related to explicit destruction. Thus we have two sets of proof rules, which are described in next sections.

## 6.2.1 Rules regarding region deallocation

The aim of this part of the certificate is to prove the preservation of the consistency property through the execution of a program. Recall from Section 3.6.1 the existence of a region instantiation  $\eta$  associating RTVs with actual runtime region numbers. If a DS is inferred to have  $\rho$  as the outermost RTV in its type, and this DS lives in region  $n$  at runtime, then  $\eta(\rho) = n$ . Consistency property tells us that the correspondence between the static region types, and the actual regions where the data structures are stored in the heap, do not contradict  $\eta$ . This part of the certificate proves that, if the initial configuration of a program is consistent, so is every intermediate configuration in its execution, and so is the final configuration. Since the execution of a program starts with an empty heap, which is trivially consistent, all the intermediate configurations are also consistent. This part of the certificate deals exclusively with regions, and not with the condemned, safe or in-danger mark of a type, so the proof rules roughly resemble the  $\vdash_{Reg}$  typing rules of Section 3.7. Thus we can instantiate the facts (6.2) and (6.3) as follows:

$$\vdash_{Reg}\text{-typed programs preserve consistency along their execution} \quad (6.4)$$

$$prog \text{ is a } \vdash_{Reg}\text{-typed program} \quad (6.5)$$

This implies that the deallocation of the topmost region of the heap cannot create dangling pointers at runtime, since  $\eta$  always maps the  $\rho_{self}$  region type to the topmost region, and this RTV cannot appear in the signature of a function, so neither the input nor the output of the function will be located in that region.

In Figure 6.3 we show the proof rules. We have typing environments  $\Gamma_R$  mapping program variables to safe types and region arguments to region types. When mechanically checking the certificate, it is convenient to have the types of already certified functions into a separate environment  $\Sigma_R$ , and a global environment  $\Gamma_C$  giving the polymorphic most general type of data constructors. Hence  $\Gamma_R$  only contains variables in scope. In the rules we have judgements of the form  $e, \Sigma_R \vdash \Gamma_R \rightsquigarrow s$ , where  $e \in \mathbf{Exp}$  and  $s \in \mathbf{SafeType}$ . They have a similar meaning to the judgement  $\Gamma_R \cup \Sigma_R \cup \Gamma_C \vdash_{Reg} e : s$ .

In addition to the  $\Gamma_R$  and  $s$ , which are provided by the compiler, the certificate needs additional information in function and constructor applications. Recall that the types occurring in these expressions may be instances of the signature of the corresponding function/constructor, so the compiler generates a *type instantiation mapping*  $\mu$  in these cases. The certificate checks whether this mapping correctly associate the types of the formal parameters with the types of the corresponding actual arguments in the function/constructor application. This motivates the following definition.

**Definition 6.1.** Given the instantiated types  $\bar{s}_i^n$ , the instantiated region types  $\bar{\rho}_j^m$ , the arguments of the application  $\bar{a}_i^n, \bar{r}_j^m$  and a region typing environment  $\Gamma_R$ , we say that the application is *argument preserving*, denoted  $argP(\bar{s}_i^n, \bar{\rho}_j^m, \bar{a}_i^n, \bar{r}_j^m, \Gamma_R)$  iff  $\forall i \in \{1..n\}. s_i = \Gamma_R(a_i)$  and  $\forall j \in \{1..m\}. \rho_j = \Gamma_R(r_j)$

Thus, given the generic type scheme  $\forall \bar{\alpha} \bar{\rho}. \bar{s}_i \rightarrow \bar{\rho}_j \rightarrow s$  of a function  $g$ , and a specific application  $g \bar{a}_i @ \bar{r}_j$ , the type instantiation mapping  $\mu$  is valid provided  $argP(\mu(\bar{s}_i), \mu(\bar{\rho}_j), \bar{a}_i, \bar{r}_j, \Gamma_R)$  holds. This is what rules  $[APP_{RegC}]$  and  $[CONS_{RegC}]$  ensure.

The certificate incrementally constructs a global environment  $\Sigma_R$  keeping the most general types of the functions already certified. Regarding constructors, the certificate demands a well-formedness condition on their types, namely, that the type of the only region argument of a constructor is the outermost RTV of the type being constructed, and that all the RTVs occurring in the type of the arguments of a constructor also occur in the type of its result.

$$\begin{array}{c}
\frac{}{c, \Sigma_R \vdash \Gamma_R \rightsquigarrow B} [LIT_{RegC}] \\
\\
\frac{}{x, \Sigma_R \vdash \Gamma_R \rightsquigarrow \Gamma_R(x)} [VAR_{RegC}] \\
\\
\frac{\Gamma_R(x) = T \overline{s_i^m} @ \rho_1 \dots \rho_l \quad \Gamma_R(r) = \rho'}{x @ r, \Sigma_R \vdash \Gamma_R \rightsquigarrow T \overline{s_i^m} @ \rho_1 \dots \rho_{l-1} \rho'} [COPY_{RegC}] \\
\\
\frac{\Gamma_C(C) = \overline{s_i^n} \rightarrow \rho \rightarrow s_C \quad wellT(\overline{s_i^n}, \rho, s_C) \quad argP(\overline{\mu(s_i)^n}, \mu(\rho), \overline{a_i^n}, r, \Gamma_R) \quad s = \mu(s_C)}{C \overline{a_i^n} @ r, \Sigma_R \vdash \Gamma_R \rightsquigarrow s} [CONS_{RegC}] \\
\\
\frac{\Sigma_R(g) = \overline{s_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s_g \quad \rho_{self}^g \notin regions(s_g) \quad argP(\overline{\mu(s_i)^n}, \overline{\mu(\rho_j)^m}, \overline{a_i^n}, \overline{r_j^m}, \Gamma_R) \quad s = \mu(s_g)}{g \overline{a_i^n} @ \overline{r_j^m}, \Sigma_R \vdash \Gamma_R \rightsquigarrow s} [APP_{RegC}] \\
\\
\frac{e_1, \Sigma_R \vdash \Gamma_R \rightsquigarrow s_1 \quad e_2, \Sigma_R \vdash \Gamma_R \uplus [x_1 \mapsto s_1] \rightsquigarrow s_2}{\mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2, \Sigma_R \vdash \Gamma_R \rightsquigarrow s_2} [LET_{RegC}] \\
\\
\frac{\forall i. \Sigma_R(C_i) = \overline{s_{ij}^{n_i}} \rightarrow \rho \rightarrow s \quad \forall i. wellT(\overline{s_{ij}^{n_i}}, \rho, s) \quad \forall i. e_i, \Sigma_R \vdash \Gamma_R \uplus [\overline{x_{ij}} \rightarrow \mu(s_{ij})^{n_i}] \rightsquigarrow s' \quad \Gamma_R(x) = \mu(s)}{\mathbf{case} \ x \ \mathbf{of} \ \overline{C_i} \ \overline{x_{ij}^{n_i}} \rightarrow e_i^n, \Sigma_R \vdash \Gamma_R \rightsquigarrow s'} [CASE_{RegC}] \\
\\
\frac{\forall i. \Sigma_R(C_i) = \overline{s_{ij}^{m_i}} \rightarrow \rho \rightarrow s \quad \forall i. wellT(\overline{s_{ij}^{m_i}}, \rho, s) \quad \forall i. e_i, \Sigma_R \vdash \Gamma_R \uplus [x_{ij} \rightarrow \mu(s_{ij})^{m_i}] \rightsquigarrow s' \quad \Gamma_R(x) = \mu(s)}{\mathbf{case!} \ x \ \mathbf{of} \ \overline{C_i} \ \overline{x_{ij}^{m_i}} \rightarrow e_i^n, \Sigma_R \vdash \Gamma_R \rightsquigarrow s'} [CASE!_{RegC}] \\
\\
\frac{\Gamma_R = [\overline{x_i} \mapsto \overline{s_i^n}, \overline{r_j} \mapsto \overline{\rho_j^m}, self \mapsto \rho_{self}] \quad e_f, \Sigma_R \uplus \{f \mapsto \overline{s_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s_f\} \vdash \Gamma_R \rightsquigarrow s_f \quad f \overline{x_i^n} @ \overline{r_j^m} = e_f}{e_f, \Sigma_R \vdash \Gamma_R \rightsquigarrow s_f} [REC_{RegC}]
\end{array}$$

Figure 6.3: Proof rules regarding regions.

**Definition 6.2.** Predicate  $wellT(\overline{s}_i^n, \rho, s)$ , read *well-typed*, is defined as follows:

$$wellT(\overline{s}_i^n, \rho, T \overline{s}_i^m @ \overline{\rho}_j) \equiv \rho_m = \rho \wedge \bigcup_{i=1}^n regions(s_i) \subseteq regions(T \overline{s}_i^m @ \overline{\rho}_j)$$

The first phase of the region inference algorithm (Section 4.3) generates type schemes for constructors that always satisfy this property. There is an additional rule  $[REC_{RegC}]$  for typing the body of a function definition, which is similar to the  $[FUN]$  rule of Figure 3.9. This rule states that whenever we are able to prove a judgement  $e_f, \Sigma_R \vdash \Gamma_R \rightsquigarrow s_f$  by assuming that  $\Sigma_R$  contains a type scheme for  $f$  matching the contents of  $\Gamma_R$ , we can do the same without that assumption. The correctness proof of this rule is technically involved and we refer to [37] for more details on it.

So far for the static concepts aimed at proving (6.5). We move now to the dynamic or runtime ones in order to prove (6.4). First we define the notion of admissible region instantiation.

**Definition 6.3.** Assume  $k$  denotes the topmost region of a given heap. We say that the region instantiation  $\eta$  is admissible, denoted  $admissible(\eta, k)$  iff:

$$\rho_{self}^f \in \text{dom } \eta \wedge \eta(\rho_{self}^f) = k \wedge \forall \rho \in \text{dom } \eta \setminus \{\rho_{self}^f\}. \eta(\rho) < k$$

Basically, an admissible  $\eta$  maps the type of the *self* region to the topmost region in the heap, and the rest of the RTVs to the regions situated below it. The fact that a region instantiation connects the static information  $\Gamma_R$  with the dynamic one  $(E, h)$  is formalised by means of the important notion of consistency, defined as follows:

**Definition 6.4.** We say that the mappings  $\Gamma_R$ ,  $\eta$ , the runtime environment  $E$ , and the heap  $h$  are *consistent*, denoted  $consistent(\Gamma_R, \eta, E, h)$ , if:

1.  $\forall x \in \text{dom } E. consistent(\Gamma_R(x), \eta, E(x), h)$  where:
 

$consistent(B, \eta, c, h)$	$=$	$true$
$consistent(\alpha, \eta, v, h)$	$=$	$true$
$consistent(s, \eta, p, h)$	$\Leftarrow$	$p \notin \text{dom } h$
$consistent(T \overline{s}_i^m @ \overline{\rho}_j^l, \eta, p, h)$	$\Leftarrow$	$\exists j \ C \ \overline{v}_k^n \ \mu \ \overline{s}_{kC}^n \ \overline{\rho}_{jC}^l. h(p) = (j, C \ \overline{v}_k^n)$ $\wedge \ \rho_l \in \text{dom } \eta \wedge \eta(\rho_l) = j$ $\wedge \ \Gamma_C(C) = \overline{s}_{kC}^n \rightarrow \rho_{lC} \rightarrow T \overline{s}_{iC}^m @ \overline{\rho}_{jC}^l$ $\wedge \ \mu(T \overline{s}_{iC}^m @ \overline{\rho}_{jC}^l) = T \overline{s}_i^m @ \overline{\rho}_j^l$ $\wedge \ \forall k \in \{1..n\}. consistent(\mu(s_{kC}), \eta, v_k, h)$
2.  $\forall r \in \text{dom } E. \Gamma_R(r) \in \text{dom } \eta \wedge E(r) = (\eta \circ \Gamma_R)(r)$
3.  $self \in \text{dom } E \wedge \Gamma_R(self) = \rho_{self}^f$

Notice the similarity with the notion of consistency introduced in Section 3.6.1. In fact,  $consistent(\Gamma_R, \eta, E, h)$  holds if and only if  $\eta$  is consistent with  $build^*(h, E, \Gamma_R)$ .

We are ready to specify (6.4) by means of a *static assertion* relating the static and dynamic properties referred to regions. A judgement of the form  $e : \llbracket \Gamma_R, s \rrbracket$  defines that, if the expression  $e$  is evaluated with an environment  $E$ , a heap  $h$  with  $k$  regions, and an admissible mapping  $\eta$  consistent with  $\Gamma_R$ , then the final heap  $h'$  and the final value  $v$  are consistent with  $\eta$ . Formally:

**Definition 6.5.** Let us define the following predicates:

$$\begin{aligned}
PR1 &\equiv E \vdash h, k, e \Downarrow h', k, v \\
PR2 &\equiv \text{dom } E \subseteq \text{dom } \Gamma_R \\
PR3 &\equiv \text{admissible}(\eta, k) \\
PR4 &\equiv \text{consistent}(\Gamma_R, \eta, E, h) \\
PR5 &\equiv \text{consistent}(s, \eta, v, h')
\end{aligned}$$

An expression  $e$  satisfies the pair  $(\Gamma_R, s)$ , denoted  $e : \llbracket \Gamma_R, s \rrbracket$  iff:

$$\forall E, h, k, h', v, \eta. PR1 \wedge PR2 \wedge PR3 \wedge PR4 \rightarrow PR5$$

Notice that this definition only ensures the consistency of the final configuration assuming the consistency of the initial one, but it does not prevent us from having a non-consistent intermediate configuration. The following theorem states that this does not happen if the initial configuration is admissible and consistent.

**Theorem 6.6** (consistency). *Let us assume a  $\Downarrow$ -judgement:*

$$E \vdash h, k, e \Downarrow h', k, v \tag{6.6}$$

For every  $\Gamma_R$  and  $s$  such that  $e : \llbracket \Gamma_R, s \rrbracket$ : if  $PR2(E, \Gamma_R)$ ,  $PR3(\eta, k)$ ,  $PR4(\Gamma_R, \eta, E, h)$  hold, then for every judgement  $E_i \vdash h_i, k_i, e_i \Downarrow h'_i, k_i, v_i$  in the derivation of (6.6), it holds that  $PR3(\eta_i, k_i)$ ,  $PR4(\Gamma_{R,i}, \eta_i, E_i, h_i)$  and  $PR5(s_i, \eta_i, v_i, h'_i)$ .

*Proof.* It follows from the proof of Theorem 3.26 (see [35]). □

We know that  $PR2$ ,  $PR3$  and  $PR4$  trivially hold for the initial state with the empty heap, the initial runtime environment  $[self \mapsto 0]$  and the initial region type environment  $[self : \rho_{self}]$ , so  $PR3$ ,  $PR4$  and  $PR5$  hold across the whole program derivation.

The following theorem relates the proof rules of Figure 6.3 to the assertions defined in Definition 6.5.

**Theorem 6.7** (soundness). *If  $e, \Sigma_R \vdash \Gamma_R \rightsquigarrow s$ , then  $e, \Sigma_R : \llbracket \Gamma_R, s \rrbracket$ .*

*Proof.* It follows from Theorem 3.26 (see [35]). □

## 6.2.2 Rules regarding explicit deallocation

This part of the certificate deals with the preservation of the closedness property of the heap in presence of explicit destruction. A heap is said to be closed with respect an expression  $e$  if it is not possible to reach a dangling pointer from the variables occurring free in  $e$ . In a similar way to the consistency property, the certificate proves that, if the initial heap is closed, so is the final heap and the intermediate heaps taking place in the evaluation of the expression. The initial (empty) heap is always closed w.r.t. the main expression of the program (as the latter does not have free variables), so this fact proves the preservation of closedness during the whole execution of the program. Again, this part of the certificate involves proving the following facts:

$$\text{The heap remains closed as } \vdash_{Dst}\text{-typed programs are executed} \tag{6.7}$$

$$prog \text{ is a } \vdash_{Dst}\text{-typed program} \tag{6.8}$$

The proof rules are shown in Figure 6.4. In this case, the static information consists of a *mark environment*  $\Gamma_D$  assigning each variable in scope a mark in the set  $\{d, r, s\}$ , as in the  $\vdash_{Dst}$  type system. Again,

$$\begin{array}{c}
\frac{}{c, \Sigma_D \vdash (\emptyset, \emptyset)} [LIT_{DstC}] \\
\\
\frac{}{x, \Sigma_D \vdash (\{x\}, \Gamma_D + [x : s])} [VAR_{DstC}] \\
\\
\frac{x \in \text{dom } \Gamma_D \quad \Gamma_D \text{ well formed}}{x @ r, \Sigma_D \vdash (\{x\}, \Gamma_D)} [COPY_{DstC}] \\
\\
\frac{\Gamma_D = [\bar{a}_i : \bar{s}^n]}{C \bar{a}_i^n @ r, \Sigma_D \vdash (\{\bar{a}_i^n\}, \Gamma_D)} [CONS_{DstC}] \\
\\
\frac{\Sigma_D(g) = \bar{m}_i^n \rightarrow s \quad \Gamma_{D,0} = \bigoplus_{i=1}^n [a_i : m_i] \text{ defined} \quad \Gamma_D \supseteq \Gamma_{D,0} \quad \Gamma_D \text{ well formed}}{g \bar{a}_i^n @ \bar{r}_j^m, \Sigma_D \vdash (\{\bar{a}_i^n\}, \Gamma_D)} [APP_{DstC}] \\
\\
\frac{e_1, \Sigma_D \vdash (L_1, \Gamma_{D,1}) \quad x_1 \notin L_1 \quad e_2, \Sigma_D \vdash (L_2, \Gamma_{D,2} + [x_1 : m]) \quad \text{def}(\Gamma_{D,1} \triangleright^{L_2} \Gamma_{D,2}) \quad m \in \{d, s\}}{\text{let } x_1 = e_1 \text{ in } e_2, \Sigma_D \vdash (L_1 \cup (L_2 \setminus \{x_1\}), \Gamma_{D,1} \triangleright^{L_2} \Gamma_{D,2})} [LET_{DstC}] \\
\\
\frac{\forall i. (e_i, \Sigma_D \vdash (L_i, \Gamma_{D,i}) \wedge \forall j. \Gamma_{D,i}(x_{ij}) \neq d) \quad \Gamma_D \text{ well-formed} \quad \Gamma_D \supseteq \bigotimes_i (\Gamma_{D,i} \setminus \{\bar{x}_{ij}\}) \quad x \in \text{dom } \Gamma_D \quad L = \{x\} \cup (\bigcup_i (L_i \setminus \{\bar{x}_{ij}\}))}{\text{case } x \text{ of } \bar{C}_i \bar{x}_{ij} \rightarrow e_i, \Sigma_D \vdash (L, \Gamma_D)} [CASE_{DstC}] \\
\\
\frac{\forall i. (e_i, \Sigma_D \vdash (L_i, \Gamma_{D,i}) \quad \forall j. \Gamma_{D,i}(x_{ij}) = d \rightarrow j \in \text{RecPos}(C_i)) \quad \Gamma_D \text{ well formed} \quad L = \bigcup_i (L_i \setminus \{\bar{x}_{ij}\}) \quad \Gamma_D \supseteq (\bigotimes_i \Gamma_{D,i} \setminus (\{\bar{x}_{ij}\} \cup \{x\})) + [x : d] \quad \forall z \in \text{dom } \Gamma_D. \Gamma_D(z) \neq s \rightarrow (\forall i. z \notin L_i)}{\text{case! } x \text{ of } \bar{C}_i \bar{x}_{ij} \rightarrow e_i, \Sigma_D \vdash (L \cup \{x\}, \Gamma_D)} [CASE!_{DstC}] \\
\\
\frac{f \bar{x}_i^n @ \bar{r}_j^m = e_f \quad L = \{\bar{x}_i^n\} \quad \Gamma_D = [\bar{x}_i \mapsto \bar{m}_i^n] \quad e_f, \Sigma_D \uplus [f \mapsto \bar{m}_i^n \rightarrow s] \vdash (L, \Gamma_D)}{e_f, \Sigma_D \vdash (L, \Gamma_D)} [REC_{DstC}]
\end{array}$$

Figure 6.4: Proof rules for explicit deallocation.

we store the mark signatures of functions in a separate environment  $\Sigma_D$ . A judgement  $e, \Sigma_D \vdash (L, \Gamma_D)$  means that  $L$  is the set of free variables in  $e$ , and that  $\Sigma_D \cup \Gamma_D \vdash_{Dst} e : s$ . The proof rules of Figure 6.4 slightly deviate from the  $\vdash_{Dst}$  rules, since the former are based on the type system of [84]. For instance, in the  $[LET_{DstC}]$  rule, the bound variable is not allowed to have an in-danger mark. The  $\triangleright$  operator ensures that no variables with unsafe types in  $\Gamma_1$  occur free in  $L_2$ . This is expressed as an additional condition in the  $\vdash_{Dst}$  rules. The condition  $\Gamma_D$  *well-formed* specifies that, if a variable  $x$  occurs in  $\Gamma_D$  with a condemned mark, all the variables in  $sharec(x, e)$  must also occur in  $\Gamma_D$  with an unsafe mark. The result of the  $\otimes$  operator occurring in  $[CASE_{DstC}]$  and  $[CASE!_{DstC}]$  is defined if the given environments map common variables to the same type, and its result is the union of the bindings of the environments to which it is applied. The  $[REC_{DstC}]$  rule plays a similar role as in the rules for region deallocation.

Once we have defined the proof rules, we have to specify how to prove them correct. This amounts to proving (6.7). Again, we do this by relating the static properties of the rules (mark environments) to the dynamic properties of the language (runtime closedness). Given a *Core-Safe* expression, its corresponding static assertion is a judgement of the form  $e : \llbracket L, \Gamma_D \rrbracket$ , where  $L$  contains the set of free variables in  $e$ . An expression  $e$  is said to satisfy this assertion if  $fv(e) \subseteq L$  and some others conditions shown below hold. The generic part of the certificate generates a proof different from that described in Section 3.6.4. The definition shown below, which is based on [84], relies on the disjointness of the  $S$  and  $R$  sets corresponding to a given configuration  $(L, \Gamma_D, E, h)$ . These sets are defined as follows:

$$\begin{aligned} S_{L, \Gamma_D, E, h} &\stackrel{\text{def}}{=} \bigcup_{x \in L, \Gamma_D(x)=s} \{closure(h, E(x))\} \\ R_{L, \Gamma_D, E, h} &\stackrel{\text{def}}{=} \bigcup_{x \in L, \Gamma_D(x)=d} \{p \in live(h, E, L) \mid p \rightarrow_h^* recReach(h, E(x))\} \end{aligned}$$

In other words, the disjointness of  $S$  and  $R$  implies that there are no safe variables pointing to recursive children of condemned DSs (otherwise they would be in-danger). Now we give a formal meaning of an expression satisfying a given assertion.

**Definition 6.8.** Given the following properties:

$$\begin{aligned} PD1 &\equiv E \vdash h, k, e \Downarrow h', k, v \\ PD2 &\equiv \text{dom } \Gamma_D \subseteq \text{dom } E \\ PD3 &\equiv L \subseteq \text{dom } \Gamma_D \\ PD4 &\equiv fv(e) \subseteq L \\ PD5 &\equiv \forall x \in \text{dom } E. \forall z \in L. \\ &\quad \Gamma_D(z) = d \wedge recReach(h, E(z)) \cap closure(h, E(x)) \neq \emptyset \rightarrow x \in \text{dom } \Gamma_D \wedge \Gamma_D(x) \neq s \\ PD6 &\equiv \forall x \in \text{dom } E. closure(h, E(x)) \neq closure(h', E(x)) \rightarrow x \in \text{dom } \Gamma_D \wedge \Gamma_D(x) \neq s \\ PD7 &\equiv S_{L, \Gamma_D, E, h} \cap R_{L, \Gamma_D, E, h} = \emptyset \\ PD8 &\equiv closed(h, E, L) \\ PD9 &\equiv closed(h', v) \end{aligned}$$

we say that the expression  $e$  satisfies the static assertion  $\llbracket L, \Gamma_D \rrbracket$ , denoted  $e : \llbracket L, \Gamma_D \rrbracket$ , iff

$$PD3 \wedge PD4 \wedge (\forall E, h, k, h', v. PD1 \wedge PD2 \rightarrow (PD5 \wedge PD6 \wedge (PD7 \wedge PD8 \rightarrow PD9)))$$

In a similar way as before, we also define a notion of satisfaction relative to the validity of a mark environment  $\Sigma_D$ , denoted  $e, \Sigma_D : \llbracket L, \Gamma_D \rrbracket$  (see [37] for details).

Property  $PD1$  defines the runtime evaluation of  $e$ . Properties  $PD2$  to  $PD4$  just guarantee that each free variable has a type and a value. Properties  $PD5$  to  $PD7$  formalise the dynamic meaning of safe and

condemned types: if a variable can share a recursive descendant of a condemned one, or its closure changes during evaluation, it has to occur in the environment with an unsafe type. The key properties are *PD8* and *PD9*. If we prove them to hold for every judgement in the derivation of *PD1*, we guarantee that the live part of the heap remains closed during the execution of  $e$ , and hence that there are not dangling pointers at runtime. The following theorem shows that *PD8* is an invariant which is propagated upwards through the evaluation tree, whereas *PD9* is propagated downwards.

**Theorem 6.9** (closedness). *Let us assume a  $\Downarrow$ -judgement:*

$$E \vdash h, k, e \Downarrow h', k, v \quad (6.9)$$

*For every  $L$  and  $\Gamma_D$  such that  $e : \llbracket L, \Gamma_D \rrbracket$ : if *PD2*( $\Gamma_D, E$ ), *PD7*( $L, \Gamma_D, E, h$ ) and *PD8*( $E, L, h$ ) hold, then for every judgement  $E_i \vdash h_i, k_i, e_i \Downarrow h'_i, k_i, v_i$  occurring in the derivation of (6.9) it holds that *PD8*( $E_i, L_i, h_i$ ) and *PD9*( $v_i, h'_i$ ).*

*Proof.* (see [37, 35]). □

Since *PD2*, *PD7* and *PD8* hold trivially for the empty heap, the empty mark environment, and the empty set of free variables, which are the ones corresponding to the initial expression, *PD8* and *PD9* are guaranteed to hold across the whole derivation of the program.

Similarly as in the previous section, the last step is to prove the soundness relation between the proof rules and these static assertions.

**Theorem 6.10** (soundness). *If  $e, \Sigma_D \vdash (L, \Gamma_D)$ , then  $e, \Sigma_D : \llbracket L, \Gamma_D \rrbracket$ .*

*Proof.* It follows from the proof described in [84] (see [35] for more details). □

### 6.3 Certificate generation

Once we have established the correctness of the proof rules of Figures 6.3 and 6.4, this section explains how to generate, an Isabelle/HOL script establishing that these proof rules can be applied to the program being certified. We assume that the generic part of the certificate (given by the theorems in the last section) has been proved in advance, and that its corresponding theorems are available in a database of proven facts. This part of the certificate has been proved interactively with Isabelle/HOL [35]. In this section we focus on the program-specific part. Assuming we are given a *Core-Safe* program with the type information inferred in Chapters 4 and 5, the certifier generates a plain text file containing an Isabelle/HOL script. This script contains, for every function definition of the program, the following information:

- An Isabelle/HOL definition of the abstract syntax tree (AST) of the function's body.
- A set of Isabelle/HOL definitions of the static objects inferred by the compiler: sets of free variables, mark environments, region typing environments, type instantiation mappings, etc.
- A set of Isabelle/HOL proof scripts proving a lemma for each expression. Each proof consists in first checking the premises of the proof rule associated to the syntactic form of the expression, and then applying this proof rule.

<b>LExp</b> $\ni$ $\varepsilon \rightarrow$	$a$	{atom: literal $c$ or variable $x$ }
	$x @ r$	{copy}
	$a_1 \oplus a_2$	{basic operator application}
	$C \bar{a}_i @ r$	{constructor application}
	$f \bar{a}_i @ \bar{r}_j$	{function application}
	$\mathbf{let} \ x_1 = m_1 \ \mathbf{in} \ m_2$	{let declaration: $m_1, m_2 \in \mathbb{N}$ }
	$\mathbf{case} \ x \ \mathbf{of} \ \bar{C}_i \ \bar{x}_{ij}^{n_i} \rightarrow m_i$	{read-only pattern matching: $m_i \in \mathbb{N}$ }
	$\mathbf{case!} \ x \ \mathbf{of} \ \bar{C}_i \ \bar{x}_{ij}^{n_i} \rightarrow m_i$	{destructive pattern matching: $m_i \in \mathbb{N}$ }

Figure 6.5: Syntax of labelled *Core-Safe* expressions.

With respect to the latter part, the certificate proves the lemmas of compound expressions in a bottom-up way. For example, when proving that the judgement  $\mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2, \Sigma_R \vdash \Gamma_R \rightsquigarrow s$  holds, the certificate must have proved the validity of the judgements corresponding to the sub-expressions  $e_1$  and  $e_2$ . For this reason, there must be a way to uniquely identify every sub-expression in a given function definition, so each sub-expression can be referred to when proving the lemma of the parent expression. Hence, our first goal is to split the body of the function being certified into a list of its sub-expressions. Each of these is identified by a natural number. When dealing with compound expressions (i.e. **let**, **case** and **case!**), we replace each sub-expression by the number identifying it. As a consequence, we have to slightly adapt the syntax of our *Core-Safe* expressions in order to support these identifiers. The set of *labelled Core-Safe expressions* is defined by the grammar of Figure 6.5. Notice the absence of nested expressions: these have been replaced by their identifier.

**Example 6.11.** Consider the following function definition:

$$\begin{aligned}
 \mathit{unshuffle} &:: [\alpha]! \rightarrow ([\alpha], [\alpha]) \\
 \mathit{unshuffle} \ []! &= ([], []) \\
 \mathit{unshuffle} \ (x : xs)! &= (x : \mathit{ys2}, \mathit{ys1}) \ \mathbf{where} \ (\mathit{ys1}, \mathit{ys2}) = \mathit{unshuffle} \ xs
 \end{aligned}$$

Its translation into *Core-Safe* yields the following result:

$$\begin{aligned}
 \mathit{unshuffle} &:: [\alpha]!@ \rho \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow ([\alpha]@ \rho_1, [\alpha]@ \rho_2)@ \rho_3 \\
 \mathit{unshuffle} \ xs @ r_1 \ r_2 \ r_3 &= \\
 &\mathbf{case!} \ xs \ \mathbf{of} \\
 &\quad [] \quad \rightarrow \mathbf{let} \ x_1 = [] @ r_1 \ \mathbf{in} \\
 &\quad \quad \mathbf{let} \ x_2 = [] @ r_2 \ \mathbf{in} \ (x_1, x_2)@ r_3 \\
 &\quad (x : xx) \rightarrow \mathbf{let} \ x_4 = \mathit{unshuffle} \ xx @ r_2 \ r_1 \ \mathbf{self} \ \mathbf{in} \\
 &\quad \quad \mathbf{let} \ x_5 = \mathbf{case} \ x_4 \ \mathbf{of} \ (a, b) \rightarrow a \ \mathbf{in} \\
 &\quad \quad \mathbf{let} \ x_6 = \mathbf{case} \ x_4 \ \mathbf{of} \ (c, d) \rightarrow d \ \mathbf{in} \\
 &\quad \quad \mathbf{let} \ x_7 = (x : x_6)@ r_1 \ \mathbf{in} \ (x_7, x_5)@ r_3
 \end{aligned}$$

If we split the body of this function into its sub-expressions, we get the list of labelled expressions shown in Figure 6.6. The elements of the list are arranged bottom-up (i.e. from simple to compound expressions), because this is the order required by Isabelle/HOL for applying the proof rules.  $\square$

Given the above, the first phase consists in translating the function being certified into a set of tuples containing every sub-expression in its labelled form, together with its related static information. Each

Identifier ( $m$ )	Expression ( $\varepsilon_m$ )
1	<i>unshuffle</i> $xx @ r_2 r_1 self$
2	$a$
3	<b>case</b> $x_4$ <b>of</b> $(a, b) \rightarrow 2$
4	$d$
5	<b>case</b> $x_4$ <b>of</b> $(c, d) \rightarrow 4$
6	$(x_7, x_5) @ r_3$
7	$(x : x_6) @ r_1$
8	<b>let</b> $x_7 = 7$ <b>in</b> 6
9	<b>let</b> $x_6 = 5$ <b>in</b> 8
10	<b>let</b> $x_5 = 3$ <b>in</b> 9
11	<b>let</b> $x_4 = 1$ <b>in</b> 10
12	$(x_1, x_2) @ r_3$
13	$[] @ r_2$
14	$[] @ r_1$
15	<b>let</b> $x_2 = 13$ <b>in</b> 12
16	<b>let</b> $x_1 = 14$ <b>in</b> 15
17	<b>case!</b> $xs$ <b>of</b> $\{[] \rightarrow 16; (x : xx) \rightarrow 11\}$

Figure 6.6: Labelled expressions resulting from splitting the body of *unshuffle*. The components of **let** and **case** expressions have been replaced by their identifiers.

tuple  $\phi_m$  has the following components:

$$\phi_m \equiv (m, \varepsilon_m, L_m, \Gamma_{D,m}, \Gamma_{R,m}, s_m, \mu_m) \quad (6.10)$$

where  $m$  is the tuple identifier,  $\varepsilon_m$  is the corresponding sub-expression,  $L_m$  is the set of its free variables,  $\Gamma_{D,m}$  and  $\Gamma_{R,m}$  denote the mark and region environments typing the sub-expression,  $s_m$  is its safe type, and  $\mu_m$  is the type instantiation mapping needed when  $\varepsilon_m$  is a function or a constructor application.

**Example 6.12.** Given the *unshuffle* function of Example 6.11 above, in Figure 6.7 we show the components of each tuple. Function and constructor applications also have the extra component  $\mu_m$  which is the instantiation mappings used in each case, and assuming the following generic type schemes for the list and tuple constructors.

$$\begin{aligned} (\cdot) &:: \alpha \rightarrow [\alpha]@ \rho \rightarrow \rho \rightarrow [\alpha]@ \rho \\ [] &:: \rho \rightarrow [\alpha]@ \rho \\ (\cdot, \cdot) &:: \alpha_1 \rightarrow \alpha_2 \rightarrow \rho \rightarrow (\alpha_1, \alpha_2)@ \rho \end{aligned}$$

□

There is also a tuple for every function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . Each one has the following components:

$$\Phi_f \equiv (f, \bar{x}_i^n, \bar{r}_j^m, \bar{m}_i^n, \Sigma_{D,f}, \Sigma_{R,f}, e_f, \bar{\phi}_i^q) \quad (6.11)$$

The first three components are self-explanatory, as well as  $e_f$ . The list  $\bar{m}_i^n$  denotes the marks inferred for the function parameters. The  $\Sigma_{D,f}$  environment contains the mark signatures of the functions that have been certified before  $f$ , together with the mark signature of the latter. Thus the signature of every function being called in  $e_f$  is assumed to belong to this environment. The  $\Sigma_{R,f}$  component has a similar

$m$	$L_m$	$\Gamma_{D,m}$	$\Gamma_{R,m}$	$S_m$
1	$\{xx\}$	$[xx : d, xs : r]$	$\Gamma_{R,11}$	$([\alpha]@_{\rho_2}, [\alpha]@_{\rho_1})@_{\rho_3}$
2	$\{a\}$	$[a : s]$	$\Gamma_{R,10} + [a : [\alpha]@_{\rho_2}, b : [\alpha]@_{\rho_1}]$	$[\alpha]@_{\rho_2}$
3	$\{x_4\}$	$[x_4 : s]$	$\Gamma_{R,10}$	$[\alpha]@_{\rho_2}$
4	$\{d\}$	$[d : s]$	$\Gamma_{R,9} + [c : [\alpha]@_{\rho_2}, d : [\alpha]@_{\rho_1}]$	$[\alpha]@_{\rho_1}$
5	$\{x_4\}$	$[x_4 : s]$	$\Gamma_{R,9}$	$[\alpha]@_{\rho_1}$
6	$\{x_7, x_5\}$	$[x_7 : s, x_5 : s]$	$\Gamma_{R,8} + [x_7 : [\alpha]@_{\rho_1}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
7	$\{x, x_6\}$	$[x : s, x_6 : s]$	$\Gamma_{R,8}$	$[\alpha]@_{\rho_1}$
8	$\{x, x_6, x_5\}$	$[x : s, x_6 : s, x_5 : s]$	$\Gamma_{R,9} + [x_6 : [\alpha]@_{\rho_1}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
9	$\{x_4, x, x_5\}$	$[x : s, x_4 : s, x_5 : s]$	$\Gamma_{R,10} + [x_5 : [\alpha]@_{\rho_2}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
10	$\{x_4, x\}$	$[x_4 : s, x : s]$	$\Gamma_{R,11} + [x_4 : ([\alpha]@_{\rho_2}, [\alpha]@_{\rho_1})@_{\rho_{self}}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
11	$\{xx, x\}$	$[xx : d, xs : r, x : s]$	$\Gamma_{R,17} + [x : \alpha, xx : [\alpha]@_{\rho}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
12	$\{x_1, x_2\}$	$[x_1 : s, x_2 : s]$	$\Gamma_{R,15} + [x_2 : [\alpha]@_{\rho_2}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
13	$\emptyset$	$[\ ]$	$\Gamma_{R,15}$	$[\alpha]@_{\rho_2}$
14	$\emptyset$	$[\ ]$	$\Gamma_{R,17}$	$[\alpha]@_{\rho_1}$
15	$\{x_1\}$	$[x_1 : s]$	$\Gamma_{R,17} + [x_1 : [\alpha]@_{\rho_1}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
16	$\emptyset$	$[\ ]$	$\Gamma_{R,17}$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$
17	$\{xs\}$	$[xs : d]$	$[xs : [\alpha]@_{\rho}, r_1 : \rho_1, r_2 : \rho_2, r_3 : \rho_3, self : \rho_{self}]$	$([\alpha]@_{\rho_1}, [\alpha]@_{\rho_2})@_{\rho_3}$

Figure 6.7: Free variables and typing information attached to each labelled expression.

$m$	$\mu_m$
1	$[\rho_1 \mapsto \rho_2, \rho_2 \mapsto \rho_1, \rho_3 \mapsto \rho_{self}]$
6	$[\alpha_1 \mapsto [\alpha]@_{\rho_1}, \alpha_2 \mapsto [\alpha]@_{\rho_2}, \rho \mapsto \rho_3]$
7	$[\alpha \mapsto \alpha, \rho \mapsto \rho_1]$
12	$[\alpha_1 \mapsto [\alpha]@_{\rho_1}, \alpha_2 \mapsto [\alpha]@_{\rho_2}, \rho \mapsto \rho_3]$
13	$[\alpha \mapsto \alpha, \rho \mapsto \rho_2]$
14	$[\alpha \mapsto \alpha, \rho \mapsto \rho_1]$

Figure 6.8: Type instantiation mappings of the function and constructor applications occurring in *unshuffle*.

meaning, but it contains safe type schemes, instead of mark signatures. Lastly, the  $\overline{\phi}_i^q$  component contain the list of tuples into which the body  $e_f$  is split.

There also exists a tuple for the whole program. It has the form  $(\Gamma_C, \overline{\Phi}_{f_i}, \overline{\phi}_j)$  and contains the type schemes of the data constructors being defined in the program, the list of tuples corresponding to each function definition, and the list of tuples into which the main expression of the program is split.

The task of generating all these tuples from the input *Core-Safe* program is straightforward, and it will not be described here. These tuples provide a sequential representation of the expressions of a program which is more amenable to the final step of the certification process: the generation of the Isabelle/HOL script.

Even when we have separated the generic proofs of the certificate from the program-specific ones, a considerable part of the program-specific part is still boilerplate code which can be reused in every certificate with little alteration. In order to ease the programming of the certifier we have used a *template system* to deal with this code. A template is a fragment of Isabelle/HOL code, which is parametric on some parts. These parameters are given as *placeholders*. Templates have the following syntax:

```
%.template [TemplateName]

These are the contents of the template with ${PlaceHolders}

%.end
```

A placeholder is denoted with the syntax  $\${P}$ , being P its name. Templates are stored in a plain text file which is read and processed by the compiler. The certifying phase of the compiler substitutes the contents of the tuples  $\phi_i$  shown above for the placeholders in the template. Usually the compiler has to apply a given template several times and join the results. We shall explain this with an example:

**Example 6.13.** The generation of a mark environment is given by the following templates:

```
%.template EnvBindingsSet

${Bindings}

%.end

%.template EnvBinding

"${Variable}" \<mapsto> ${Type}

%.end

%.template Mark

${Mark}"

%.end
```

Given the environment  $\Gamma_D = [x : d, y : s, z : r]$ , the compiler firstly applies the `EnvBinding` template to the binding  $x : d$  as follows: it instantiates the `Mark` template with the  $d$  mark, to get the string  $d$ , and then puts this string into the `{Type}` placeholder in `EnvBinding`. It also substitutes `{Variable}` for  $x$ , so as to get `"x" \<mapsto> d`. If we do the same with the bindings  $y : s$  and  $z : r$ , we obtain two additional strings:

`"y" \<mapsto> s`                      `"z" \<mapsto> r`

Finally, the three strings are joined with a `(,)` between them. The result is substituted for the `{Bindings}` placeholder in `EnvBindingsSet` and we obtain:

`{"x" \<mapsto> d, "y" \<mapsto> s, "z" \<mapsto> r}`

□

The overall structure of the certificate is given by the template shown in Figure 6.9. The `{ModuleName}` placeholder contains the name of the program being certified. The  $\Gamma_C$  environment containing the signatures of the constructors is stored in `{CSigs}`. The region and mark environments corresponding to every function definition (components  $\Sigma_{R,f}$  and  $\Sigma_{D,f}$  of the tuple (6.11)) are placed in `{SigmaDefs}`. These environments are generated by the following template:

```
%template SigmaEntry

constdefs \<Sigma>_{FuncName}_reg ::
  "string \<rightarrow>
    TypeExpression list \<times> VarType list \<times> TypeExpression"
  "\<Sigma>_{FuncName}_reg \<equiv> [ {SigmaRegEnv} ]"
```

```
constdefs \<Sigma>_{FuncName}_mark ::
  "string \<rightarrow> Mark list"
  "\<Sigma>_{FuncName}_mark \<equiv> [ {SigmaMarkEnv} ]"
```

```
%end
```

This template is filled in for every function definition appearing in the program. Region signature environments are defined as partial mappings (`\<rightarrow>`) from function names (`string`) to triples containing the type of the parameters (`TypeExpression list`), the types of the region parameters (`VarType list`) and the type of the result (`TypeExpression`). The environments themselves are generated as in Example 6.13 and placed in `{SigmaRegEnv}` and `{SigmaMarkEnv}` respectively.

**Example 6.14.** When applying these templates to our *unshuffle* function (Example 6.11) we get the following Isabelle/HOL definitions:

```
constdefs  $\Sigma_{\text{unshuffle\_reg}}$  :: "string  $\rightarrow$ 
  TypeExpression list  $\times$  VarType list  $\times$  TypeExpression"
 $\Sigma_{\text{unshuffle\_reg}} \equiv [$ 
```

```

%.template Main

(* Title:      Pointer safety certificate for ${Filename}
   Generated:  ${Timestamp}
   Copyright:  2010 Universidad Complutense de Madrid
*)
header {* Safety certificate for ${Filename} *}
theory ${ModuleName} imports THRegionsDatabase
begin

(***** CONSTRUCTOR TYPE SIGNATURES *****)

defs constructorSignature_def : "constructorSignature \<equiv> ${CSigs}"

(***** SIGMA ENVIRONMENT DEFINITIONS *****)

${SigmaDefs}

(***** REGION ENVIRONMENT DEFINITIONS *****)

${RegEnvDefs}

(***** L SET DEFINITIONS *****)

${LDefs}

(***** MARK ENVIRONMENT DEFINITIONS *****)

${MarkEnvDefs}

(***** TYPE DEFINITIONS *****)

${TDefs}

(***** MU INSTANTIATION MAPPING DEFINITIONS *****)

${MuDefs}

(***** PROGRAM EXPRESSIONS *****)

${ExpDefs}

(***** REGION SAFETY PROOFS *****)

${RegProofs}

(***** EXPLICIT DESTRUCTION SAFETY PROOFS *****)

${DstProofs}

end

%.end

```

Figure 6.9: Global structure of the program-specific part of the certificate

```

''_MOD_'' ↦ ([ConstrT intType [] [], ConstrT intType [] []], [],
             ConstrT intType [] []),
''_PLUS_'' ↦ ([ConstrT intType [] [], ConstrT intType [] []], [],
              ConstrT intType [] []),
...
''unshuffle'' ↦ ([ConstrT ''ListT'' [VarT ''a'' ] [''rho'']],
                 [''rho2'', ''rho3'', ''rho4''],
                 ConstrT ''TupleTT_2'' [ConstrT ''ListT'' [VarT ''a'' ] [''rho2'']],
                                     ConstrT ''ListT'' [VarT ''a'' ] [''rho3'']]
                 [''rho4''])
]''

constdefs Σ_unshuffle_mark :: "string → Mark list"
"Σ_unshuffle_mark ≡ [
  ''_MOD_'' ↦ [s'', s''],
  ''_PLUS_'' ↦ [s'', s''],
  ...
  ''unshuffle'' ↦ [d'']
]''

```

These initial environments contain the types of built-in operators and some basic functions. It also contains a representation of the type of *unshuffle*. □

The remaining definitions in the certificate are the components of the tuple (6.10). To avoid name clashes in the certificate, the name of each Isabelle/HOL definition contains the name of the function to which it belongs, and the identifier of the tuple. As an example, we show the template for generating the  $L$  sets of free variables of each sub-expression.

```

%.template LDef

constdefs L_ ${FuncName} ${ExpNum} :: "string set"
           "L_ ${FuncName} ${ExpNum} \<equiv> { ${Vars} }"

%.end

```

An instance of this template is generated for each expression in the program. The  $\text{\${FuncName}}$  placeholder contains the name of the context function and  $\text{\${ExpNum}}$  contains the identifier of the corresponding tuple.

**Example 6.15.** Back to our *unshuffle* function, the certifier yields<sup>1</sup> the following definitions for the labelled expression  $\varepsilon_8$ :

```

constdefs Γ_D_unshuffle_8 :: "TypeEnvironment"
"Γ_D_unshuffle_8_reg ≡ [ ''x'' ↦ s'', ''x6'' ↦ s'', ''x5'' ↦ s'']"

constdefs L_unshuffle_8 :: "string set"
"L_unshuffle_8 ≡ { ''x'', ''x6'', ''x5'' }"

constdefs Γ_R_var_unshuffle_8 :: "TypeMapping"
"Γ_R_var_unshuffle_8 ≡ [

```

<sup>1</sup>Variables have been renamed to match those of Figure 6.7, for the sake of clarity.

```

''x6'' ↦ ConstrT ''ListT'' [VarT ''a'' ] [''rho1''],
''x5'' ↦ ConstrT ''ListT'' [VarT ''a'' ] [''rho2''],
''x4'' ↦ ConstrT ''TupleTT_2'' [
    ConstrT ''ListT'' [VarT ''a'' ] [''rho2''],
    ConstrT ''ListT'' [VarT ''a'' ] [''rho1'']] [''rho_self''],
''xx'' ↦ ConstrT ''ListT'' [VarT ''a'' ] [''rho''],
''x'' ↦ VarT ''a'',
''xs'' ↦ ConstrT ''ListT'' [VarT ''a'' ] [''rho'']] "

constdefs  $\Gamma_{R\_reg\_unshuffle\_8}$  :: "RegMapping"
 $\Gamma_{R\_reg\_unshuffle\_8}$   $\equiv$  [''r1'' ↦ ''rho1'' , ''r2'' ↦ ''rho2'' , ''r3'' ↦ ''rho3'' ,
    ''self'' ↦ ''rho_self'' ]"

constdefs Type_unshuffle_8 :: "TypeExpression"
Type_unshuffle_8  $\equiv$  ConstrT ''TupleTT_2''
    [ConstrT ''ListT'' [VarT ''a'' ] [''rho1''],
    ConstrT ''ListT'' [VarT ''a'' ] [''rho2'']] [''rho3'']"

```

The certificate splits the environment  $\Gamma_R$  into two: one containing the types of variables ( $\Gamma_{R\_var}$ ) and another one containing the types of region variables ( $\Gamma_{R\_reg}$ ).  $\square$

The last two placeholders of the certificate ( $\{\text{RegProofs}\}$  and  $\{\text{DstProofs}\}$ ) contain, for every expression  $e$ , the proof of the respective judgements  $e, \Sigma_D \vdash (L, \Gamma_D)$  and  $e, \Sigma_R \vdash \Gamma_R \rightsquigarrow s$ , where  $L$  is the set of free variables in  $e$ ,  $s$  is its type, and  $\Gamma_D$  and  $\Gamma_R$  are their corresponding environments. Each proof starts with the statement of the lemma to be proved.

```

%template DstProof

lemma e_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  _DST:
    "def_e_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  : \<lbrace> L_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  ,
        \<Gamma>_D_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  \<rbrace>"

 $\{\text{ProofBody}\}$ 

%end

%template RegProof

lemma e_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  _REG:
    "def_e_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  \<turnstile>
    (\<Gamma>_R_var_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  ,
        \<Gamma>_R_reg_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  )
        \<leadsto> Type_  $\{\text{FuncName}\}$  _  $\{\text{ExpNum}\}$  "

 $\{\text{ProofBody}\}$ 

%end

```

The  $\{\text{ProofBody}\}$  placeholder contains the script proving the corresponding lemma. This script depends on the form of the expression to which the lemma refers. Regarding judgements of the form

```

%.template ProofLet1S

apply (unfold def_e_{{FuncName}}_{{ExpNum}}_def)
apply (rule_tac ?L1.0="L_{{FuncName}}_{{ExpNum1}}" and
      ?\<Gamma>1.0="\<Gamma>_{{FuncName}}_{{ExpNum1}}" and
      ?L2.0="L_{{FuncName}}_{{ExpNum2}}" and
      ?\<Gamma>2'.0="\<Gamma>_{{FuncName}}_{{ExpNum2}}"
      in SafeDA_LET1_s)
apply (rule e_{{FuncName}}_{{ExpNum1}})
apply (rule e_{{FuncName}}_{{ExpNum2}})
apply (unfold \<Gamma>_{{FuncName}}_{{ExpNum2}}_def)
apply (simp add: disjointUnionEnv_def)?
apply (simp add: unionEnv_def)?
apply (rule ext, simp, force)?
apply (simp add: def_disjointUnionEnv_def)
apply (unfold \<Gamma>_{{FuncName}}_{{ExpNum1}}_def,
      unfold L_{{FuncName}}_{{ExpNum2}}_def,
      simp add: def_pp_def)
apply (unfold unsafe_def, force)?
apply (unfold L_{{FuncName}}_{{ExpNum1}}_def, force)
apply (unfold L_{{FuncName}}_{{ExpNum}}_def, force)
apply (unfold \<Gamma>_{{FuncName}}_{{ExpNum}}_def)
apply (simp add: pp_def)
apply (simp add: disjointUnionEnv_def, simp add: unionEnv_def)?
apply (rule ext, simp)
by (unfold def_e_{{FuncName}}_{{ExpNum1}}_def, simp)

%.end

```

Figure 6.10: Proof script template for **let** expressions (explicit deallocation).

$e, \Sigma_D : \llbracket L, \Gamma_D \rrbracket$ , the proof script is generated as a sequence of Isabelle/HOL tactics and rules. For instance, Figure 6.10 shows the script corresponding **let** expressions in which the bound variable gets an *s* mark in the main expression. The second line in this script consists in applying the  $[LET_{DstC}]$  rule. This rule generates several proof obligations: one for each of its premises. The subsequent lines of the script are aimed at proving these proof obligations.

With respect to the judgements of the form  $e, \Sigma_R \vdash \Gamma_R \rightsquigarrow s$ , de Dios has defined in [35] an Isabelle/HOL tactic for each of the rules in Figure 6.3. So, the proof of each of the lemmas generated by the RegProof template shown above consists in applying the corresponding tactic. Each tactic applies a type rule and discharges its associated proof obligations in a single proof step. This reduces both the size and checking time of the certificate. For instance, Figure 6.11 shows the template which applies the  $[LET_{RegC}]$  rule and discharges all its associated proof obligations. It is part of ongoing work to define tactics for the proofs related to explicit destruction.

**Example 6.16.** By applying these templates to the expression  $\varepsilon_8$  above, we get the following proof:

```

lemma e_unshuffle_8_REG:
  "def_e_unshuffle_8  $\vdash$ 

```

```

%.template RegProofLet

by (tactic "SafeDARegion_LET1_tac
  (@{thm def_e_ ${FuncName} _ ${ExpNum} _def},
   @{thm \<Gamma>_R_var_ ${FuncName} _ ${ExpNum} _def},
   @{thm \<Gamma>_R_reg_ ${FuncName} _ ${ExpNum} _def},
   @{thm Type_ ${FuncName} _ ${ExpNum} _def},
   @{thm def_e_ ${FuncName} _ ${ExpNum1} _def},
   @{thm \<Gamma>_R_var_ ${FuncName} _ ${ExpNum1} _def},
   @{thm \<Gamma>_R_reg_ ${FuncName} _ ${ExpNum1} _def},
   @{thm Type_ ${FuncName} _ ${ExpNum1} _def},
   @{thm e_ ${FuncName} _ ${ExpNum1} _REG},
   @{thm def_e_ ${FuncName} _ ${ExpNum2} _def},
   @{thm \<Gamma>_R_var_ ${FuncName} _ ${ExpNum2} _def},
   @{thm \<Gamma>_R_reg_ ${FuncName} _ ${ExpNum2} _def},
   @{thm Type_ ${FuncName} _ ${ExpNum2} _def},
   @{thm e_ ${FuncName} _ ${ExpNum2} _REG},
   "\<Gamma>_R_var_ ${FuncName} _ ${ExpNum1} _",
   "\<Gamma>_R_reg_ ${FuncName} _ ${ExpNum1} _",
   "Type_ ${FuncName} _ ${ExpNum1} _",
   "\<Gamma>_R_var_ ${FuncName} _ ${ExpNum2} _",
   "\<Gamma>_R_reg_ ${FuncName} _ ${ExpNum2} _",
   "Type_ ${FuncName} _ ${ExpNum2} _")")

%.end

```

Figure 6.11: Proof script template for **let** expressions (region deallocation). There is a single step consisting in applying an user-defined tactic.

Name	Core-Safe code	Size of the certificate (lines)				Ratio
	Size (lines)	Definitions	Proofs (regs.)	Proofs (dest.)	Total	
basic	306	2667	1634	2218	6619	21.63
unshuffle	49	429	271	407	1107	22.59
mergesort	189	1748	1134	1753	4635	24.52
inssort	67	651	406	616	1673	24.97
pascal	106	1105	658	911	2674	25.22
compiler	2058	16131	10262	13471	39864	19.37
AVLTrees	502	4680	3008	4484	12472	24.84

Figure 6.12: Comparison of sizes between the certificates and the source *Core-Safe* code.

```

(Γ_R_var_unshuffle_8, Γ_R_reg_unshuffle_8) ~> Type_unshuffle_8"
by (tactic "SafeDARegion_LET1_tac
  (@{thm def_e_unshuffle_8_def},
   @{thm Γ_R_var_unshuffle_8_def},
   @{thm Γ_R_reg_unshuffle_8_def},
   @{thm Type_unshuffle_8_def},
   @{thm def_e_unshuffle_7_def},
   @{thm Γ_R_var_unshuffle_7_def},
   @{thm Γ_R_reg_unshuffle_7_def},
   @{thm Type_unshuffle_7_def},
   @{thm e_unshuffle_7_REG},
   @{thm def_e_unshuffle_6_def},
   @{thm Γ_R_var_unshuffle_6_def},
   @{thm Γ_R_reg_unshuffle_6_def},
   @{thm Type_unshuffle_6_def},
   @{thm e_unshuffle_6_REG},
   "Γ_R_var_unshuffle_7",
   "Γ_R_reg_unshuffle_7",
   "Type_unshuffle_7",
   "Γ_R_var_unshuffle_7",
   "Γ_R_reg_unshuffle_7",
   "Type_unshuffle_7"
  )

```

□

## 6.4 Case studies

In this section we aim to measure the sizes of the resulting certificates in comparison with the size of their respective source code. The table in Figure 6.12 shows the results. For each certificate, we have measured the number of lines generated for the definitions involved in the proofs (such as typing environments, mark environments, instantiation mappings, sets of free variables, etc.) and the number of lines of the proofs themselves. The column *Proofs (regs.)* refers to the part of the certificate related to region deallocation, whereas *Proofs (dest.)* measures the part regarding explicit destruction. The basic file contains several example functions on lists and binary search trees (*append*, *appendC*, *insertT*, etc.), most of which were defined in Chapter 2. The pascal example was shown in Example 4.20, whereas compiler corresponds to the small compiler developed at the end of Chapter 4 (Example 4.22). Lastly, AVLTrees contains the code of the functions of a small library for handling AVL Trees, which was shown in Example 3.14.

From these results we can draw the following conclusions:

1. The size of the certificate grows linearly with the size of the input *Core-Safe* program. The expansion factor (last column in the table) ranges between 19 and 25.
2. A considerable part of the certificate (around 40%) is devoted to the definitions of the elements in the tuples shown in (6.10) and (6.11). The proofs of region consistency preservation make up approximately 25% of the certificate, whereas the remaining 35% deals with the proofs related to explicit deallocation. We expect the latter to decrease when we replace the actual proof scripts by the application of used-defined tactics, as it was done with the part related to region consistency.

## 6.5 Conclusions and related work

In this chapter we have shown how to encode, in Isabelle/HOL, the proof that a *Core-Safe* program is well-typed, so that it can be mechanically checked for that program. The hardest part of the proof (correctness of the type system) is assumed to have been done in advance, and once for all. This results in smallest certificates, and shorter checking times. In spite of that, certificates are approximately twenty times bigger than the program being certified. Another approach would have been to implement a certified type checker, and that the code consumer applies this checker to the input *Core-Safe* program. However, this delegates the type-checking task to the code consumer. It is necessary to find a balance between the size of the certificates, and the amount of work to be done by the code consumer. With regard to this balance, it is worth taking into account that the certification of a program is associated with its deployment in the target machine, but neither with the development of that program, nor with its execution on the target machine. Therefore, deployment is not a so frequent task in software development, as to consider the certificate checking time to be decisive.

Introducing pointers in a Hoare-style assertion logic and using a proof assistant for proving the correctness of programs with pointers goes back to the late seventies [74], where the Stanford Pascal Program Verifier was used. A more recent reference is [20], using the Jape proof editor. A formalisation of Bornat's ideas in Isabelle/HOL was done by Mehta and Nipkow in [78], where they add a complete soundness proof. None of these papers address the problem of automatically generating a certificate.

Connecting the results of a static analysis with the generation of certificates was done from the beginning of the PCC paradigm (see for instance [90]). A more recent work is [15].

Our work is more closely related to [18], where a resource consumption property obtained by Hofmann and Jost's type system [58] is transformed into a certificate. The compiler is able to infer a linear upper bound on heap consumption and to certify this property by emitting an Isabelle/HOL script proving it. Our static assertions have been inspired by their *derived assertions*, used also there to connect static with dynamic properties. However, their heap is simpler to deal with than ours since it essentially consists of a free list of cells, and the only data type available is the list. We must also deal with regions and with any user-defined data type. This results in our complex notion of consistency. The authors of [18] conjecture a set of proof rules and claim they could be used to prove the safety of destruction in Hofmann and Aspinall's type system [11]. But in fact they do not provide an Isabelle/HOL or a manual proof of these rules. The lessons that can be drawn from [37, 35] about proving the correctness of proof rules of Section 6.2 in a proof assistant is that it is a rather daunting task full of unexpected difficulties which have forced us to frequently modify the proof rules.

## Chapter 7

# Memory consumption analysis

### 7.1 Introduction

In this chapter we present a static analysis aimed at inferring upper bounds to the heap and stack consumption made by a *Safe* program. Since the memory needs of a program usually depends on its input, the bounds we obtain in our analysis are functions on the sizes of the inputs. The bounds given by this memory consumption analysis are considered correct if they are equal or greater than the actual worst-case consumptions of the program being analysed.

The inference of the space complexity of a program is a very complex task, even for a first-order language like *Safe*. There are three separate aspects that can be independently studied and solved:

1. Inference of upper bounds to the size of the call-tree deployed at runtime by each recursive *Safe* function.
2. Inference of the sizes of the data structures involved in a *Safe* function.
3. From the results of (1) and (2), inference of upper bounds to the stack and heap space costs of a *Safe* function.

In this chapter we deal exclusively with the latter point. The first two points are subject of ongoing work, and not part of this thesis. The first one is closely related to the termination of a *Safe* program. Although termination is an undecidable program property (as well as the properties given in (2) and (3)), several approaches have been proposed in the literature for approximating it. In particular, we can infer multivariate polynomials of any degree as solutions by using the approach described in [73]. Another possibility is the translation of a *Safe* function into a recurrence relation modelling the size of the call-tree, and use already existing tools (PURRS [13], PUBS [4]) for inferring closed-form expressions to the result of these recurrence relations. This allows us to infer expressions beyond polynomial bounds (in particular, logarithmic and exponential). With regard to (2), in [105] the authors describe how to obtain polynomial, non-monotonic size relations between the input and the output of a function. This approach can also be combined with the PUBS recurrence solver in order to tackle (1), as described in [87]. Another possibility is the approach described in [97], which infers linear size relations between the input and the output of a *Safe* function.

For the purposes of this chapter, we shall assume that the results of (1) and (2) are given by the programmer, so we can concentrate on (3) instead. In principle, the analysis described here can accommodate several complexity classes, as long as their functions are monotonic w.r.t. the input sizes.

Obviously, the complexity class of the result critically depends of the complexity class of the inputs given to the analysis (points (1) and (2) above). In principle, we infer bounds to the costs of *Core-Safe* function definitions without taking destructive pattern matching into account (*case!* expressions are handled as if they were non-destructive), and assuming absence of region-polymorphic recursion. At the end of this chapter we draft some ideas on inferring bounds for full-fledged *Core-Safe* programs.

This chapter is an extended and improved version of [83]. A major difference w.r.t. the latter is that, in this chapter, we present a method for flattening an expression into sequences of basic expressions. This makes the algorithm simpler when considering the base and recursive cases of a function definition. Flattening also makes the algorithm more precise.

The techniques described in this chapter are based on abstract interpretation [33]. Given a function definition of  $n$  parameters, the abstract domain is the set of  $n$ -ary monotonic functions from real non-negative numbers to a real non-negative result<sup>1</sup>. This domain turns out to be a complete lattice under the usual  $\sqsubseteq$  relation on functions. This domain is detailed in Section 7.2. In Section 7.3 we define an interpretation on expressions which is monotonic w.r.t. this domain, and in Section 7.4 we prove this interpretation correct. The interpretation itself allows us to devise a space analysis aimed at non-recursive function definitions. Recursive function definitions are dealt with in Section 7.5. After this, we show in Section 7.6 how the algorithm behaves with several examples and case studies. Finally, in Section 7.7 we sketch some ideas on inferring bounds in presence of destructive pattern matching and polymorphic recursion on regions, and Section 7.8 concludes.

## 7.2 Function signatures

Let us recall the resource-aware semantics of Section 2.7. There we annotated the evaluation of an expression  $e$  with a resource vector  $(\delta, m, s)$ , where  $\delta$  gives the difference between the number of cells in each region after and before the evaluation of  $e$ , and  $m$  and  $s$  represent the number of heap cells (resp. stack words) needed for evaluating  $e$ . A space consumption analysis is expected to compute, given an expression  $e$ , upper bounds to each component  $(\delta, m, s)$  of the resource vector resulting from its evaluation. However, this resource vector does not solely depend on the expression  $e$  itself, but also on the initial value environment  $E$  and heap  $h$ , which contain the *input* to our expression. As a consequence, there are two different ways in which a memory consumption analysis can be designed.

1. The memory consumption analysis is given the input as a parameter, in addition to the expression  $e$  to be analysed. In this way, the analysis returns a resource vector  $(\delta', m', s')$  bounding the actual  $(\delta, m, s)$  components.
2. We do not assume a fixed input, but the analysis returns a resource vector  $(\delta', m', s')$  as a *function* of the given input.

The second option seems more reasonable from the programmer's point-of-view, since the input of a program is usually only known at runtime, whereas a memory consumption analysis is done at compile time. Nevertheless, the functions returned by the analysis would be complex and hard to interpret, unless we devise a simpler representation of the input, rather than a pair  $(E, h)$ . It is more useful, in practice, to abstract the inputs by their *sizes*. For example, we usually say that the mergesort algorithm has  $\mathcal{O}(n \log n)$  time complexity, being  $n$  the *number* of elements of the input list, without regard to the elements themselves. We will apply this abstraction in our analysis.

---

<sup>1</sup>By allowing real-valued functions instead of integer-valued, we can accommodate size models different from the one explained in this chapter.

Our first step is to define a suitable notion of *size*, so that the memory costs of a *Safe* function can be described as a function of this size. One may be tempted to define the size of a DS as the number of cells needed for storing it. However, this definition excludes the possibility of obtaining useful results for functions that receive an integer as an argument. An integer would always have zero size (since integers do not take heap cells by themselves), which would result in, for instance, a constant function giving the stack consumption of the *fib* function, whereas a linear function is expected. The notion of the size of the input should give an idea on how easy or difficult is to process that input.

**Definition 7.1.** Given a heap  $h$  and a value  $v$ , the *size* function is defined as follows:

- If  $v$  is a pointer  $p$ , *size* returns the number of cells taken by the recursive spine of the DSs which  $v$  points to. More formally:

$$\text{size}(h[p \mapsto (j, C \bar{v}_i^n)], p) = 1 + \sum_{i \in \text{RecPos}(C)} \text{size}(h, v_i)$$

- If  $v$  is a literal  $c \in \mathbf{Int}$ , then  $\text{size}(h, c) = c$ .
- If  $v$  is a literal  $c \in \mathbf{Bool}$ , then  $\text{size}(h, c) = 0$ .

It is worth noting that, with this definition, the size of a list with  $n$  elements is  $n + 1$ , since it is made of  $n$  cells with the  $(:)$  constructor plus an additional cell with the  $[]$  constructor. Analogously, the size of a binary search tree (as defined in Section 2.2) with  $n$  elements is always  $2n + 1$  ( $n$  cells with the *Node* constructor and  $n + 1$  cells with the *Empty* constructor).

Our space consumption analysis bounds the  $m$  and  $s$  components by means of numeric functions that depend on the size of the input. The  $\delta$  component deserves special attention, since it is not a single number, but a mapping from heap region identifiers (natural numbers) to integers. We cannot know, at compile time, which regions are used to build DSs during the execution of an expression, since this information is only known at runtime by means of the value environment  $E$ , which associates region variables with actual region identifiers. As a consequence, we have to abstract the region identifiers by the type  $\rho$  of their corresponding region variable. To compute a bound to the  $\delta$  component, our algorithm will compute a bound to the costs charged to each region type.

Recall that a *Core-Safe* function receives  $n + m$  variables as parameters:

$$\begin{aligned} f &:: t_1 \rightarrow \cdots \rightarrow t_n \rightarrow \rho_1 \rightarrow \cdots \rightarrow \rho_m \rightarrow t \\ f &x_1 \cdots x_n @ r_1 \cdots r_m = e_f \end{aligned} \tag{7.1}$$

Upper bounds to cost and sizes are represented by numbers in  $\mathbb{R}_\infty^+ \stackrel{\text{def}}{=} \mathbb{R}^+ \cup \{+\infty\}$ . The special value  $+\infty$  denotes an absence of upper bounds (either because there are no such bounds, or because the algorithm is not able to infer them). So, memory heap and stack needs are represented as functions of the following set:

$$\mathbb{F} = \{\zeta : ((\mathbb{R}_\infty^+)^{\perp})^n \rightarrow (\mathbb{R}_\infty^+)^{\perp} \mid \zeta \text{ is monotonic and strict}\}$$

The notation  $D^{\perp}$  denotes the set  $D \cup \{\perp\}$ . In this chapter we use the special value  $\perp$  to make undefinedness of functions explicit. If  $\zeta \in \mathbb{F}$ , we say that  $\zeta$  is undefined for some sizes  $\bar{x}_i^n$  if and only if  $\zeta \bar{x}_i^n = \perp$ . Hence, the *domain* of a function  $\zeta \in \mathbb{F}$  (denoted by  $\text{dom } \zeta$ ) is defined as follows:

$$\text{dom } \zeta = \{\bar{x}_i \in (\mathbb{R}_\infty^+)^n \mid \zeta \bar{x}_i \neq \perp\}$$

The intuitive meaning of a cost function returning  $\perp$  for a given size is that the function whose cost is being inferred does not evaluate to any value for that size. The strictness condition of space cost functions demands that the result is undefined if at least one of the arguments is undefined.

**Example 7.2.** Assume the following *Full-Safe* definition:

$$\text{dropTail } xs = \text{case } xs \text{ of } (x : xx) \rightarrow [x]$$

The function application  $\text{dropTail } []$  does not evaluate to any value. Hence the memory needs of this function can be specified as follows:

$$\zeta = \lambda xs. \begin{cases} 2 & \text{if } xs \geq 2 \\ \perp & \text{otherwise} \end{cases}$$

□

For the sake of clarity, we shall use curried notation when dealing with functions in  $\mathbb{F}$  (in order to avoid an excessive amount of parentheses) and  $\lambda$ -notation when defining these functions (for instance, as in  $\lambda x.x + 1$ ). We use  $\zeta, \zeta_i, \dots$  to denote a generic element of  $\mathbb{F}$ . If we use an element of  $\mathbb{F}$  for bounding *heap* memory needs (that is, the  $m$  component in the resource vector  $(\delta, m, s)$ ) we shall use  $\mu, \mu_1, \dots$  as metavariables, whereas in the context of *stack* memory needs (the  $s$  component) we use  $\sigma, \sigma_i, \dots$ .

The following guard notation will be convenient in the following sections. Given a boolean function  $G$  of  $n$  variables and a function  $\zeta \in \mathbb{F}$ , the notation  $[G \rightarrow \zeta]$  denotes the following function:

$$[G \rightarrow \zeta] = \lambda \bar{x}_i. \begin{cases} \zeta \bar{x}_i & \text{if } G \bar{x}_i \text{ holds} \\ \perp & \text{otherwise} \end{cases}$$

By abuse of notation we omit the  $\lambda \bar{x}_i.$  prefix in  $G$ . For instance, the function  $\zeta$  in Example 7.2 above can be expressed more succinctly as  $[xs \geq 2 \rightarrow 2]$ .

The usual ordering in  $\mathbb{R}^+$  can be extended to  $(\mathbb{R}_\infty^+)^{\perp}$  in the usual way:

$$x \leq y \Leftrightarrow_{\text{def}} x = \perp \vee y = +\infty \vee (x, y \in \mathbb{R}^+ \wedge x \leq y)$$

This order can be further extended to functions in  $\mathbb{F}$  as follows:

$$\zeta_1 \sqsubseteq \zeta_2 \Leftrightarrow_{\text{def}} \forall \bar{x}_i \in (\mathbb{R}_\infty^+)^n. \zeta_1 \bar{x}_i \sqsubseteq \zeta_2 \bar{x}_i$$

The arithmetic operations  $+$  and  $*$  can also be extended to  $(\mathbb{R}_\infty^+)^{\perp}$  as usual:

$$\left. \begin{aligned} (+\infty) + x &= x + (+\infty) = +\infty \\ (+\infty) * x &= x * (+\infty) = +\infty \end{aligned} \right\} \text{for every } x \in \mathbb{R}_\infty^+$$

$$\left. \begin{aligned} \perp + x &= x + \perp = \perp \\ \perp * x &= x * \perp = \perp \end{aligned} \right\} \text{for every } x \in (\mathbb{R}_\infty^+)^{\perp}$$

We have two different kinds of operators for denoting least upper bounds in  $(\mathbb{R}_\infty^+)^{\perp}$ ,  $\sqcup$  and  $\sqcup$ . The first one ignores undefined values, whereas the second one returns  $\perp$  if at least one of the elements to which it is applied is  $\perp$ . For example,  $\sqcup\{2, 5, \perp, 1\} = 5$ , but  $\sqcup\{2, 5, \perp, 1\} = \perp$ .

The  $+$ ,  $*$ ,  $\sqcup$  and  $\sqcap$  operators can be trivially extended to  $(\mathbb{R}_\infty^+)^{\perp}$ -valued functions in the usual way. It is easy to see these operators are monotonic on its arguments, and hence the  $\mathbb{F}$  set is closed under these operators. Moreover, we obtain the following result:

**Proposition 7.3.** *Given a fixed function definition  $f \bar{x}_i @ \bar{r}_j = e_f$ , its associated ordered set of space cost functions  $(\mathbb{F}, \sqsubseteq)$  is a complete lattice, whose bottommost element is  $\lambda \bar{x}_i. \perp$  and the topmost one is  $\lambda \bar{x}_i. + \infty$ .*

*Proof.* It follows directly from the fact that  $((\mathbb{R}_\infty^+)^{\perp}, \leq)$  is a complete lattice (see [92, Section A.2]).  $\square$

With regard to the  $\delta$  component of the resource vector we need to take the different region type variables into account. Assuming the function definition given in (7.1), its body can only charge space costs to the regions given as parameters (of types  $\rho_1 \dots \rho_m$ ) and to the working region *self* (of type  $\rho_{self}$ ). If we denote by  $R_f$  the set  $\{\rho_1, \dots, \rho_m\}$ , the function that gives the memory charges separated by region belong to the following class of functions:

$$\mathbb{D}^* = \{\Delta : ((\mathbb{R}_\infty^+)^{\perp})^n \rightarrow (R_f \cup \{\rho_{self}\}) \rightarrow \mathbb{R}_\infty^+ \mid \Delta \text{ is monotonic and strict}\}$$

and we use the variables  $\Delta, \Delta_1, \dots$  to denote functions in this set. The fact that  $\Delta$  is monotonic should be understood with regard to the following order relation:

$$\begin{aligned} \Delta \bar{x}_i \leq \Delta \bar{y}_i &\Leftrightarrow_{\text{def}} \forall \rho \in R_f \cup \{\rho_{self}\}. \Delta \bar{x}_i \rho \leq \Delta \bar{y}_i \rho \\ &\perp \leq \Delta \bar{y}_i \quad \text{for every } \bar{y}_i \end{aligned}$$

which works in a componentwise basis.

When considering a function definition from outside, its charges to the *self* region are not visible, since this region is disposed of when the function finishes. In these cases it is more convenient to remove the  $\rho_{self}$  region from  $\Delta$ , so as to get the following set of functions:

$$\mathbb{D} = \{\Delta : (\mathbb{R}_\infty^+)^n \rightarrow (R_f \rightarrow \mathbb{R}_\infty^+)^{\perp} \mid \Delta \text{ is monotonic}\}$$

The elements of  $\mathbb{D}$  and  $\mathbb{D}^*$  are called *abstract heaps*. Given an abstract heap  $\Delta \in \mathbb{D}^*$ , we denote by  $[\Delta]$  the function  $\lambda \bar{x}_i. (\Delta \bar{x}_i) |_{R_f}$ , which always belongs to  $\mathbb{D}$ . In other words,  $[\Delta]$  discards the information regarding the *self* region.

We define the addition  $+$ , and the least upper bound  $\sqcup$  between abstract heaps in a component-by-component basis, as usual. The same applies to the  $\sqsubseteq$  relation. Slightly different is the multiplication operator  $*$ , which is defined in  $\mathbb{F} \times \mathbb{D}$ , and not between two abstract heaps, as expected. Given  $\zeta \in \mathbb{F}$  and  $\Delta \in \mathbb{D}$ , the result of  $\zeta * \Delta$  is defined as follows:

$$\zeta * \Delta = \lambda \bar{x}_i. \begin{cases} \lambda \rho \in R_f. (\zeta \bar{x}_i) * (\Delta \bar{x}_i \rho) & \text{if } \Delta \bar{x}_i \neq \perp \\ \perp & \text{otherwise} \end{cases}$$

The guarded notation described previously can also be used with the elements of  $\mathbb{D}$  to obtain functions such as  $[G \rightarrow \Delta]$  for some  $\Delta \in \mathbb{D}$ . Finally, we introduce the  $|\cdot|$  operator as a function in  $\mathbb{D} \rightarrow \mathbb{F}$ , defined as follows:

$$|\Delta| = \lambda \bar{x}_i. \begin{cases} \sum_{\rho \in R_f} \Delta \bar{x}_i \rho & \text{if } \Delta \bar{x}_i \neq \perp \\ \perp & \text{otherwise} \end{cases}$$

Intuitively, the  $|\cdot|$  operator adds the charges made to all RTVs into a single function, and it resembles the  $|\delta|$  notation defined in Section 2.7.

All the operators  $+$ ,  $*$ ,  $\sqcup$ ,  $|\cdot|$  and the  $\sqsubseteq$  relation can be easily extended to  $\mathbb{D}^*$ .

**Example 7.4.** Given the set  $R_f = \{\rho_1, \rho_2, \rho_3\}$ , let us define:

$$\Delta_1 = \lambda x. \left[ \begin{array}{l} \rho_1 \mapsto x + 1 \\ \rho_2 \mapsto 2 \\ \rho_3 \mapsto x^2 + 3 \end{array} \right] \quad \Delta_2 = [x \geq 2 \rightarrow \Delta_1] \quad \tilde{\zeta} = \lambda x. 2x$$

then:

$$\Delta_1 + \Delta_2 = \lambda x. \left[ x \geq 2 \rightarrow \left[ \begin{array}{l} \rho_1 \mapsto 2x + 2 \\ \rho_2 \mapsto 4 \\ \rho_3 \mapsto 2x^2 + 6 \end{array} \right] \right]$$

$$\tilde{\zeta} * \Delta_1 = \lambda x. \left[ \begin{array}{l} \rho_1 \mapsto 2x^2 + 2x \\ \rho_2 \mapsto 4x \\ \rho_3 \mapsto 2x^3 + 6x \end{array} \right]$$

$$|\Delta_2| = \lambda x. [x \geq 2 \rightarrow x^2 + x + 6]$$

and  $\Delta_2 \sqsubseteq \Delta_1$  holds, but not  $\Delta_1 \sqsubseteq \Delta_2$ . □

**Proposition 7.5.** Given a fixed function definition  $f \bar{x}_i @ \bar{r}_j = e_f$ , its associated ordered set  $(\mathbb{D}, \sqsubseteq)$  is a complete lattice, whose bottommost element is  $\lambda \bar{x}_i. \perp$  and the topmost one is  $\lambda \bar{x}_i. \lambda \rho \in R_f. + \infty$ .

*Proof.* Since  $(\mathbb{R}_{\infty}^+, \leq)$  is a complete lattice, so is  $(R_f \rightarrow \mathbb{R}_{\infty}^+, \sqsubseteq)$  with the component-wise ordering [92, Section A.2]. As a consequence,  $((R_f \rightarrow \mathbb{R}_{\infty}^+)^+, \sqsubseteq)$  is a complete lattice, and so is  $\mathbb{D}$ , for the same reason. □

These definitions allow us to set up a correspondence between the actual memory consumption of an expression (given by the resource vector  $(\delta, m, s)$  resulting from its evaluation) and the results of our analysis, which will be triples of the form  $(\Delta, \mu, \sigma)$ , where  $\Delta \in \mathbb{D}$  (or  $\mathbb{D}^*$ , depending on the context) and  $\mu, \sigma \in \mathbb{F}$ . The abstract heap  $\Delta$  is meant to be an upper bound to the  $\delta$  component, whereas  $\mu$  and  $\sigma$  are upper bounds to the  $m$  and  $s$  components, respectively. The main difference is that the vector  $(\delta, m, s)$  represents a *particular* execution of  $e$  for a given input size, and  $(\Delta, \mu, \sigma)$  captures *every possible execution* of the expression for every possible input size. As a consequence, we can abstract the resource behaviour of a function definition by means of its function signature.

**Definition 7.6.** A function signature for  $f$  is a triple  $(\Delta, \mu, \sigma)$ , where  $\Delta$  belongs to  $\mathbb{D}$ , and  $\mu, \sigma$  belongs to  $\mathbb{F}$ .

The next step in this section is to precisely define what a *correct* function signature is. Assume the following  $\Downarrow$ -judgement of the body  $e_f$  of the function definition given in (7.1), whose derivation we call the *context derivation*.

$$E_0 \vdash h_0, k_0, e_f \Downarrow h_f, k_0, v_f, (\delta_f, m_f, s_f) \tag{7.2}$$

Intuitively, we expect a correct signature  $(\Delta, \mu, \sigma)$  to be an upper bound to the actual vector  $(\delta_f, m_f, s_f)$  for every input size. In order to check that  $\mu$  is a correct approximation of  $m_f$ , we compute the size of the function parameters from the value environment  $E_0$  and the heap  $h_0$ , so we define  $s_i \stackrel{\text{def}}{=} \text{size}(h_0, E_0(x_i))$

for every  $i \in \{1..n\}$ . If  $\mu \bar{s}_i^n \geq m_f$ , then  $\mu$  has given a correct result for this particular execution of  $e_f$ . If we are able to prove that this holds for every possible execution of  $e_f$ , then  $\mu$  correctly approximates the memory needs of  $e_f$ . For the  $\sigma$  we can proceed similarly. However, the case of  $\Delta$  is more involved, since we have to take into account the correspondence between the RTVs  $\rho_1 \dots \rho_m$  and the actual regions given by  $E_0(r_1) \dots E_0(r_m)$ . For this purpose we use the concept of region instantiation which was introduced when proving the correctness of *Safe's* type system (see Definition 3.15 on page 96). If  $\Gamma_0$  is the environment typing the expression  $e_f$ , then a region instantiation  $\eta$  is consistent with  $E_0$  and  $\Gamma_0$  it does not contradict the region instantiation defined by from  $E_0$  and  $\Gamma_0$ , i.e. common RTVs are bound to the same actual region:

$$E_0(r) = i \wedge \Gamma_0(r) = \rho \Rightarrow \eta(\rho) = i$$

In case  $\Gamma_0$  is injective, then  $\eta = (E_0 \circ \Gamma_0^{-1})|_{R_f \cup \{\rho_{self}\}}$  is the only region instantiation consistent with  $E_0$  and  $\Gamma_0$ . This is the case for our function bodies, since the region inference algorithm (Chapter 4) assigns different types to different region parameters. Henceforth we will abbreviate  $\eta = (E_0 \circ \Gamma_0^{-1})|_{R_f \cup \{\rho_{self}\}}$ .

**Definition 7.7.** Given a sequence of sizes  $\bar{s}_i^n$  of the input arguments of the context function  $f$ , a number  $k$  of regions and a region instantiation  $\eta$ , we say that:

- $\Delta$  is an upper bound to  $\delta$  in the context of  $\bar{s}_i^n$ ,  $k$  and  $\eta$ , denoted  $\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta$  iff

$$\forall j \in \{0..k\}. \sum_{\eta(\rho)=j} \Delta \bar{s}_i^n \rho \geq \delta(j)$$

- $\mu$  is an upper bound to  $m$  in the context of  $\bar{s}_i^n$ , denoted  $\mu \succeq_{\bar{s}_i^n} m$ , iff  $\mu \bar{s}_i^n \geq m$ .
- $\sigma$  is an upper bound to  $s$  in the context of  $\bar{s}_i^n$ , denoted  $\sigma \succeq_{\bar{s}_i^n} s$ , iff  $\sigma \bar{s}_i^n \geq s$ .

**Example 7.8.** Assume a function definition  $f \ x \ y @ r_1 \ r_2 = e_f$  and the execution of its body  $e_f$  under a environment  $E$  and a heap  $h$  with three regions (that is,  $k = 2$ ). We also assume that  $s_x = size(h, E(x)) = 4$ ,  $s_y = size(h, E(y)) = 2$  and that the function is typable under an environment  $\Gamma$  such that  $\Gamma(r_i) = \rho_i$  for  $i \in \{1, 2\}$ . If  $E(r_1) = 0$ ,  $E(r_2) = 0$  and  $E(self) = 1$ , then we get  $\eta = [\rho_1 \mapsto 0, \rho_2 \mapsto 0, \rho_{self} \mapsto 1]$ . Let us define  $\Delta$  as follows:

$$\Delta = \lambda x \ y. \begin{bmatrix} \rho_1 & \mapsto & 2x + y \\ \rho_2 & \mapsto & xy \\ \rho_{self} & \mapsto & 7x + 5 \end{bmatrix}$$

Then  $\Delta$  is a correct upper bound to  $\delta = [0 \mapsto 10, 1 \mapsto 32]$  in the context of  $s_x$ ,  $s_y$ ,  $k$  and  $\eta$ , since

$$\Delta 4 \ 2 = \begin{bmatrix} \rho_1 & \mapsto & 10 \\ \rho_2 & \mapsto & 8 \\ \rho_{self} & \mapsto & 32 \end{bmatrix} \text{ and } \begin{cases} 10 + 8 \geq \delta(0) \\ 32 \geq \delta(1) \end{cases}$$

□

The following properties on the  $\succeq$  will be useful in the following. In particular, we prove that the  $\succeq$  relation on  $\mathbb{D}$  (or  $\mathbb{D}^*$ ) is preserved by  $+$  and  $|\cdot|$  operators.

**Lemma 7.9.** If  $\Delta_1 \succeq_{\bar{s}_i^n, k, \eta} \delta_1$  and  $\Delta_2 \succeq_{\bar{s}_i^n, k, \eta} \delta_2$  then  $\Delta_1 + \Delta_2 \succeq_{\bar{s}_i^n, k, \eta} \delta_1 + \delta_2$ .

*Proof.* Let  $j \in \{0..k\}$ . Then:

$$\sum_{\eta(\rho')=j} (\Delta_1 + \Delta_2) \bar{s}_i^n \rho = \sum_{\eta(\rho')=j} \Delta_1 \bar{s}_i^n \rho + \sum_{\eta(\rho')=j} \Delta_2 \bar{s}_i^n \rho \geq \delta_1(j) + \delta_2(j) = (\delta_1 + \delta_2)(j)$$

□

**Lemma 7.10.** *If  $\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta$  then  $|\Delta| \bar{s}_i^n \geq |\delta|$*

*Proof.* We know that  $\text{dom } \eta = R_f$ , so for every  $\rho \in \text{dom } \eta$  there exists some  $i \in \{0..k\}$  such that  $\eta(\rho) = i$ . Thus we get:

$$|\Delta| \bar{s}_i^n = \sum_{\rho \in \text{dom } \eta} \Delta \bar{s}_i^n \rho = \sum_{i=0}^k \sum_{\eta(\rho)=i} \Delta \bar{s}_i^n \rho \geq \sum_{i=0}^k \delta(i) = |\delta|$$

□

Given these definitions, we are ready to give a formal notion of a correct function signature.

**Definition 7.11** (Correct signature). Let  $(\Delta, \mu, \sigma)$  be the signature of a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  and  $\Gamma$  the type environment inferred for  $e_f$ . This signature is said to be correct if for all  $h, h', k, E_f, \bar{v}_i^n, \bar{i}_j^m, v, \delta, m, s, \bar{s}_i^n, \eta$  such that:

1.  $E_f \vdash h, k+1, e_f \Downarrow h', k+1, v, (\delta, m, s)$  where  $E_f = [\bar{x}_i \mapsto \bar{v}_i^n, \bar{r}_j \mapsto \bar{i}_j^m, \text{self} \mapsto k+1]$
2. For each  $i \in \{1..n\}$ ,  $s_i = \text{size}(h, v_i)$
3.  $\eta$  is the consistent region instantiation determined by  $E_f$  and  $\Gamma$ .

then  $\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta|_k$ ,  $\mu \succeq_{\bar{s}_i^n} m$ , and  $\sigma \succeq_{\bar{s}_i^n} s$ .

### 7.3 Abstract interpretation

Once we have defined *what* a correct signature is, we need to know *how* to infer a correct signature for a given function definition. In this section we provide an abstract interpretation function which addresses this problem for non-recursive function definitions, and also serves as a basis for the inference of recursive function definitions.

Given an expression  $e$ , our aim is to find a tuple  $(\Delta, \mu, \sigma)$  which is an upper bound to its memory consumption. Firstly we have to distinguish between two kinds of *Core-Safe* expressions: literals, variables, copy expressions and constructor applications make up the set of *basic expressions*, whereas the set of *compound expressions* include, additionally, **let** and **case** expressions. The motivation for doing this is that basic expressions can be dealt with in an uniform way when inferring each of the components of the tuple  $(\Delta, \mu, \sigma)$ , while the processing of **let** and **case** expressions differs substantially depending on the specific component. Therefore, we redefine the grammar of *Core-Safe* expressions as follows:

$$\begin{aligned} \mathbf{BExp} \ni be &::= c \mid x \mid x @ r \mid a_1 \oplus a_2 \mid C \bar{a}_i @ r \mid f \bar{a}_i^n @ \bar{r}_j^m \\ \mathbf{Exp} \ni e &::= be \mid \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \mid \mathbf{case} \ x \ \mathbf{of} \ \bar{C}_i \ \bar{x}_{ij}^{n_i} \rightarrow e_i \end{aligned}$$

Let us start defining the abstract interpretation function for basic expressions. This function is defined as a set of rules in Figure 7.1. The interpretation of a basic expression  $\llbracket e \rrbracket$  receives a signature environment  $\Sigma : \mathbf{Fun} \rightarrow \mathbb{D} \times \mathbb{F} \times \mathbb{F}$  which maps each function name  $g$  appearing in  $e$  to its memory consumption signature  $(\Delta_g, \mu_g, \sigma_g)$ , as in Definition 7.6. It also receives a typing environment  $\Gamma : \mathbf{RegVar} \rightarrow \mathbf{RegType}$ , giving the type of each region variable in scope, and the statically determined length  $td$  of the runtime environment when calling a function application. The latter has the same meaning as in the resource-aware semantics of Figure 2.26. The interpretation function is also

$$\begin{array}{c}
\overline{\llbracket c \rrbracket \Sigma \Gamma td = ([ ]_f, 0, 1)} \\
\overline{\llbracket x \rrbracket \Sigma \Gamma td = ([ ]_f, 0, 1)} \\
\overline{\llbracket a_1 \oplus a_2 \rrbracket \Sigma \Gamma td = ([ ]_f, 0, 2)} \\
\overline{\llbracket x @ r \rrbracket \Sigma \Gamma td = ([\Gamma(r) \mapsto |x|]_f, |x|, 2)} \\
\overline{\llbracket C \bar{a}_i @ r \rrbracket \Sigma \Gamma td = ([\Gamma(r) \mapsto 1]_f, 1, 1)} \\
\begin{array}{l}
\Sigma(g) = (\Delta_g, \mu_g, \sigma_g) \quad \theta = \text{unify}(\Gamma, g, \bar{r}_j) \\
G = \lambda \bar{x}_j. \left( \Delta_g \overline{|a_i| \bar{x}_j^l} \neq \perp \wedge \mu_g \overline{|a_i| \bar{x}_j^l} \neq \perp \wedge \sigma_g \overline{|a_i| \bar{x}_j^l} \neq \perp \right) \\
\Delta = [G \rightarrow \theta \downarrow_{\overline{|a_i|}} \Delta_g] \quad \mu = [G \rightarrow \lambda \bar{x}_j. \mu_g (\overline{|a_i| \bar{x}_j^l})] \quad \sigma = [G \rightarrow \lambda \bar{x}_j. \sigma_g (\overline{|a_i| \bar{x}_j^l})]
\end{array} \\
\hline
\overline{\llbracket g \bar{a}_i^l @ \bar{r}_j^q \rrbracket \Sigma \Gamma td = (\Delta, \mu, \sqcup \{l + q, l + q - td + \sigma\})}
\end{array}$$

Figure 7.1: Abstract interpretation function for basic expressions.

parametric on the context function definition  $f$ . However, since we assume this definition fixed (and to avoid excessive subscripting) we consider this parameter implicit. The notation  $|x|$  denotes the size function  $(\mathbb{R}_\infty^+)^n \rightarrow \mathbb{R}_\infty^+$  associated with  $x$ . These functions are assumed to have been inferred by an external size analysis. If  $R_f$  is the set of the context function's region parameters, the  $[ ]_f$  notation stands for the abstract heap  $\lambda \bar{x}_i. \lambda \rho. 0$  where  $\rho \in R_f$ . Similarly, the binding  $[\rho' \mapsto \xi]_f$  denotes the following abstract heap:

$$[\rho' \mapsto \xi]_f \stackrel{\text{def}}{=} \lambda \bar{x}_i. \lambda \rho. \begin{cases} \xi \bar{x}_i & \text{if } \rho = \rho' \\ 0 & \text{otherwise} \end{cases}$$

The integer values occurring in the rules should be understood as constant functions (that is  $n \stackrel{\text{def}}{=} \lambda \bar{x}_i. n$ ). We omit the  $\lambda \bar{x}_i$  for the sake of simplicity.

Notice the similarity between the results of the abstract interpretation in Figure 7.1 and the resource vectors of the semantic rules (Figure 2.26 on page 53). The only rule worth explaining is that of function application. Given a function definition  $g \bar{y}_i^l @ \bar{r}_j^m$ , assume we want to infer a particular function application  $g \bar{a}_i^l @ \bar{r}_j^q$ . Firstly we retrieve the signature of  $g$  from the signature environment  $\Sigma$ , so let  $(\Delta_g, \mu_g, \sigma_g) = \Sigma(g)$ . Each component is a function which depends on the sizes of its  $l$  parameters  $\bar{y}_i^l$ , so we have to pass the sizes of the actual arguments  $\overline{|a_i|}$ , which, in turn, are functions of the parameters of the caller (that is why we have  $\overline{|a_i| \bar{x}_j^l}$ ). The guard  $G$  discards those values  $\bar{x}_i$  giving place to sizes  $\overline{|a_i| \bar{x}_j^l}$  not belonging to the domain of  $\Delta_g, \mu_g$  or  $\sigma_g$ . Notice that, with regard to the computation of  $\Delta$ , the type of the arguments in the function application may be instances of the function's type. Hence we have to find a correspondence between the RTVs of  $g$  and the RTVs of the particular instance used in the call. That is what the function *unify* does.

**Definition 7.12.** Given a type environment  $\Gamma$ , a function name  $g \in \mathbf{Fun}$  and a sequence of region variables  $\bar{r}_j^q$ , we say that  $\theta = \text{unify}(\Gamma, g, \bar{r}_j^q)$ , iff  $\Gamma(g) = \forall \bar{\alpha} \rho. \bar{t}_i^l \rightarrow \bar{\rho}_j^q \rightarrow t$ , and  $\forall j \in \{1..q\}. \theta(\rho_j) = \Gamma(r_j)$ .

If there are several RTVs of  $g$ 's type being mapped to the same RTV  $\rho$  in the function application, then the charges done to the region of type  $\rho$  are the sum of the charges made by  $g$  to the RTVs  $\rho'$  such

that  $\theta(\rho') = \rho$ . The  $\downarrow$  operator does this computation. It is defined as follows:

$$\theta \downarrow_{\overline{|a_i|}} \Delta = \lambda \overline{x_i}. \lambda \rho. \sum_{\rho' \in \theta^{-1}(\rho)} \Delta \overline{|a_i|} \overline{x_i}^l \rho'$$

where  $\rho \in R_f \cup \{\rho_{self}\}$ .

**Example 7.13.** Assume a function  $g$  of type

$$g :: \forall \alpha, \rho_1, \rho_2, \rho_3. [\alpha] @ \rho_1 \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow T \alpha @ \rho_2 \rho_3$$

and the function call  $g x @ self r_1 self$ , where  $x$  has type  $[Int]@_{\rho_{self}}$ ,  $r_1$  has type  $\rho$ , and  $self$  has type  $\rho_{self}$ . Let us compare the general type of  $g$  with the particular instance of the function application:

$$\begin{array}{l} \text{General type :} \quad [\alpha]@_{\rho_1} \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow T \alpha @ \rho_2 \rho_3 \\ \text{Function application:} \quad [Int]@_{\rho_{self}} \rightarrow \rho_{self} \rightarrow \rho \rightarrow \rho_{self} \rightarrow T Int @ \rho \rho_{self} \end{array}$$

Hence we obtain  $\theta = unify(\Gamma, g, self r_1 self) = [\rho_1 \mapsto \rho_{self}, \rho_2 \mapsto \rho, \rho_3 \mapsto \rho_{self}]$ . If  $\Delta_g$  is defined as follows:

$$\Delta_g = \lambda y. \begin{bmatrix} \rho_1 \mapsto y^2 + 3 \\ \rho_2 \mapsto 3y \\ \rho_3 \mapsto 2y + 1 \end{bmatrix}$$

and  $\xi = \lambda x. 2x$ , then we get:

$$\theta \downarrow_{\xi} \Delta_g = \lambda x. \begin{bmatrix} \rho_{self} \mapsto ((2x)^2 + 3) + (2 \cdot (2x) + 1) \\ \rho \mapsto 3 \cdot (2x) \end{bmatrix} = \lambda x. \begin{bmatrix} \rho_{self} \mapsto 4x^2 + 4x + 4 \\ \rho \mapsto 6x \end{bmatrix}$$

□

So far we have obtained a tuple  $(\theta \downarrow_{\overline{|a_i|} \overline{x_i}^l} \Delta_g, \mu \overline{|a_i|} \overline{x_i}, \sigma \overline{|a_i|} \overline{x_i})$ , which is an upper bound to the memory costs of the function's *body*. From these we can easily compute the costs of the function application itself: we just have to proceed as in rule  $[App]$  of Figure 2.26. We use the strict  $\sqcup$  operator, since we want the stack cost function of the whole application to be undefined in those values where  $G$  does not hold. If we used  $\sqcup$  instead we would get in these values  $\sqcup\{l + q, l + q - td + \perp\} = l + q$ , where an undefined result is expected.

In the following we shall use the notation  $\llbracket be \rrbracket_{\Delta} \Sigma \Gamma td$ ,  $\llbracket be \rrbracket_{\mu} \Sigma \Gamma td$  and  $\llbracket be \rrbracket_{\sigma} \Sigma \Gamma td$  to refer to the first, second and third components of  $\llbracket be \rrbracket \Sigma \Gamma td$  respectively. In the first two cases we will omit the  $td$  since they are not relevant to the computation of  $\Delta$  and  $\mu$ , whereas with the third component we omit the  $\Gamma$  environment for the same reason. Sometimes we shall even leave out entirely the  $\Sigma$  and  $\Gamma$  when they are clear from the context.

An important result is the fact that the results given by the  $\llbracket \cdot \rrbracket_{\Delta}$ ,  $\llbracket \cdot \rrbracket_{\mu}$ , and  $\llbracket \cdot \rrbracket_{\sigma}$  abstract interpretations, when applied to basic expressions, have the same definition domain.

**Lemma 7.14.** For every basic expression  $be$  and every  $\Gamma, \Sigma, td$  we get:

$$dom \llbracket be \rrbracket_{\Delta} \Sigma \Gamma = dom \llbracket be \rrbracket_{\mu} \Sigma \Gamma = dom \llbracket be \rrbracket_{\sigma} \Sigma td$$

*Proof.* If  $be$  is not a function application the Lemma follows trivially, since the definition domain is  $\mathbb{R}^+$

in these cases. If  $be$  is a function application, it holds because the results of  $\llbracket be \rrbracket_\Delta$ ,  $\llbracket be \rrbracket_\mu$  and  $\llbracket be \rrbracket_\sigma$  are guarded by the same  $G$ , and the three are always distinct from  $\perp$  if the guard holds.  $\square$

For inferring heap space consumption in presence of **let** and **case** expressions we have to apply a *flattening* transformation to the expression being analysed. This works as follows: an expression  $e$  is transformed into a set of *sequences* of basic expressions. Each sequence represents a possible execution flow of the expression. The whole set of sequences capture all the possible execution flows that may arise when executing this expression. This transformation destroys the structure of the program: instead of having nested expressions, the result comprises all the basic expressions being executed into a single “level” of nesting (that is why this transformation is called flattening). However, this transformation does not affect heap space consumption, since the latter does not depend on the structure of the expression. The following example provides an intuition on this fact.

**Example 7.15.** Given the following expressions:

$$e \equiv \mathbf{let} \ x_1 = be_1 \ \mathbf{in} \ (\mathbf{let} \ x_2 = be_2 \ \mathbf{in} \ be_3)$$

$$e' \equiv \mathbf{let} \ x_2 = (\mathbf{let} \ x_1 = be_1 \ \mathbf{in} \ be_2) \ \mathbf{in} \ be_3$$

Let  $(\delta_i, m_i, s_i)$  the resource vector associated to the execution of the expression  $be_i$  for  $i \in \{1..3\}$ . If  $x_1 \notin \text{fv}(be_3)$  and we can execute  $e$ , we can also execute  $e'$ . Let  $(\delta, m, s)$  and  $(\delta', m', s')$  be the resources of the execution of  $e$  and  $e'$  respectively. With the  $\delta$  component, we get:

$$\delta = \delta_1 + (\delta_2 + \delta_3) \quad \delta' = (\delta_1 + \delta_2) + \delta_3$$

with regard to the  $m$  component we obtain:

$$m = \sqcup\{m_1, |\delta_1| + \sqcup\{m_2, |\delta_2| + m_3\}\} \quad m' = \sqcup\{\sqcup\{m_1, |\delta_1| + m_2\}, |\delta_1 + \delta_2| + m_3\}$$

The equality  $\delta = \delta'$  follows trivially. The equality  $m = m'$  follows from the fact that  $|\delta_1 + \delta_2| = |\delta_1| + |\delta_2|$  and the associativity of the  $\sqcup$  operator, so we get:

$$m = m' = \sqcup\{m_1, |\delta_1| + m_2, |\delta_1| + |\delta_2| + m_3\} \quad (7.3)$$

It is easy to show that, in both cases, the expression  $be_1$  is the first to be executed, then  $be_2$  and  $be_3$  comes in the last place. That is why we can represent both expressions as a single sequence  $[be_1, be_2, be_3]$ .  $\square$

When considering sequences of expressions we lose the pattern guards of the **case** expressions. Nevertheless, these guards provide useful information on the size of the DS being matched against. For instance, assume the following expression:

$$\begin{array}{l} \mathbf{case} \ x \ \mathbf{of} \\ \quad [] \rightarrow e_1 \\ \quad (x : xx) \rightarrow e_2 \end{array}$$

If the branch  $e_2$  is executed, we know for sure<sup>2</sup> that the size of  $x$  must be greater or equal than 2. In general, this size information will be included in the execution sequences as *guards*, specifying under

<sup>2</sup>At this point we assume the absence of dangling pointers in the heap, since the expression is well-typed.

$$\begin{aligned}
seqs\ be &= \{[be]\} \\
seqs\ (\mathbf{let}\ x_1 = e_1\ \mathbf{in}\ e_2) &= \{seq_1 ++ seq_2 \mid seq_1 \in seqs\ e_1, seq_2 \in seqs\ e_2\} \\
seqs\ (\mathbf{case}\ x\ \mathbf{of}\ C_i\ \overline{x_{ij}^{n_i}} \rightarrow e_i) &= \bigcup_{i=1}^n (seqs\ e_i \wedge \lambda \overline{x_i}. (|\overline{x}| \geq 1 + |RecPos(C_i)|))
\end{aligned}$$

Figure 7.2: Definition of *seqs*, which converts an expression into a set of sequences.

which sizes of the **case** discriminant the execution of each sequence may take place. This motivates the following definition:

**Definition 7.16.** A sequence is a list of basic expressions  $be_1, \dots, be_n$  preceded by a boolean function  $G$ . We use the notation  $[G \rightarrow be_1, \dots, be_n]$  to denote sequences. The notation  $[be_1, \dots, be_n]$  stands for the sequence  $[true \rightarrow be_1, \dots, be_n]$ .

We define the concatenation ( $++$ ) of sequences as follows:

$$[G_1 \rightarrow be_1, \dots, be_n] ++ [G_2 \rightarrow be'_1, \dots, be'_m] = [G_1 \wedge G_2 \rightarrow be_1, \dots, be_n, be'_1, \dots, be'_m]$$

If *seq* is a sequence, we can strengthen its guard with the notation  $seq \wedge G$ , which stands for

$$[G_1 \rightarrow be_1, \dots, be_n] \wedge G = [G_1 \wedge G \rightarrow be_1, \dots, be_n]$$

and this notation can be extended to sets of sequences as follows:

$$S \wedge G = \{seq \wedge G \mid seq \in S\}$$

The function *seqs* transforms a *Core-Safe* expression into a set of sequences representing all the possible execution paths. It is defined in Figure 7.2. In order to transform a **let** expression, *seqs* gathers all the sequences of the auxiliary expression  $e_1$  and the main expression  $e_2$  and considers all the combinations. With respect to **case** expressions, it collects the sequences of each branch and adds the corresponding size guard imposed by the recursive positions of the pattern: the size of the discriminant must be one plus the number of recursive positions of its constructor.

**Example 7.17.** Assume the function *append* for joining two lists:

$$\begin{aligned}
append\ xs\ ys\ @\ r &= \mathbf{case}\ xs\ \mathbf{of} \\
&\quad [] \rightarrow ys \\
&\quad (x : xx) \rightarrow \mathbf{let}\ x_1 = append\ xx\ ys\ @\ r\ \mathbf{in}\ (x : x_1)@r
\end{aligned}$$

Its body  $e_{append}$  can be flattened into two sequences as follows:

$$seqs\ e_{append} = \{[xs \geq 1 \rightarrow ys], [xs \geq 2 \rightarrow append\ xx\ ys, (x : x_1)@r]\}$$

□

The extension of the abstract interpretation rules to compound expressions will be done in terms of their decomposition into sequences of basic expressions. Therefore our first step is to define the  $\llbracket \cdot \rrbracket_\Delta$  and  $\llbracket \cdot \rrbracket_\mu$  interpretations of such sequences. Example 7.15 gives us an intuition on how to define these interpretations. There we had three basic expressions with their associated costs  $(\delta_1, m_1)$ ,  $(\delta_2, m_2)$  and  $(\delta_3, m_3)$ . The  $\delta$  component is additive, in the sense that the  $\delta$  component of the whole compound

expression is the sum of the  $\delta$ 's of its components. It is reasonable to think that its abstract counterpart ( $\Delta$ ) should be additive, as well. Hence, if  $(\Delta_1, \mu_1)$ ,  $(\Delta_2, \mu_2)$  and  $(\Delta_3, \mu_3)$  are the inferred costs of the respective basic expressions, we would obtain the following result:

$$\llbracket [be_1, be_2, be_3] \rrbracket_{\Delta} = \Delta_1 + \Delta_2 + \Delta_3 \quad (7.4)$$

With regard to the  $\mu$  component, from the expression (7.3) we can define an “abstract” counterpart as follows:

$$\llbracket [be_1, be_2, be_3] \rrbracket_{\mu} = \sqcup \{ \mu_1, |\Delta_1| + \mu_2, |\Delta_1| + |\Delta_2| + \mu_3 \} \quad (7.5)$$

The intuition behind this expression is depicted in Figure 2.30, and its extension to three expressions. The reason of choosing  $\sqcup$  instead of  $\sqcap$  is that, if the costs of one of the basic expressions return an undefined value for some input sizes, we want to cancel out the whole sequence.

Now we can proceed to the generalization of the expressions given in (7.4) and (7.5) to guarded sequences of an arbitrary number of expressions. This is done as follows:

$$\begin{aligned} \llbracket [G \rightarrow be_1, \dots, be_n] \rrbracket_{\Delta} \Sigma \Gamma &= [G \rightarrow \sum_{i=1}^n (\llbracket be_i \rrbracket_{\Delta} \Sigma \Gamma)] \\ \llbracket [G \rightarrow be_1, \dots, be_n] \rrbracket_{\mu} \Sigma \Gamma &= \left[ G \rightarrow \sqcup_{i=1}^n \left( \sum_{j=1}^{i-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma \Gamma| + \llbracket be_i \rrbracket_{\mu} \Sigma \Gamma \right) \right] \end{aligned}$$

It is easy to see that Lemma 7.14 can be extended when applying the interpretation to sequences of expressions:

**Lemma 7.18.** *For every sequence of expressions  $seq$ , every  $\Sigma$  and  $\Gamma$ , we get:*

$$dom (\llbracket seq \rrbracket_{\Delta} \Sigma \Gamma) = dom (\llbracket seq \rrbracket_{\mu} \Sigma \Gamma)$$

*Proof.* Let  $seq = [G \rightarrow be_1, \dots, be_n]$ . Firstly we prove the  $\subseteq$  inclusion. Let  $\bar{x}_i \in dom (\llbracket seq \rrbracket_{\Delta} \Sigma \Gamma)$ . Then:

1.  $G(\bar{x}_i)$  holds, and
2.  $\bar{x}_i \in dom \llbracket be_i \rrbracket_{\Delta}$  for each  $i \in \{1..n\}$ , which implies:
  - (a)  $\bar{x}_i \in dom |\llbracket be_i \rrbracket_{\Delta}|$  for each  $i \in \{1..n\}$ , by definition of  $|\cdot|$  operator.
  - (b)  $\bar{x}_i \in dom \llbracket be_i \rrbracket_{\mu}$  for each  $i \in \{1..n\}$ , by Lemma 7.14.

Hence  $\bar{x}_i \in dom \llbracket seq \rrbracket_{\mu}$ , by (1), (2a) and (2b). The  $\supseteq$  inclusion can be proven similarly. □

**Example 7.19.** Back to our *append* function, assume a signature  $\Sigma$  in which  $\Sigma(append) = (f_{\Delta}, f_{\mu}, \lambda xs ys.0)$  for some functions  $f_{\Delta} \in \mathbb{ID}$  and  $f_{\mu} \in \mathbb{IF}$  such that  $dom f_{\Delta} = \{(xs, ys) \mid xs \geq 2\}$  and  $dom f_{\mu} = \{(xs, ys) \mid xs \geq 3\}$ . Given the flattening of  $e_{append}$  shown in Example 7.17:

$$seqs e_{append} = \underbrace{\{ [xs \geq 1 \rightarrow ys] \}}_{seq_1}, \underbrace{\{ [xs \geq 2 \rightarrow append \ xx \ ys, (x : x_1)@r] \}}_{seq_2}$$

Let us apply the  $\llbracket \cdot \rrbracket_{\Delta}$  and  $\llbracket \cdot \rrbracket_{\mu}$  interpretations to  $seq_1$  and  $seq_2$  under an environment  $\Gamma = [r \mapsto \rho_2]$  and assuming that  $|xx| = \lambda xs ys.xs - 1$ . We start with the first sequence:

$$\begin{aligned}\llbracket ys \rrbracket_{\Delta} &= \lambda xs \ ys. [\rho_2 \mapsto 0] \\ \llbracket seq_1 \rrbracket_{\Delta} &= \lambda xs \ ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 0]]\end{aligned}$$

$$\begin{aligned}\llbracket ys \rrbracket_{\mu} &= \lambda xs \ ys. 0 \\ \llbracket seq_1 \rrbracket_{\mu} &= \lambda xs \ ys. [xs \geq 1 \rightarrow 0]\end{aligned}$$

whereas with the second sequence we get the following guard,

$$\begin{aligned}G &= \lambda xs \ ys. f_{\Delta}(xs) \neq \perp \wedge f_{\mu}(xs) \neq \perp \\ &= \lambda xs \ ys. xs \geq 2 \wedge xs \geq 3 \\ &= \lambda xs \ ys. xs \geq 3\end{aligned}$$

which is used in the respective interpretations of the *append* function application:

$$\begin{aligned}\llbracket append \ xs \ ys \ @ \ r \rrbracket_{\Delta} &= \lambda xs \ ys. [G \ (xs - 1) \ ys \rightarrow f_{\Delta} \ (xs - 1) \ ys] \\ &= \lambda xs \ ys. [xs - 1 \geq 3 \rightarrow [\rho_2 \mapsto f_{\Delta} \ (xs - 1) \ ys \ \rho_2]] \\ &= \lambda xs \ ys. [xs \geq 4 \rightarrow [\rho_2 \mapsto f_{\Delta} \ (xs - 1) \ ys \ \rho_2]]\end{aligned}$$

$$\llbracket append \ xs \ ys \ @ \ r \rrbracket_{\mu} = \lambda xs \ ys. [xs \geq 4 \rightarrow f_{\mu} \ (xs - 1) \ ys]$$

$$\llbracket (x : x_1) @ r \rrbracket_{\Delta} = \lambda xs \ ys. [\rho_2 \mapsto 1]$$

$$\llbracket (x : x_1) @ r \rrbracket_{\mu} = \lambda xs \ ys. 1$$

$$\begin{aligned}\llbracket seq_2 \rrbracket_{\Delta} &= \lambda xs \ ys. [xs \geq 2 \rightarrow [xs \geq 4 \rightarrow [\rho_2 \mapsto f_{\Delta} \ (xs - 1) \ ys \ \rho_2]] + [\rho_2 \mapsto 1]] \\ &= \lambda xs \ ys. [xs \geq 2 \rightarrow [xs \geq 4 \rightarrow [\rho_2 \mapsto 1 + f_{\Delta} \ (xs - 1) \ ys \ \rho_2]]] \\ &= \lambda xs \ ys. [xs \geq 4 \rightarrow [\rho_2 \mapsto 1 + f_{\Delta} \ (xs - 1) \ ys \ \rho_2]]\end{aligned}$$

$$\begin{aligned}\llbracket seq_2 \rrbracket_{\mu} &= \lambda xs \ ys. [xs \geq 2 \rightarrow \sqcup \{ [xs \geq 4 \rightarrow f_{\mu} \ (xs - 1) \ ys], [xs \geq 4 \rightarrow f_{\Delta} \ (xs - 1) \ ys \ \rho_2 + 1] \}] \\ &= \lambda xs \ ys. [xs \geq 2 \rightarrow [xs \geq 4 \rightarrow \sqcup \{ f_{\mu} \ (xs - 1) \ ys, f_{\Delta} \ (xs - 1) \ ys \ \rho_2 + 1 \}]] \\ &\quad \lambda xs \ ys. [xs \geq 4 \rightarrow \sqcup \{ f_{\mu} \ (xs - 1) \ ys, f_{\Delta} \ (xs - 1) \ ys \ \rho_2 + 1 \}]\end{aligned}$$

Notice that, for every sequence, both interpretations have the same domain. □

Sometimes it is useful to express the  $\llbracket \cdot \rrbracket_{\Delta}$  and  $\llbracket \cdot \rrbracket_{\mu}$  interpretations of a sequence in terms of its subsequences. This is given by the following Lemma.

**Lemma 7.20.** *Given two sequences  $seq_1$  and  $seq_2$ , the following facts hold:*

1.  $\llbracket seq_1 ++ seq_2 \rrbracket_{\Delta} \Sigma \Gamma = \llbracket seq_1 \rrbracket_{\Delta} \Sigma \Gamma + \llbracket seq_2 \rrbracket_{\Delta} \Sigma \Gamma$
2.  $\llbracket seq_1 ++ seq_2 \rrbracket_{\mu} \Sigma \Gamma = \sqcup \{ \llbracket seq_1 \rrbracket_{\mu} \Sigma \Gamma, |\llbracket seq_1 \rrbracket_{\Delta} \Sigma \Gamma| + \llbracket seq_2 \rrbracket_{\mu} \Sigma \Gamma \}$

*Proof.* Let us assume  $seq_1 = [G_1 \rightarrow be_1, \dots, be_k]$  and  $seq_2 = [G_2 \rightarrow be'_1, \dots, be'_m]$ . We also define  $[be_{k+1}, \dots, be_{k+m}] = [be'_1, \dots, be'_m]$ . Firstly we prove that the functions occurring in both sides of each equation have the same domain. With regard to the first fact we get

$$\begin{aligned}
& \bar{x}_i \in \text{dom } \llbracket seq_1 ++ seq_2 \rrbracket_\Delta \\
\Leftrightarrow & G_1 \bar{x}_i \wedge G_2 \bar{x}_i \wedge \forall j \in \{1..k+m\}. \bar{x}_i \in \text{dom } \llbracket be_j \rrbracket_\Delta \\
\Leftrightarrow & G_1 \bar{x}_i \wedge \forall j \in \{1..k\}. \bar{x}_i \in \text{dom } \llbracket be_j \rrbracket_\Delta \wedge G_2 \bar{x}_i \wedge \forall j \in \{k+1..k+m\}. \bar{x}_i \in \text{dom } \llbracket be_j \rrbracket_\Delta \\
\Leftrightarrow & G_1 \bar{x}_i \wedge \forall j \in \{1..k\}. \bar{x}_i \in \text{dom } \llbracket be_j \rrbracket_\Delta \wedge G_2 \bar{x}_i \wedge \forall l \in \{1..m\}. \bar{x}_i \in \text{dom } \llbracket be'_l \rrbracket_\Delta \\
\Leftrightarrow & \bar{x}_i \in \text{dom } \llbracket seq_1 \rrbracket_\Delta \wedge \bar{x}_i \in \text{dom } \llbracket seq_2 \rrbracket_\Delta \\
\Leftrightarrow & \bar{x}_i \in \text{dom } (\llbracket seq_1 \rrbracket_\Delta + \llbracket seq_2 \rrbracket_\Delta)
\end{aligned}$$

whereas in the second one we get:

$$\begin{aligned}
& \bar{x}_i \in \text{dom } \llbracket seq_1 ++ seq_2 \rrbracket_\mu \\
\Leftrightarrow & G_1 \bar{x}_i \wedge G_2 \bar{x}_i \wedge \forall j \in \{1..k+m\}. \bar{x}_i \in \text{dom } \sum_{l=1}^{j-1} |\llbracket be_l \rrbracket_\Delta| + \llbracket be_j \rrbracket_\mu \\
\Leftrightarrow & G_1 \bar{x}_i \wedge \forall j \in \{1..k\}. \bar{x}_i \in \text{dom } \sum_{l=1}^{j-1} |\llbracket be_l \rrbracket_\Delta| + \llbracket be_j \rrbracket_\mu \\
& \wedge G_2 \bar{x}_i \wedge \forall j \in \{1..m\}. \bar{x}_i \in \text{dom } \sum_{l=1}^{j-1} |\llbracket be'_l \rrbracket_\Delta| + \llbracket be'_j \rrbracket_\mu \\
\Leftrightarrow & G_1 \bar{x}_i \wedge \forall j \in \{1..k\}. \bar{x}_i \in \text{dom } \sum_{l=1}^{j-1} |\llbracket be_l \rrbracket_\Delta| + \llbracket be_j \rrbracket_\mu \\
& \wedge G_2 \bar{x}_i \wedge \forall j \in \{1..m\}. \bar{x}_i \in \text{dom } \sum_{l=1}^{j-1} |\llbracket be'_l \rrbracket_\Delta| + \llbracket be'_j \rrbracket_\mu \\
& \wedge \forall j \in \{1..k\}. \bar{x}_i \in \text{dom } |\llbracket be_j \rrbracket_\Delta| \\
\Leftrightarrow & \bar{x}_i \in \text{dom } \llbracket seq_1 \rrbracket_\mu \wedge \bar{x}_i \in \text{dom } \llbracket seq_2 \rrbracket_\mu \wedge \bar{x}_i \in \text{dom } |\llbracket seq_1 \rrbracket_\Delta| \\
\Leftrightarrow & \bar{x}_i \in \text{dom } \sqcup \{ \llbracket seq_1 \rrbracket_\mu, |\llbracket seq_1 \rrbracket_\Delta| + \llbracket seq_2 \rrbracket_\mu \}
\end{aligned}$$

Now let us assume that  $D = \text{dom } \llbracket seq_1 ++ seq_2 \rrbracket_\Delta$  (and hence  $D = \text{dom } \llbracket seq_1 ++ seq_2 \rrbracket_\mu$ , by Lemma 7.14). If we restrict ourselves to this definition domain  $D$ , we can prove the first fact as follows:

$$\begin{aligned}
\llbracket seq_1 ++ seq_2 \rrbracket_\Delta &= \llbracket [be_1, \dots, be_k, be_{k+1}, \dots, be_{k+m}] \rrbracket_\Delta \\
&= \sum_{i=1}^{k+m} \llbracket be_i \rrbracket_\Delta \\
&= \sum_{i=1}^k \llbracket be_i \rrbracket_\Delta + \sum_{i=k+1}^{k+m} \llbracket be_i \rrbracket_\Delta \\
&= \sum_{i=1}^k \llbracket be_i \rrbracket_\Delta + \sum_{j=1}^m \llbracket be_{k+j} \rrbracket_\Delta \\
&= \sum_{i=1}^k \llbracket be_i \rrbracket_\Delta + \sum_{j=1}^m \llbracket be'_j \rrbracket_\Delta \\
&= \llbracket seq_1 \rrbracket_\Delta + \llbracket seq_2 \rrbracket_\Delta \quad (\text{in } D)
\end{aligned}$$

Under the same conditions, we prove the second fact:

$$\begin{aligned}
\llbracket seq_1 ++ seq_2 \rrbracket_\mu &= \llbracket [be_1, \dots, be_k, be_{k+1}, \dots, be_{k+m}] \rrbracket_\mu \\
&= \sqcup_{i=1}^{k+m} \left( \sum_{j=1}^{i-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_i \rrbracket_\mu \right) \\
&= \sqcup \left\{ \sqcup_{i=1}^k \left( \sum_{j=1}^{i-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_i \rrbracket_\mu \right), \sqcup_{i=k+1}^{k+m} \left( \sum_{j=1}^{i-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_i \rrbracket_\mu \right) \right\} \\
&= \sqcup \left\{ \llbracket seq_1 \rrbracket_\mu, \sqcup_{i=k+1}^{k+m} \left( \sum_{j=1}^k |\llbracket be_j \rrbracket_\Delta| + \sum_{j=k+1}^{i-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_i \rrbracket_\mu \right) \right\} \\
&= \sqcup \left\{ \llbracket seq_1 \rrbracket_\mu, \sum_{j=1}^k |\llbracket be_j \rrbracket_\Delta| + \sqcup_{i=k+1}^{k+m} \left( \sum_{j=k+1}^{i-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_i \rrbracket_\mu \right) \right\} \\
&= \sqcup \left\{ \llbracket seq_1 \rrbracket_\mu, \sum_{j=1}^k |\llbracket be_j \rrbracket_\Delta| + \sqcup_{l=1}^m \left( \sum_{j=k+1}^{k+l-1} |\llbracket be_j \rrbracket_\Delta| + \llbracket be_{k+l} \rrbracket_\mu \right) \right\} \\
&= \sqcup \left\{ \llbracket seq_1 \rrbracket_\mu, |\llbracket seq_1 \rrbracket_\Delta| + \sqcup_{l=1}^m \left( \sum_{p=1}^{l-1} |\llbracket be_{k+p} \rrbracket_\Delta| + \llbracket be'_l \rrbracket_\mu \right) \right\} \\
&= \sqcup \left\{ \llbracket seq_1 \rrbracket_\mu, |\llbracket seq_1 \rrbracket_\Delta| + \sqcup_{l=1}^m \left( \sum_{p=1}^{l-1} |\llbracket be'_p \rrbracket_\Delta| + \llbracket be'_l \rrbracket_\mu \right) \right\} \\
&= \sqcup \{ \llbracket seq_1 \rrbracket_\mu, |\llbracket seq_1 \rrbracket_\Delta| + \llbracket seq_2 \rrbracket_\mu \} \quad (\text{in } D)
\end{aligned}$$

□

In case a compound expression gives place to several sequences, we have to take the least upper bound of all of them. So, the abstract interpretation of a set  $S$  of sequences is defined as follows:

$$\begin{aligned} \llbracket S \rrbracket_{\Delta} \Sigma \Gamma &= \bigsqcup_{seq \in S} \llbracket seq \rrbracket_{\Delta} \Sigma \Gamma \\ \llbracket S \rrbracket_{\mu} \Sigma \Gamma &= \bigsqcup_{seq \in S} \llbracket seq \rrbracket_{\mu} \Sigma \Gamma \end{aligned}$$

Finally, the  $\Delta$  and  $\mu$  components of the abstract interpretation of a compound expression are given by the following definition:

$$\begin{aligned} \llbracket e \rrbracket_{\Delta} \Sigma \Gamma &= \llbracket seqs \ e \rrbracket_{\Delta} \Sigma \Gamma \\ \llbracket e \rrbracket_{\mu} \Sigma \Gamma &= \llbracket seqs \ e \rrbracket_{\mu} \Sigma \Gamma \end{aligned}$$

**Example 7.21.** Given the same signature and type environments of Example 7.19, we get:

$$\begin{aligned} \llbracket e_{append} \rrbracket_{\Delta} &= \lambda xs \ ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 0]] \sqcup [xs \geq 4 \rightarrow [\rho_2 \mapsto 1 + f_{\Delta}(xs - 1) \ ys \ \rho_2]] \\ \llbracket e_{append} \rrbracket_{\mu} &= \lambda xs \ ys. [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 4 \rightarrow \sqcup \{f_{\mu}(xs - 1) \ ys, f_{\Delta}(xs - 1) \ ys \ \rho_2 + 1\}] \end{aligned}$$

both of which can be expressed with piecewise functions as follows:

$$\begin{aligned} \llbracket e_{append} \rrbracket_{\Delta} &= \lambda xs \ ys. \begin{cases} \perp & xs < 1 \\ [\rho_2 \mapsto 0] & 1 \leq xs < 4 \\ [\rho_2 \mapsto 1 + f_{\Delta}(xs - 1) \ ys \ \rho_2] & 4 \leq xs \end{cases} \\ \llbracket e_{append} \rrbracket_{\mu} &= \lambda xs \ ys. \begin{cases} \perp & xs < 1 \\ 0 & 1 \leq xs < 4 \\ \sqcup \{f_{\mu}(xs - 1) \ ys, f_{\Delta}(xs - 1) \ ys \ \rho_2 + 1\} & 4 \leq xs \end{cases} \end{aligned}$$

□

Regarding the  $\sigma$  component for bounding the stack costs, we cannot apply the flattening-based approach, since this transformation breaks the structure of an expression and, unlike heap costs, the stack costs *do* depend on this structure, as the following example shows.

**Example 7.22.** Back to the expressions  $e$  and  $e'$  of Example 7.15, and assuming that  $s_i$  represent the stack costs of  $be_i$ , for  $i \in \{1, 2, 3\}$ . We get for  $e$  the following costs:

$$s = \sqcup \{2 + s_1, 1 + \sqcup \{2 + s_2, 1 + s_3\}\} = \sqcup \{2 + s_1, 3 + s_2, 2 + s_3\}$$

whereas with  $e'$  we obtain:

$$s' = \sqcup \{2 + \sqcup \{2 + s_1, 1 + s_2\}, 1 + s_3\} = \sqcup \{4 + s_1, 3 + s_2, 1 + s_3\}$$

which is clearly different from  $s$ .

□

Instead of defining the  $\llbracket \cdot \rrbracket_{\sigma}$  interpretation as a function of the sequences originated from a compound expressions, we have to define it in terms of the compound expression itself. The  $\llbracket \cdot \rrbracket_{\sigma}$  interpreta-

tion of **let** expressions shown below roughly resembles its counterpart in the resource-aware semantics of Figure 2.26. With respect to **case** expressions, we proceed in a similar way, but we take the least upper bound of all the branches and include the corresponding guard:

$$\begin{aligned} \llbracket \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \rrbracket_{\sigma} \Sigma \ td &= \sqcup \{2 + \llbracket e_1 \rrbracket_{\sigma} \Sigma \ 0, 1 + \llbracket e_2 \rrbracket_{\sigma} \Sigma \ (td + 1)\} \\ \llbracket \mathbf{case} \ x \ \mathbf{of} \ \overline{C_i \ x_{ij}^{n_i} \rightarrow e_i^n} \rrbracket_{\sigma} \Sigma \ td &= \sqcup_{i=1}^n [|x| \geq 1 + |\mathit{RecPos}(C_i)| \rightarrow n_i + \llbracket e_i \rrbracket_{\sigma} \Sigma \ (td + n_i)] \end{aligned}$$

**Example 7.23.** The  $\llbracket \cdot \rrbracket_{\sigma}$  interpretation of our running example  $e_{\mathit{append}}$  leads to the following results with  $td = 3$ , if we assume that  $\Sigma(\mathit{append}) = ([\ ]_f, 0, f_{\sigma})$  such that  $\mathit{dom} \ f_{\sigma} = \{(xs, ys) \mid xs \geq 1\}$ :

$$\begin{aligned} \llbracket e_{\mathit{append}} \rrbracket_{\sigma} \Sigma \ 3 &= [xs \geq 1 \rightarrow \llbracket ys \rrbracket_{\sigma} \Sigma \ 3] \sqcup [xs \geq 2 \rightarrow 2 + \llbracket \mathbf{let} \ x_1 = \dots \rrbracket_{\sigma} \Sigma \ 5] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 2 + \sqcup \{2 + \llbracket \mathit{append} \ xx \ ys \ @ \ r \rrbracket_{\sigma} \Sigma \ 0, 1 + \llbracket (x : x_1) @ r \rrbracket_{\sigma} \Sigma \ 6\}] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 2 + \sqcup \{2 + \sqcup \{3, 3 + [xs - 1 \geq 1 \rightarrow f_{\sigma}(xs - 1) \ ys]\}, 1 + 1\}] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 2 + \sqcup \{[xs \geq 2 \rightarrow \sqcup \{5, 5 + f_{\sigma}(xs - 1) \ ys]\}, 2\}] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 2 + \sqcup \{[xs \geq 2 \rightarrow 5 + f_{\sigma}(xs - 1) \ ys], 2\}] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow \sqcup \{7 + f_{\sigma}(xs - 1) \ ys, 7\}] \\ &= [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 7 + f_{\sigma}(xs - 1) \ ys] \end{aligned}$$

□

One may wonder why we have not followed the same approach with the  $\llbracket \cdot \rrbracket_{\Delta}$  and  $\llbracket \cdot \rrbracket_{\mu}$  interpretations. We could have directly defined these in terms of the **let** and **case** expressions, without any kind of flattening:

$$\begin{aligned} \llbracket \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \rrbracket_{\Delta} \Sigma \ \Gamma &= \llbracket e_1 \rrbracket_{\Delta} \Sigma \ \Gamma + \llbracket e_2 \rrbracket_{\Delta} \Sigma \ \Gamma \\ \llbracket \mathbf{case} \ x \ \mathbf{of} \ \overline{C_i \ x_{ij}^{n_i} \rightarrow e_i^n} \rrbracket_{\Delta} \Sigma \ \Gamma &= \sqcup_{i=1}^n [|x| \geq 1 + |\mathit{RecPos}(C_i)| \rightarrow \llbracket e_i \rrbracket_{\Delta} \Sigma \ \Gamma] \\ \llbracket \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \rrbracket_{\mu} \Sigma \ \Gamma &= \sqcup \{ \llbracket e_1 \rrbracket_{\mu} \Sigma \ \Gamma, |\llbracket e_1 \rrbracket_{\Delta} \Sigma \ \Gamma| + \llbracket e_2 \rrbracket_{\mu} \Sigma \ \Gamma \} \\ \llbracket \mathbf{case} \ x \ \mathbf{of} \ \overline{C_i \ x_{ij}^{n_i} \rightarrow e_i^n} \rrbracket_{\mu} \Sigma \ \Gamma &= \sqcup_{i=1}^n [|x| \geq 1 + |\mathit{RecPos}(C_i)| \rightarrow \llbracket e_i \rrbracket_{\mu} \Sigma \ \Gamma] \end{aligned} \tag{7.6}$$

However, the interpretation given with these equations (as done in [83]) is, in some cases, less precise than the one given in terms of sequences, as the following example shows:

**Example 7.24.** Assume the following expression in the context of a function depending on a parameter  $b$ :

$$\begin{aligned} e \equiv \mathbf{let} \ x_1 = \mathbf{case} \ b \ \mathbf{of} \\ \quad \mathit{True} \rightarrow be_1 \\ \quad \mathit{False} \rightarrow be_2 \\ \mathbf{in} \ be_3 \end{aligned}$$

and let  $(\Delta_i, \mu_i)$  be the heap costs inferred for the basic expression  $be_i$  ( $i \in \{1, 2, 3\}$ ). If we split  $e$  into sequences we get:

$$\mathit{seqs} \ e = \{[be_1, be_3], [be_2, be_3]\}$$

With the interpretation applied to sequences we get:

$$\mu = \sqcup \left\{ \begin{array}{l} \sqcup \{ \mu_1, |\Delta_1| + \mu_3 \} \\ \sqcup \{ \mu_2, |\Delta_2| + \mu_3 \} \end{array} \right\}$$

whereas with the equations (7.6) we would obtain:

$$\mu' = \sqcup\{\sqcup\{\mu_1, \mu_2\}, |\sqcup\{\Delta_1, \Delta_2\}| + \mu_3\}$$

Then  $\mu' \sqsupseteq \mu$  if and only if  $|\sqcup\{\Delta_1, \Delta_2\}| \sqsupseteq |\sqcup\{|\Delta_1|, |\Delta_2|\}|$ . Assume  $R_f = \{\rho_1, \rho_2\}$  and the following values of  $\Delta_1$  and  $\Delta_2$ :

$$\Delta_1 = \lambda b. \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 2 \end{bmatrix} \quad \Delta_2 = \lambda b. \begin{bmatrix} \rho_1 & \mapsto & 2 \\ \rho_2 & \mapsto & 1 \end{bmatrix}$$

We would get:

$$|\sqcup\{\Delta_1, \Delta_2\}| = \left| \lambda b. \begin{bmatrix} \rho_1 & \mapsto & 2 \\ \rho_2 & \mapsto & 2 \end{bmatrix} \right| = \lambda b.4, \quad \text{but} \quad |\sqcup\{|\Delta_1|, |\Delta_2|\}| = \sqcup\{\lambda b.3, \lambda b.3\} = \lambda b.3$$

So, in this particular case,  $\mu'$  is strictly less precise than  $\mu$ . □

## 7.4 Correctness of the abstract interpretation

In this section we aim to prove the following fact:

*The tuple  $(\Delta, \mu, \sigma)$  resulting from the abstract interpretation of  $e$  is an upper bound to the actual resource vector  $(\delta, m, s)$  resulting from the execution of  $e$ .*

The meaning of  $(\Delta, \mu, \sigma)$  being an upper bound to  $(\delta, m, s)$  was already discussed in Section 7.2, in which we defined a  $\succeq$  relation. However, this relation is parametric with respect some sizes of the input arguments. Hence, in order to prove the claim above we cannot consider  $e$  to be an expression in its own, but contained within a context function definition (7.1) with its set of input parameters. In the same way, we consider that the execution of  $e$ , represented as a  $\Downarrow$ -judgement, belongs to the context derivation specified in (7.2). So, our sizes  $\bar{s}_i^n$  occurring in the  $\succeq$  relations are obtained from the values of the *parameters*  $\bar{x}_i$  in the value environment  $E_0$  and heap  $h_0$  of the context judgement. In other words, for every  $i \in \{1..n\}$ ,  $s_i = \text{size}(h_0, E_0)$ .

We shall assume the admissibility condition (Definition 2.14 on page 36) on the context derivation:

$$\forall j \in \{1..m\}. E_0(r_j) < k_0 \quad \text{and} \quad E(\text{self}) = k_0$$

By Proposition 2.15 we know that this condition is propagated through the subderivations contained within it. This admissibility property is also propagated to the consistent region instantiation defined by  $E_0$  and the typing environment  $\Gamma_0$  of  $e_f$ , since  $\Gamma_0(r_j) = \rho_j$  for all  $j \in \{1..m\}$  and  $\Gamma_0(\text{self}) = \rho_{\text{self}}$ , hence

$$\forall j \in \{1..m\}. \eta(\rho_j) < k_0 \quad \text{and} \quad \eta(\rho_{\text{self}}) = k_0$$

The result of the abstract interpretation depends on some elements that are given as parameters to it, namely:

- The function signatures contained within the signature environment  $\Sigma$  passed to  $\llbracket e \rrbracket$  as a parameter. If we want  $\llbracket e \rrbracket$  to return correct results we have to ensure that the function signatures in  $\Sigma$  are correct approximations to the costs of their respective functions.
- The size functions  $|\cdot|$  used in copy expressions and function applications. Although we do not tackle the problem of size inference, it is necessary to pose a condition of *correct size analysis* in

order to prove that the results of  $\llbracket e \rrbracket$  are correct. In other words, we have to ensure that the sizes given by the analysis are exact or upper approximations of the actual runtime sizes of its corresponding DSs.

The first problem has been already addressed in Definition 7.11. With respect to the second one, we formulate the following notion of a correct size analysis.

**Definition 7.25** (Correct size analysis). Let  $f \bar{x}_i @ \bar{r}_j = e_f$  be the context function. The size analysis  $|\cdot|$  is correct if, given any initial environment  $E_0$  such that the judgement (7.2) is derivable, then for all subexpressions  $e$  of  $e_f$  such that the judgement:

$$E \vdash h, k_0, td, e \Downarrow h', k_0, v, (\delta, m, s)$$

belongs to the derivation of (7.2) it holds that

$$\forall x \in \text{dom } E : |x|_{\bar{s}_i^n} \geq \text{size}(h, E(x)) \quad \text{where } s_i = \text{size}(h_0, E_0(x_i)) \text{ for each } i \in \{1..n\}$$

In other words, a size function  $|x|$  is correct if it is always greater than the size of the data structure pointed to by  $x$  at runtime.

The correctness proof of the abstract interpretation relies on the fact that both the signature environment and the size analysis are correct.

**Theorem 7.26** (Correctness of the abstract interpretation). Let  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  a context function,  $\Gamma$  the inferred global type environment for  $e_f$ ,  $\Sigma$  containing correct signatures for all the functions called from  $e_f$ , an initial environment  $E_0$  and a heap  $h_0$  such that the judgement (7.2) is derivable. For each subexpression  $e$  of  $e_f$  and  $E, td, \Delta, \mu, \sigma, h, h', v, t, \delta, m, s, S$  such that:

1.  $S = \text{seqs } e$ .
2. For every  $\text{seq} \in S$ , every occurrence of  $|x|$  in the evaluation of  $\llbracket \text{seq} \rrbracket_{\Delta} \Sigma \Gamma$  and  $\llbracket \text{seq} \rrbracket_{\mu} \Sigma \Gamma$  has been inferred with a correct size analysis.
3.  $E \vdash h, k_0, td, e \Downarrow h', k_0, v, (\delta, m, s)$  belongs to the derivation of (7.2).

then:

1.  $\llbracket e \rrbracket_{\sigma} \Sigma td \succeq_{\bar{s}_i^n} s$ , and
2. There exists some  $\text{seq} \in S$  such that  $\llbracket \text{seq} \rrbracket_{\Delta} \Sigma \Gamma \succeq_{\bar{s}_i^n, k_0, \eta} \delta$  and  $\llbracket \text{seq} \rrbracket_{\mu} \Sigma \Gamma \succeq_{\bar{s}_i^n} m$ .

where  $\forall i \in \{1..n\}. s_i = \text{size}(h, E_0(x_i))$ , and  $\eta$  is the consistent region instantiation determined by  $E$  and  $\Gamma$ .

*Proof.* By induction on the structure of  $e$ . In the following we will leave out the  $\bar{s}_i^n$  and  $k_0$  subscripts in the  $\succeq$  relations for a better readability. We distinguish cases:

- **Cases**  $e \equiv c, e \equiv x$

We get a single sequence  $[e]$ , for which  $\llbracket [e] \rrbracket_{\Delta} = \lambda \bar{x}_i^n. \lambda \rho. 0$ ,  $\llbracket [e] \rrbracket_{\mu} = \lambda \bar{x}_i^n. 0$ . Besides this, we get  $\llbracket [e] \rrbracket_{\sigma} = \lambda \bar{x}_i^n. 1$ . With respect to the execution, we obtain  $(\delta, m, s) = ([ ]_{k_0}, 0, 1)$ . Thus we get:

1.  $\llbracket [e] \rrbracket_{\sigma} \succeq s$ , since  $\llbracket [e] \rrbracket_{\sigma} \bar{s}_i^n = 1 = s$ .

2.  $\llbracket [e] \rrbracket_{\Delta} \succeq \delta$  and  $\llbracket [e] \rrbracket_{\mu} \succeq m$ , since for every  $j \in \{0..k_0\}$

$$\sum_{\eta(\rho)=j} \llbracket [e] \rrbracket_{\Delta} \bar{s}_i^n \rho = 0 = \delta(j)$$

$$\llbracket [e] \rrbracket_{\mu} \bar{s}_i^n = 0 = m$$

- **Case**  $e \equiv a_1 \oplus a_2$

It is similar to the previous case. The only difference is that  $s = 2$  instead of 1, and  $\llbracket [e] \rrbracket_{\sigma} = \lambda \bar{x}_i^n . 2$  instead of  $\lambda \bar{x}_i^n . 1$ .

- **Case**  $e \equiv x @ r$

Assume that  $\Gamma(r) = \rho$  and  $E(r) = j$ . If we denote  $m = \text{size}(h, E(x))$ , the resulting resource vector is  $([j \mapsto m]_k, m, 2)$ , whereas the result of the analysis is  $([\rho \mapsto |x|], |x|, 2)$ . Then:

1.  $\llbracket [e] \rrbracket_{\sigma} \succeq s$  holds since  $\sigma \bar{s}_i^n = 2 = s$ .

2.  $\llbracket [e] \rrbracket_{\Delta} \succeq \delta$ . Let  $i \in \{0..k_0\}$ . If  $i = j$  then  $\eta(\rho) = (E \circ \Gamma^{-1})(\rho) = i$ . Hence:

$$\begin{aligned} \sum_{\eta(\rho')=i} \Delta \bar{s}_i^n \rho' &\geq \Delta \bar{s}_i^n \rho \\ &= |x| \bar{s}_i^n \\ &\geq \text{size}(h, E(x)) \quad \{\text{by Definition 7.25}\} \\ &= \delta(i) \end{aligned}$$

If  $i \neq j$  we get:

$$\sum_{\eta(\rho')=i} \Delta \bar{s}_i^n \rho' \geq 0 = \delta(i)$$

The proof of  $\llbracket [e] \rrbracket_{\mu} \succeq m$  is similar:

$$\llbracket [e] \rrbracket_{\mu} \bar{s}_i^n = |x| \bar{s}_i^n \geq \text{size}(h, E(x)) = m$$

- **Case**  $e \equiv C \bar{a}_i @ r$

Assume  $\Gamma(r) = \rho$  and  $E(r) = j$ . Then  $(\delta, m, s) = ([j \mapsto 1]_k, 1, 1)$  and  $\llbracket [e] \rrbracket_{\Sigma \Gamma td} = ([\rho \mapsto 1]_f, 1, 1)$ . Again, we have a single sequence  $\text{seq} = [e]$ , so we prove:

1.  $\llbracket [e] \rrbracket_{\sigma} \succeq s$ . It follows from the fact that  $\sigma \bar{s}_i^n = 1 = s$ .

2.  $\llbracket [e] \rrbracket_{\Delta} \succeq \delta$ . Let  $i \in \{0..k_0\}$ , if  $i \neq j$  we get:

$$\sum_{\eta(\rho')=i} \Delta \bar{s}_i^n \rho' \geq 0 = \delta(i)$$

and with regard to  $j$ , we know that  $\eta(\rho) = (E \circ \Gamma^{-1})(\rho) = j$ , so

$$\sum_{\eta(\rho')=j} \Delta \bar{s}_i^n \rho' \geq \Delta \bar{s}_i^n \rho = 1 = \delta(j)$$

The proof of  $\llbracket [e] \rrbracket_{\mu} \succeq m$  follows from the definitions of  $\mu$  and  $m$ :

$$\mu \bar{s}_i^n = 1 = m$$

- **Case**  $e \equiv g \bar{a}_i^l @ \bar{r}_j^{lq}$

We assume that  $\Sigma g \equiv g \bar{y}_i^l @ \bar{r}_j^{lq} = e_g$  and, by using the corresponding rule:

$$E_g \vdash h, k_0 + 1, l + q, e_g \Downarrow h', k_0 + 1, v, (\delta_g, m_g, s_g)$$

$$\text{where } E_g = \left[ \overline{y_i \mapsto E(a_i)^l}, \overline{r_j^{lq} \mapsto E(r_j^{lq})^q}, \text{self} \mapsto k_0 + 1 \right]$$

The correctness of the signature  $(\Delta_g, \mu_g, \sigma_g)$  follows by assumption. Moreover,  $g$  is well-typed and if  $\Gamma(g) = \forall \bar{\alpha} \bar{\rho}. \bar{t}_i^l \rightarrow \bar{\rho}_j^{lq} \rightarrow t$ , the global type  $\Gamma_g$  being inferred for  $e_g$  contains the bindings  $[\bar{r}_j^{lq} : \bar{\rho}_j^{lq}]$ . On the other hand, if  $s_{i,g}$  denote the size of the  $i$ -th actual argument before evaluating the function's body (i.e.  $\forall i \in \{1 \dots l\} : s_{i,g} = \text{size}(h, E_g(y_i))$ ) then, by our definition of correct signature:

$$\Delta_g \succeq_{\bar{s}_{i,g}^{-l}, k_0, \eta'} \delta_g|_{k_0} \quad \mu_g \succeq_{\bar{s}_{i,g}^{-l}} m_g \quad \sigma_g \succeq_{\bar{s}_{i,g}^{-l}} s_g$$

where  $\eta' = E_g \cdot \Gamma_g^{-1}$ . This implies, in particular, that  $\Delta_g \bar{s}_{i,g}^{-l} \neq \perp$ ,  $\mu_g \bar{s}_{i,g}^{-l} \neq \perp$  and  $\sigma_g \bar{s}_{i,g}^{-l} \neq \perp$ . By Definition 7.25 we get, for each  $i \in \{1 \dots l\}$

$$|a_i| \bar{s}_i^{-n} \geq \text{size}(h, E(a_i)) = \text{size}(h, E_g(y_i)) = s_{i,g} \quad (7.7)$$

So, by monotonicity of  $\Delta_g, \mu_g$  and  $\sigma_g$  we get  $\Delta_g (\overline{|a_i| \bar{s}_i^{-n}})^l \neq \perp$ ,  $\mu_g (\overline{|a_i| \bar{s}_i^{-n}})^l \neq \perp$  and  $\sigma_g (\overline{|a_i| \bar{s}_i^{-n}})^l \neq \perp$ , from which  $G (\overline{|a_i| \bar{s}_i^{-n}})^l$  holds. Now we prove:

1.  $\llbracket e \rrbracket_\sigma \succeq s$ . Since  $\sigma_g$  is monotonic:

$$\begin{aligned} \llbracket e \rrbracket_\sigma \bar{s}_i^{-n} &= \sqcup \{l + q, \sigma_g (\overline{|a_i| \bar{s}_i^{-n}})^l - td + l + q\} \\ &\geq \sqcup \{l + q, \sigma_g \bar{s}_{j,g}^{-l} - td + l + q\} \\ &\geq \sqcup \{l + q, s_g - td + l + q\} \\ &= s \end{aligned}$$

2.  $\llbracket [e] \rrbracket_\Delta \succeq_{\bar{s}_i^{-n}, k_0, \eta} \delta$ . Let  $i \in \{0 \dots k_0\}$ . Let us define  $\Delta = \theta \downarrow_{\overline{|a_i|}} \Delta_g$ :

$$\sum_{\eta(\rho)=i} \Delta \bar{s}_i^{-n} \rho = \sum_{\eta(\rho)=i} \sum_{\theta(\rho')=\rho} \Delta_g \overline{|a_i| \bar{s}_i^{-n}}^l \rho'$$

where  $\theta = \text{unify}(\Gamma, g, \bar{r}_j^{lq})$ .

Because of monotonicity of  $\Delta_g$ :

$$\sum_{\eta(\rho)=i} \Delta \bar{s}_i^{-n} \rho \geq \sum_{\eta(\rho)=i} \sum_{\theta(\rho')=\rho} \Delta_g \bar{s}_{i,g}^{-l} \rho' = \sum_{(\eta \circ \theta)(\rho')=i} \Delta_g \bar{s}_{i,g}^{-l} \rho'$$

By definition of  $\Delta_g \succeq_{\bar{s}_{i,g}^{-l}, k_0, (\eta \circ \theta)} \delta_g|_{k_0}$  and because of the fact that  $i \neq k_0 + 1$ , we can get the desired result:

$$\sum_{\eta(\rho)=i} \Delta \bar{s}_i^{-n} \rho \geq \delta_g|_{k_0}(i) = \delta(i)$$

provided  $\eta \circ \theta = \eta'$ . However, for each  $j \in \{1..q\}$ :

$$\begin{aligned}
\eta(\theta(\rho_j)) &= E(\Gamma^{-1}(\Gamma(r'_j))) \quad \{\text{by definition of } \eta \text{ and } \theta\} \\
&= E(r'_j) \\
&= E_g(r''_j) \quad \{\text{by definition of } E_g\} \\
&= E_g(\Gamma_g^{-1}(\rho_j)) \quad \{\text{by definition of } \Gamma_g\} \\
&= \eta'(\rho_j) \quad \{\text{by definition of } \eta'\}
\end{aligned}$$

So  $\llbracket [e] \rrbracket_\Delta = \Delta \succeq_{\bar{s}_i^n, k_0, \eta} \delta$ . Finally, we prove that  $\llbracket [e] \rrbracket_\mu \succeq_{\bar{s}_i^n} m$ :

$$\begin{aligned}
\mu \bar{s}_i^n &= \mu_g \overline{|a_i| \bar{s}_i^{n_l}} \\
&\geq \mu_g \bar{s}_{i,g}^l \quad \{\text{because of (7.7) and monotonicity of } \mu_g\} \\
&\geq m_g \quad \{\text{since } \mu_g \succeq_{\bar{s}_{i,g}^l} m_g\} \\
&= m
\end{aligned}$$

• **Case  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$**

Let us denote by  $S_1$  and  $S_2$  the results of  $\text{seqs } e_1$  and  $\text{seqs } e_2$  respectively. Assume that the execution of  $e_1$  returns  $(\delta_1, m_1, s_1)$  as a resource vector, and the execution of  $e_2$  returns  $(\delta_2, m_2, s_2)$ . By induction hypothesis we get the following facts:

$$\llbracket [e_1] \rrbracket_\sigma \Sigma 0 \succeq_{\bar{s}_i^n} s_1 \quad \llbracket [e_2] \rrbracket_\sigma \Sigma (td + 1) \succeq_{\bar{s}_i^n} s_2 \quad (7.8)$$

$$\exists \text{seq}_1 \in S_1. \llbracket [\text{seq}_1] \rrbracket_\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta_1 \wedge \llbracket [\text{seq}_1] \rrbracket_\mu \succeq_{\bar{s}_i^n} m_1 \quad (7.9)$$

$$\exists \text{seq}_2 \in S_2. \llbracket [\text{seq}_2] \rrbracket_\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta_2 \wedge \llbracket [\text{seq}_2] \rrbracket_\mu \succeq_{\bar{s}_i^n} m_2 \quad (7.10)$$

Now we prove:

1.  $\llbracket [e] \rrbracket_\sigma \succeq_{\bar{s}_i^n} s$ . This follows from (7.8) because:

$$\begin{aligned}
\llbracket [e] \rrbracket_\sigma \Sigma td \bar{s}_i^n &= \sqcup \{2 + (\llbracket [e_1] \rrbracket_\sigma \Sigma 0) \bar{s}_i^n, 1 + (\llbracket [e_2] \rrbracket_\sigma \Sigma (td + 1)) \bar{s}_i^n\} \\
&\geq \sqcup \{2 + s_1, 1 + s_2\} \\
&= s
\end{aligned}$$

2.  $\llbracket [\text{seq}_1 ++ \text{seq}_2] \rrbracket_\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta$  and  $\llbracket [\text{seq}_1 ++ \text{seq}_2] \rrbracket_\mu \succeq_{\bar{s}_i^n} m$ . We get:

$$\llbracket [\text{seq}_1 ++ \text{seq}_2] \rrbracket_\Delta = \llbracket [\text{seq}_1] \rrbracket_\Delta + \llbracket [\text{seq}_2] \rrbracket_\Delta \succeq_{\bar{s}_i^n, k, \eta} \delta_1 + \delta_2 = \delta$$

$$\llbracket [\text{seq}_1 ++ \text{seq}_2] \rrbracket_\mu = \sqcup \{ \llbracket [\text{seq}_1] \rrbracket_\mu, | \llbracket [\text{seq}_1] \rrbracket_\Delta | + \llbracket [\text{seq}_2] \rrbracket_\mu \} \succeq_{\bar{s}_i^n} \sqcup \{ m_1, |\delta_1| + m_2 \} = m$$

In both cases the first step follows from Lemma 7.20, and the second one follows from (7.9), (7.10), and Lemmas 7.9 and 7.10.

• **Case  $e \equiv \text{case } x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i}} \rightarrow e_i$**

Assume the  $r$ -th branch being executed ( $r \in \{1..n\}$ ). If  $(\delta_r, m_r, s_r)$  is the resource vector associated to this branch, we get  $\delta = \delta_r$ ,  $m = m_r$  and  $s = n_r + s_r$ . By induction hypothesis we obtain:

$$\llbracket [e_r] \rrbracket_\sigma \Sigma (td + n_r) \succeq_{\bar{s}_i^n} s_r \quad (7.11)$$

$$\exists seq \in seqs e_r. \llbracket seq \rrbracket_{\Delta} \succeq_{\bar{s}_i^n, k, \eta} \delta_r = \delta \wedge \llbracket seq \rrbracket_{\mu} \succeq_{\bar{s}_i^n} m_r = m \quad (7.12)$$

For each  $i \in \{1..n\}$  let us denote by  $G_i$  the guard  $\lambda \bar{x}_i. |x| \bar{x}_i \geq 1 + RecPos(C_i)$ . Since the initial configuration is closed, then  $size(h, E(x)) \geq 1 + RecPos(C_r)$ . By the correctness property of the size analysis,  $|x| \bar{s}_i^n \geq size(h, E(x)) \geq 1 + RecPos(C_r)$ , so  $G_r(\bar{s}_i^n)$  holds. We prove below the conclusions of the Theorem:

1.  $\llbracket e \rrbracket_{\sigma} \succeq_{\bar{s}_i^n} n_r + s_r$ . We get:

$$\begin{aligned} \llbracket e \rrbracket_{\sigma} \Sigma td \bar{s}_i^n &= \bigsqcup_{i=1}^n [G_i \rightarrow n_i + \llbracket e_i \rrbracket_{\sigma} \Sigma \Gamma (td + n_i)] \bar{s}_i^n \\ &\geq [G_r \rightarrow n_r + \llbracket e_r \rrbracket_{\sigma} \Sigma \Gamma (td + n_r)] \bar{s}_i^n \\ &\geq n_r + s_r \end{aligned}$$

The last step being justified by (7.11) and the fact that  $G_r(\bar{s}_i^n)$  holds.

2. The existence of a sequence  $seq \in seqs e$  satisfying the conclusions of the theorem follows trivially from (7.12), since  $seqs e = \bigcup_{i=1}^n (seqs e_i \wedge G_i)$  and  $G_r(\bar{s}_i^n)$  holds.

□

Notice that we have not established that  $\llbracket e \rrbracket_{\Delta}$  and  $\llbracket e \rrbracket_{\mu}$  are upper bounds to the actual  $\delta$  and  $m$  components, respectively. The previous theorem states a stronger property: there exists a sequence in  $seqs e$  whose  $\llbracket \cdot \rrbracket_{\Delta}$  and  $\llbracket \cdot \rrbracket_{\mu}$ -interpretations are upper bounds to the  $\delta$  and  $m$ , respectively. However, and since  $\llbracket e \rrbracket_{\Delta}$  and  $\llbracket e \rrbracket_{\mu}$  are defined as  $\llbracket seqs e \rrbracket_{\Delta} = \bigsqcup_{seq \in seqs e} \llbracket seq \rrbracket_{\Delta}$  and  $\llbracket seqs e \rrbracket_{\mu} = \bigsqcup_{seq \in seqs e} \llbracket seq \rrbracket_{\mu}$ , it follows trivially that  $\llbracket e \rrbracket_{\Delta}$  and  $\llbracket e \rrbracket_{\mu}$  are correct approximations to  $\delta$  and  $m$ .

From this correctness result we can devise a way for inferring upper bounds to the heap and stack consumption of a non-recursive function, assuming that the functions called from it have already been inferred. We just have to apply the abstract interpretation to the body of the function definition.

**Example 7.27.** Let us recall the *combNumber* function, which, given two integers  $n$  and  $m$ , returns the result of  $\binom{n}{m}$  (see Example 4.20 on page 153). Assuming our signature environment  $\Sigma$  has the following information regarding the stack costs of (!) and *pascal*:

$$\begin{aligned} \sigma_{(!)} &= \lambda xs m. [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 6] \\ \sigma_{pascal} &= \lambda n. 15n + 4 \end{aligned}$$

If  $e_{combNumber}$  denotes the body of the *combNumber* function, we get:

$$\llbracket e_{combNumber} \rrbracket_{\sigma} \Sigma 2 = \lambda n m. [n \geq 0 \rightarrow 15n + 14] \sqcup [n \geq 1 \rightarrow 15n + 17]$$

which is a correct bound to the stack costs of *combNumber*, provided  $\sigma_{(!)}$  and  $\sigma_{pascal}$  are correct upper bounds to the stack consumption of their respective functions. □

The inference of recursive function definitions is far more involved, and its study is deferred to the following section.

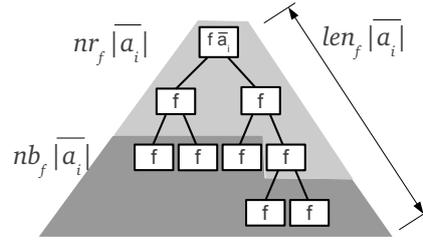


Figure 7.3: Representation of the activation tree of a given function call to  $f$ . The result of  $nr_f \overline{|a_i|}$  and  $nb_f \overline{|a_i|}$  are the number of internal nodes and leaves, respectively. The value of  $len_f \overline{|a_i|}$  represents the height of the activation tree.

## 7.5 Memory consumption of recursive function definitions

The obvious question that arises when applying the abstract interpretation to a recursive function is which signature associated to  $f$  must be stored into the signature environment  $\Sigma$ . The result of the abstract interpretation will be correct if this initial signature is correct, but this signature is what we aim to infer. We have to find a correct upper bound to the memory consumptions of a function definition by other means different from the  $\llbracket \cdot \rrbracket$ -interpretations. Once these upper bounds are computed, it still makes sense to apply the abstract interpretation under a signature environment  $\Sigma$  containing those bounds. The new results are also correct, by Theorem 7.26, but they might be more precise than the initial ones, as we will see later.

The rest of this section is devoted to the computation of these *initial approximations*, which will be called  $\Delta_0$ ,  $\mu_0$  and  $\sigma_0$ . In order to compute them, we need some information regarding the number of recursive calls produced during the evaluation of a call to  $f$ . This information will be given as a function of the input sizes  $\overline{x_i}$ . We represent these recursive calls by means of activation trees (or *call trees*), as in Figure 7.3. Before computing an upper bound to the memory costs, we assume that the following information is available as elements of  $\mathbb{F}$ :

$nr_f$  Upper bound to the number of calls to  $f$  which invoke  $f$  again. This number corresponds to the internal nodes of  $f$ 's call tree.

$nb_f$  Upper bound to the number of basic calls to  $f$  that do not invoke  $f$  again. It corresponds to the leaf nodes of  $f$ 's call tree.

$len_f$  Upper bound to the maximum length of  $f$ 's call chains. It corresponds to the height of  $f$ 's call tree.

In general these functions are not independent of each other. For instance, with linear recursion we get  $nr_f = len_f - 1$  and  $nb_f = 1$ . However, we shall not assume a fixed relation between them. The computation of these three functions is closely related to the problem of termination and the computation of *ranking functions*, and we can use the techniques described in [73] for computing them. Another possibility is to give a definition of these components as a recurrence relation and obtain a closed form by using recurrence solving tools, such as PUBS [7, 8], possibly in combination with polynomial interpolation-based techniques [87].

**Example 7.28.** Assume a call to *append*  $xs\ ys$  where  $xs = [x_1, \dots, x_n]$ . We get the following call sequence:

$$\begin{array}{ll} & \text{append } [x_1, \dots, x_n] \quad \text{1st call} \\ \rightarrow & \text{append } [x_2, \dots, x_n] \quad \text{2nd call} \\ & \vdots \\ \rightarrow & \text{append } [x_n] \quad n\text{-th call} \\ \rightarrow & \text{append } [] \quad (n+1)\text{-th call} \end{array}$$

So we obtain  $n + 1$  calls to *append*,  $n$  of which are recursive. Since the size of a list is its number of elements plus one, we get the following functions:

$$\begin{aligned} nb_{\text{append}} &= \lambda xs\ ys.1 \\ nr_{\text{append}} &= \lambda xs\ ys.xs - 1 \\ len_{\text{append}} &= \lambda xs\ ys.xs \end{aligned}$$

□

In the following subsections we shall present three algorithms for computing our initial  $\Delta_0$ ,  $\mu_0$  and  $\sigma_0$  (sections 7.5.3, 7.5.4 and 7.5.5, respectively). For each algorithm, we have to prove the following facts:

**[CR]** The obtained bounds  $\Delta_0$ ,  $\mu_0$  and  $\sigma_0$  are correct bounds to the actual memory needs of the program.

**[RD]** The results of  $\llbracket e_f \rrbracket_{\Delta}$ ,  $\llbracket e_f \rrbracket_{\mu}$  and  $\llbracket e_f \rrbracket_{\sigma}$  (being  $e_f$  the body of the function definition) under a signature environment mapping  $f$  to  $(\Delta_0, \mu_0, \sigma_0)$  are equal or more precise than the initial approximations  $(\Delta_0, \mu_0, \sigma_0)$ .

The latter fact is closely related to the concept of reductive functions over complete lattices. In the next subsection we shall briefly describe all these concepts, including *Tarki's Fixed Point Theorem*, which is of crucial importance for proving that, under some conditions on the given  $nb_f$ ,  $nr_f$  and  $len_f$ , **[RD]** implies **[CR]**. In case these conditions do not hold, **[RD]** may not hold, but **[CR]** still holds. Subsection 7.5.6 addresses this situation.

### 7.5.1 Preliminaries on fixed points in complete lattices

Let  $(L, \sqsubseteq)$  be a complete lattice and  $f$  a monotonic function on  $L$ . Given an element  $x \in L$ :

- We say that  $x$  is a *fixed point* of  $f$  iff  $f(x) = x$ .
- We say that  $f$  is *reductive* at  $x$  iff  $f(x) \sqsubseteq x$ .
- We say that  $f$  is *extensive* at  $x$  iff  $f(x) \sqsupseteq x$ .

We denote by  $Fix(f)$  the set of fixed points of  $f$ :

$$Fix(f) = \{x \in L \mid f(x) = x\}$$

Similarly, we denote by  $Red(f)$  (resp.  $Ext(f)$ ) the set of points upon which  $f$  is reductive (resp. extensive).

$$\begin{aligned} Red(f) &= \{x \in L \mid f(x) \sqsubseteq x\} \\ Ext(f) &= \{x \in L \mid f(x) \sqsupseteq x\} \end{aligned}$$

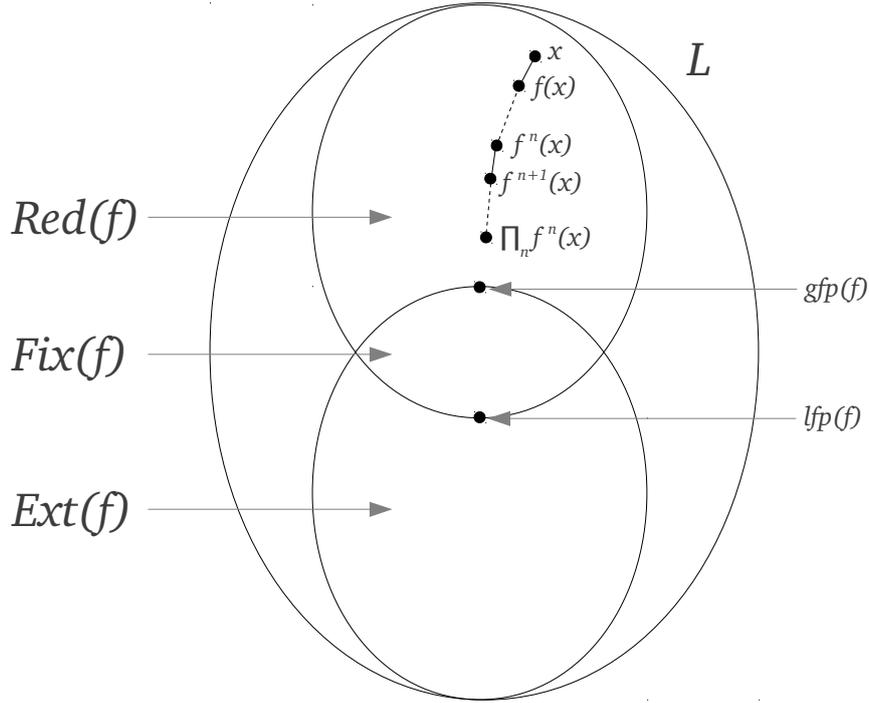


Figure 7.4: Representation of the points upon which a given function is reductive and extensive. The intersection of  $Red(f)$  and  $Ext(f)$  is the set of fixed points of  $f$ . Given an element  $x \in Red(f)$ ,  $f^n(x)$  is always above the least fixed point.

Given the fact that  $L$  is a complete lattice, the least upper bound and the greatest lower bounds of  $Fix(f)$  are both defined and respectively denoted by  $lfp(f)$  and  $gfp(f)$ . Tarski's fixed fixed point theorem [111] establishes the relation between fixed points and the reductivity/extensivity properties.

**Theorem 7.29** (Tarski 1955, taken from [92]). *Let  $(L, \sqsubseteq)$  a complete lattice and  $f : L \rightarrow L$  a monotone function. Then  $lfp(f) = \sqcap Red(f)$  and  $gfp(f) = \sqcup Ext(f)$ .*

*Proof.* (see [92, Section A.4]). □

Figure 7.4 depicts the layout of the reductive and extensive elements of  $L$  w.r.t. a function  $f$ . If  $x$  is an element belonging to  $Red(f)$ , then  $f(x) \sqsubseteq x$ . By monotonicity of  $f$ , we get  $f(f(x)) \sqsubseteq f(x)$ . By repeating this process we get the following chain,

$$x \sqsupseteq f(x) \sqsupseteq f^2(x) \sqsupseteq \dots \sqsupseteq f^n(x) \sqsupseteq \dots$$

whose elements, by Tarski's theorem, are located above the least fixed point.

Now let us apply these concepts to our particular abstract interpretation. From Propositions 7.3 and 7.5 we know that  $(\mathbb{F}, \sqsubseteq)$  and  $(\mathbb{ID}, \sqsubseteq)$  are complete lattices. Given the context function definition  $f$ , and some fixed  $\Sigma, \Gamma$  and  $td$ , the iteration of the abstract interpretation can be understood as a function on abstract heaps  $\mathbb{ID} \rightarrow \mathbb{ID}$  (resp. on cost functions  $\mathbb{F} \rightarrow \mathbb{F}$ ) which, given an input  $\Delta$  (resp.  $\mu, \sigma$ ), inserts it into the signature environment  $\Sigma$ , and apply the  $\llbracket e_f \rrbracket_{\Delta}$  interpretation (resp.  $\llbracket e_f \rrbracket_{\mu}$  and  $\llbracket e_f \rrbracket_{\sigma}$ ) to the body of  $f$ .

**Definition 7.30.** Assume a function definition  $f \bar{x}_i @ \bar{r}_j = e_f$ , and some fixed  $\Sigma, \Gamma, \Delta$  and  $td$ , such that

$f \notin \text{dom } \Sigma$ . The operators  $\mathcal{I}_f : \mathbb{D} \rightarrow \mathbb{D}$ ,  $\mathcal{M}_{\Delta,f} : \mathbb{F} \rightarrow \mathbb{F}$  and  $\mathcal{S}_f : \mathbb{F} \rightarrow \mathbb{F}$  are defined as follows:

$$\begin{aligned}\mathcal{I}_f(\Delta) &= \left[ \llbracket e_f \rrbracket_{\Delta} (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \text{td} \right] \\ \mathcal{M}_{\Delta,f}(\mu) &= \llbracket e_f \rrbracket_{\mu} (\Sigma \uplus [f \mapsto (\Delta, \mu, 0)]) \Gamma \text{td} \\ \mathcal{S}_f(\sigma) &= \llbracket e_f \rrbracket_{\sigma} (\Sigma \uplus [f \mapsto ([ ]_f, 0, \sigma)]) \Gamma \text{td}\end{aligned}$$

Notice that, strictly speaking, the  $\mathcal{I}$ ,  $\mathcal{M}$  and  $\mathcal{S}$  operators are also parametric on the given  $\Sigma$ ,  $\Gamma$  and  $\text{td}$ , but we assume all these elements fixed. Now we prove their monotonicity.

**Proposition 7.31.** *The operators  $\mathcal{I}_f$ ,  $\mathcal{M}_{\Delta,f}$  and  $\mathcal{S}_f$  are monotonic on their input arguments.*

*Proof.* From the definition of the abstract interpretation it follows that the result of  $\mathcal{I}$ ,  $\mathcal{M}$  and  $\mathcal{S}$  is computed by composing size functions with the  $\sqcup, \sqsubseteq, +$  operators, and possibly with the substraction of a constant value ( $\text{td}$ ). All these operations are monotone w.r.t. their inputs, and so is the composition of all of them.  $\square$

For each of these operators, we can define its set of fixed points, reductive elements and extensive elements in the same way as above.

**Example 7.32.** Consider the *append* function of Example 7.17. Let  $\Delta$  be an abstract heap such that  $\text{dom } \Delta = \{(xs, ys) \mid xs \geq 1\}$ . If we assume that  $\Gamma = [r_2 : \rho_2]$ , our operator  $\mathcal{I}_{\text{append}}$  is defined as follows:

$$\mathcal{I}_{\text{append}}(\Delta) \text{ } xs \text{ } ys \text{ } \rho_2 = [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1 + \Delta(xs - 1, ys)]$$

Let us consider the abstract heaps  $\Delta_1$ ,  $\Delta_2$  and  $\Delta_3$  defined as follows:

$$\begin{aligned}\Delta_1 &\stackrel{\text{def}}{=} \lambda xs \text{ } ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 0]] \\ \Delta_2 &\stackrel{\text{def}}{=} \lambda xs \text{ } ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto [xs - 1]]] \\ \Delta_3 &\stackrel{\text{def}}{=} \lambda xs \text{ } ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 2xs]]\end{aligned}$$

If we apply the  $\mathcal{I}_{\text{append}}$  iterator to each of these, we obtain the following results:

$$\begin{aligned}\mathcal{I}_{\text{append}}(\Delta_1) \text{ } xs \text{ } ys \text{ } \rho_2 &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1 + 0] \\ &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1]\end{aligned}$$

$$\begin{aligned}\mathcal{I}_{\text{append}}(\Delta_2) \text{ } xs \text{ } ys \text{ } \rho_2 &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1 + [xs - 2]] \\ &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow [xs - 1]] \\ &= [xs \geq 1 \rightarrow [xs - 1]]\end{aligned}$$

$$\begin{aligned}\mathcal{I}_{\text{append}}(\Delta_3) \text{ } xs \text{ } ys \text{ } \rho_2 &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1 + 2(xs - 1)] \\ &= [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 2xs - 1]\end{aligned}$$

Therefore,  $\Delta_1 \in \text{Ext}(\mathcal{I}_{\text{append}})$ ,  $\Delta_2 \in \text{Fix}(\mathcal{I}_{\text{append}})$ , and  $\Delta_3 \in \text{Red}(\mathcal{I}_{\text{append}})$ .  $\square$

An important result is that, if we take a fixed point of each operator, the resulting signature is correct for the given function definition. The precise statement of this fact is rather involved and beyond the scope of this work. For its purposes we consider a simplified version, and refer to Javier de Dios' PhD thesis [35] for more details.

**Theorem 7.33.** *Let  $(\Delta, \mu, \sigma)$  be a signature for  $f$ . If  $\Delta \in \text{Fix}(\mathcal{I}_f)$ ,  $\mu \in \text{Fix}(\mathcal{M}_{\Delta, f})$  and  $\sigma \in \text{Fix}(\mathcal{S}_f)$ , the signature  $(\Delta, \mu, \sigma)$  is correct for  $f$ .*

*Proof.* By induction on the maximum number of nested calls to  $f$  in the corresponding  $\Downarrow$ -derivation. The full proof can be found in [35].  $\square$

However, and by Tarski's Theorem (7.29), the reductive elements of each operator are always located above the corresponding least fixed point, so these can also be considered as correct bounds.

**Corollary 7.34.** *Let  $(\Delta, \mu, \sigma)$  be a signature for  $f$ . If  $\Delta \in \text{Red}(\mathcal{I}_f)$ ,  $\mu \in \text{Red}(\mathcal{M}_{\Delta, f})$  and  $\sigma \in \text{Red}(\mathcal{S}_f)$ , the signature  $(\Delta, \mu, \sigma)$  is correct for  $f$ .*

*Proof.* By Theorem 7.33,  $(\text{lfp}(\mathcal{I}_f), \text{lfp}(\mathcal{M}_{\Delta, f}), \text{lfp}(\mathcal{S}_f))$  is a correct upper bound to  $f$ . Let  $\Delta$ ,  $\mu$  and  $\sigma$  defined as above. By Theorem 7.29 we get  $\Delta \sqsupseteq \text{lfp}(\mathcal{I}_f) \in \text{Fix}(\mathcal{I}_f)$ ,  $\mu \sqsupseteq \text{lfp}(\mathcal{M}_{\Delta, f}) \in \text{Fix}(\mathcal{M}_{\Delta, f})$ , and  $\sigma \sqsupseteq \text{lfp}(\mathcal{S}_f) \in \text{Fix}(\mathcal{S}_f)$ . Thus,  $(\Delta, \mu, \sigma)$  is also a correct bound.  $\square$

A consequence of this Corollary is that the [RD] condition shown above is sufficient in order to prove the correctness condition [CR] of the initial bounds.

## 7.5.2 Splitting Core-Safe sequences

When computing initial approximations to the heap and stack consumption of a function definition, it is useful to separate the costs charged by its base and recursive cases. The result of flattening an expression into a sequence of basic expressions (via the *seqs* function) is quite amenable to this separation: a sequence takes part in the base case if it does not contain recursive calls, and it takes part in the recursive one if it does contain them.

**Definition 7.35.** Given the context function  $f$  whose memory consumption is being inferred, a sequence of basic expressions is said to be a *base sequence* if it does not contain any expression of the form  $f \bar{a}_i @ \bar{r}_j$  for some  $\bar{a}_i$  and  $\bar{r}_j$ . Otherwise it is said to be *recursive*.

We define the *splitExp<sub>f</sub>* function, which classifies an arbitrary set of sequences into base and recursive sequences.

**Definition 7.36.** Given a set of sequences  $S$  and a context function  $f$ , the *splitExp<sub>f</sub>* function is defined as follows:

$$\text{splitExp}_f S = (S_b, S_r) \text{ where } \begin{cases} S_b = \{seq \in S \mid seq \text{ is a base sequence}\} \\ S_r = \{seq \in S \mid seq \text{ is a recursive sequence}\} \end{cases}$$

Obviously, if  $\text{splitExp}_f S = (S_b, S_r)$ , then  $S = S_b \uplus S_r$ .

When computing the charges to the working region, it is also useful to separate the charges done until (and including) the last recursive call. Again, we can do this in terms of sequences. It is easy to see that, if  $seq$  is a recursive sequence of the body of the context function  $f$ , we can split it into two sequences  $seq_{bef}$  and  $seq_{aft}$  such that  $seq = seq_{bef} ++ seq_{aft}$ , the last element of  $seq_{bef}$  is of the form  $f \bar{a}_i @ \bar{r}_j$

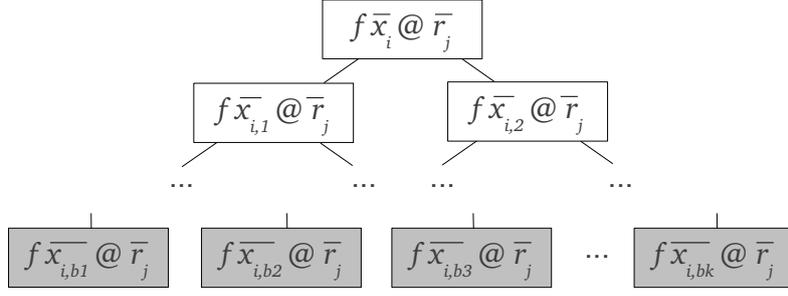


Figure 7.5: Activation tree corresponding to a function call  $f \bar{x}_i @ \bar{r}_j$ . The grey nodes correspond to base calls, whereas the white ones correspond to recursive calls.

whereas  $seq_{aft}$  does not contain expressions of this form, and the guard of  $seq_{bef}$  and  $seq_{aft}$  is the same as that of  $seq$ . The uniqueness of this decomposition allows us to define the function  $splitBA_f$  which, when applied to sequences, returns the pair  $(seq_{bef}, seq_{aft})$  resulting from this decomposition. This definition can be trivially extended to sets of sequences.

**Definition 7.37.** Given a set of sequences  $S$  and a context function  $f$ , the function  $splitBA_f$  is defined as follows:

$$splitBA_f S = \{splitBA_f seq \mid seq \in S\}$$

The  $splitBA_f$  function returns a set of pairs. We can join the first components of all these pairs in order to get a single set of *before* fragments. Similarly, the second components of these pairs can also be joined. This gives place to the definition of  $splitBA_f^*$  returning a pair of sets:

$$splitBA_f^* S = (\{seq_{bef} \mid (seq_{bef}, seq_{aft}) \in splitBA_f S\}, \{seq_{aft} \mid (seq_{bef}, seq_{aft}) \in splitBA_f S\})$$

**Example 7.38.** In our *append* running example (7.17), if  $S = seqs e_{append}$  we obtain  $splitExp_{append} S = (S_b, S_r)$  where:

$$\begin{aligned} S_b &= \{[xs \geq 1 \rightarrow ys]\} \\ S_r &= \{[xs \geq 2 \rightarrow append \ xx \ ys, (x : x_1)@r]\} \end{aligned}$$

by applying  $splitBA$  and  $splitBA^*$  to the resulting  $S_r$  set, we get:

$$\begin{aligned} splitBA_{append} S_r &= \{([xs \geq 2 \rightarrow append \ xx \ ys], [xs \geq 2 \rightarrow (x : x_1)@r])\} \\ splitBA_{append}^* S_r &= (\{[xs \geq 2 \rightarrow append \ xx \ ys]\}, \{[xs \geq 2 \rightarrow (x : x_1)@r]\}) \end{aligned}$$

□

### 7.5.3 Algorithm for computing $\Delta_0$

In order to grasp the intuitive meaning of the algorithm, let us consider the activation tree corresponding to a function call  $f \bar{x}_i @ \bar{r}_j$  (Figure 7.5). We can abstract each node by its charges on memory as follows:

- Base nodes are abstracted by the result of  $\llbracket S_b \rrbracket_{\Delta}$ , where  $S_b$  contains the base sequences of the function's body. We use  $\Delta_b$  to denote this abstract heap.

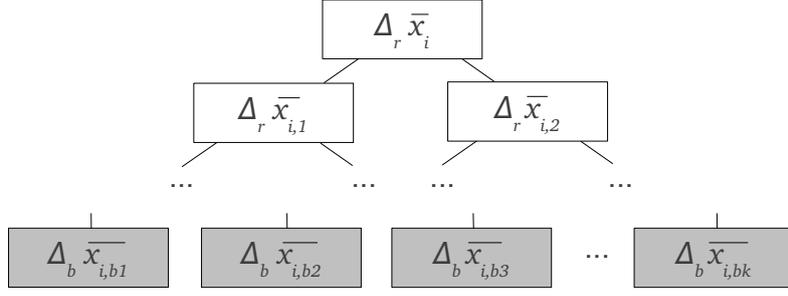


Figure 7.6: Abstraction of the charges done to each call in the activation tree.

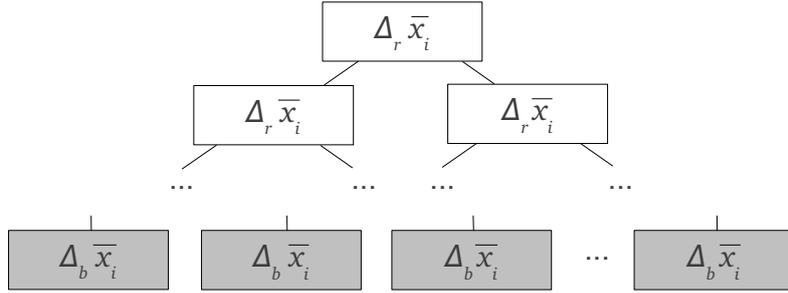


Figure 7.7: The substitution of  $\bar{x}_i$  for the parameters in each recursive call leads to an abstraction that approximates the tree shown in Figure 7.6.

- Recursive nodes are abstracted by the result of  $\llbracket S_r \rrbracket_{\Delta_r}$ , where  $S_r$  contains the recursive sequences of the function's body, and without taking into account the recursive calls occurring in  $S_r$ . We use  $\Delta_r$  to denote this abstract heap.

The result of this abstraction is depicted in Figure 7.6. In the internal nodes of the tree we get  $\Delta_r \bar{x}_{i,j}$ , where  $j$  depends on the particular recursive call. In the leaves we obtain  $\Delta_b \bar{x}_{i,bj}$ , where  $j$  depends on the particular base call. We can further approximate this abstraction by imposing some conditions on the  $\Delta_b$  and  $\Delta_r$ .

**Definition 7.39.** An abstract heap  $\Delta \in \mathbb{ID}$  (resp. a space cost function  $\zeta \in \mathbb{F}$ ) is said to be *parameter-decreasing* with respect to a function definition  $f \bar{x}_i @ \bar{r}_j = e_f$  iff, for every recursive call  $f \bar{a}_i^n @ \bar{r}_j^m$  occurring in  $e_f$ , it holds that

$$\Delta \overline{|a_i| \bar{x}_i} \sqsubseteq \Delta \bar{x}_i$$

or, respectively,

$$\zeta \overline{|a_i| \bar{x}_i} \sqsubseteq \zeta \bar{x}_i$$

Therefore, if we assume that our  $\Delta_r$  and  $\Delta_b$  are parameter-decreasing, we can substitute each  $\Delta_r \bar{x}_{i,j}$  and  $\Delta_b \bar{x}_{i,bj}$  by their counterparts  $\Delta_r \bar{x}_i$  and  $\Delta_b \bar{x}_i$ . The new tree we obtain in this way (Figure 7.7) is an upper approximation to the previous one, but now all the recursive nodes charge the same cost, and so do the base nodes. If  $nb$  and  $nr$  respectively approximate the number of base and recursive nodes, our initial approximation is given by the function  $\lfloor \Delta_b \rfloor * nb + \lfloor \Delta_r \rfloor * nr$ .

The algorithm for computing the initial approximation  $\Delta_0$  is shown in Figure 7.8. The least upper bound with  $\lfloor \Delta_b \rfloor$  handles those input sizes in which  $\lfloor \Delta_r \rfloor$  becomes undefined.

$$\begin{aligned}
\text{computeDelta } (f \bar{x}_i @ \bar{r}_j = e_f) \Sigma \Gamma \text{ nb } nr &= ([\Delta_b] * \text{nb} + [\Delta_r] * nr) \sqcup [\Delta_b] \\
\text{where } S &= \text{seqs } e_f \\
(S_b, S_r) &= \text{splitExp}_f S \\
\Delta_b &= \llbracket S_b \rrbracket_{\Delta} (\Sigma \uplus [f \mapsto ([ ]_f, 0, 0)]) \Gamma \\
\Delta_r &= \llbracket S_r \rrbracket_{\Delta} (\Sigma \uplus [f \mapsto ([ ]_f, 0, 0)]) \Gamma
\end{aligned}$$

Figure 7.8: Algorithm for computing  $\Delta_0$ .

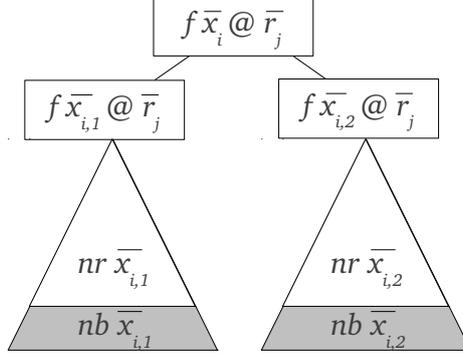


Figure 7.9: Activation tree of a call  $f \bar{x}_i @ \bar{r}_j$ . The triangles represent the activation trees of each child call.

**Example 7.40.** If we apply *computeDelta* to our *append* example by considering the *nb* and *nr* functions of Example 7.28, we get  $\Delta_b \text{ xs ys } \rho_2 = [xs \geq 1 \rightarrow 0]$  and  $\Delta_r \text{ xs ys } \rho_2 = [xs \geq 2 \rightarrow 1]$ . Therefore we get:

$$\Delta_0 \text{ xs ys } \rho_2 = [xs \geq 2 \rightarrow xs - 1] \sqcup [xs \geq 1 \rightarrow 0]$$

□

Our next step is to prove that an abstract heap  $\Delta_0$  computed by *computeDelta* falls into the reductive area of the lattice  $(\mathbb{D}, \sqsubseteq)$  with respect to the iteration of the abstract interpretation  $\mathcal{I}_f$ . As noted in Section 7.5.1, this also ensures that the result of *computeDelta* is a correct approximation to the actual charges done by *f*. Reductivity of  $\Delta_0$  holds provided some admissibility conditions on the externally given *nb* and *nr* functions hold.

**Definition 7.41** (Admissible *nb*). A function *nb* for computing the number of base calls is admissible with respect to a definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  iff for every  $\bar{x}_i^n \in \mathbb{R}^n$  the following conditions hold:

1.  $nb \bar{x}_i^n \geq 1$
2.  $\forall seq \in \text{seqs } e_f. \sum \left\{ nb \overline{|a_i| \bar{x}_i^n} \mid f \bar{a}_i^n @ \bar{r}_j^m \in seq \right\} \leq nb \bar{x}_i^n$

**Definition 7.42** (Admissible *nr*). A function *nr* for computing the number of recursive calls is admissible with respect to a definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  iff for every  $\bar{x}_i^n \in \mathbb{R}^n$  the following conditions hold:

1.  $nr \bar{x}_i^n \geq 0$
2.  $\forall seq \in \text{seqs } e_f. 1 + \sum \left\{ nr \overline{|a_i| \bar{x}_i^n} \mid f \bar{a}_i^n @ \bar{r}_j^m \in seq \right\} \leq nr \bar{x}_i^n$

In order to get an intuitive idea on these conditions, let us consider the activation tree shown in Figure 7.9. The root of this tree is a call to *f* giving place to two recursive calls, each one with an

approximation of its number of leafs  $nb \overline{x_{i,j}}$  and internal nodes  $nr \overline{x_{i,j}}$ , with  $j = 1, 2$ . Hence, the whole tree has  $nb \overline{x_{i,1}} + nb \overline{x_{i,2}}$  leafs or less. Admissibility on  $nb$  holds if the approximation  $nb \overline{x_i}$  given for the root node is greater or equal than this number. Analogously,  $nr$  is admissible if the approximation  $nr \overline{x_i}$  of the root node is greater or equal than  $1 + nr \overline{x_{i,1}} + nr \overline{x_{i,2}}$ .

Given these conditions, we are ready to prove the reductivity of  $\Delta_0$ :

**Theorem 7.43.** *Let us define  $\Delta_0 = \text{computeDelta } (f \overline{x_i} @ \overline{r_j} = e_f) \Sigma \Gamma nb nr$ . If  $nr$  and  $nb$  are admissible, and the  $\Delta_b$  and  $\Delta_r$  occurring in the definition of  $\text{computeDelta}$  are parameter-decreasing, then:*

$$\llbracket [seq]_{\Delta} (\Sigma \uplus [f \mapsto (\Delta_0, 0, 0)]) \Gamma \rrbracket \sqsubseteq \Delta_0 \quad \text{for every } seq \in \text{seqs } e_f$$

Therefore,  $\Delta_0 \in \text{Red}(\mathcal{I}_f)$

*Proof.* Let us denote by  $\Sigma'$  the signature environment  $\Sigma \uplus [f \mapsto (\Delta_0, 0, 0)]$ , and by  $\Sigma_0$  the signature environment  $\Sigma \uplus [f \mapsto ([ ]_f, 0, 0)]$ . It is easy to see that  $\text{dom } \Sigma'(f) \subseteq \text{dom } \Sigma_0(f)$ . Assume  $(S_b, S_r) = \text{splitExp } (\text{seqs } e_f)$ . Then,  $\text{seqs } e_f = S_b \uplus S_r$ . Consider a sequence  $seq \in \text{seqs } e_f$ . We distinguish cases:

- **Case  $seq \in S_b$ .** That is,  $seq$  is a base sequence. By the definition of  $\Delta_b$  we get  $\llbracket [seq]_{\Delta} \Sigma_0 \Gamma \rrbracket \sqsubseteq \Delta_b$ . The signature associated with  $f$  is not relevant, since  $seq$  does not contain calls to  $f$ . Therefore we get:

$$\llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket = \llbracket [seq]_{\Delta} \Sigma_0 \Gamma \rrbracket \sqsubseteq \llbracket \Delta_b \rrbracket \sqsubseteq \Delta_0$$

- **Case  $seq \in S_r$ .** That is,  $seq$  is a recursive sequence. Let us define  $D = \text{dom } (\llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket)$ . In  $\mathbb{R}^n \setminus D$  we get  $\llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket = \perp \sqsubseteq \Delta_0$  and the Theorem follows trivially. Thus, in the following, we consider the points belonging to  $D$ . In those points we get  $\llbracket \Delta_r \rrbracket \neq \perp$ , since  $seq \in S_r$  and hence  $D \subseteq \text{dom } \llbracket [S_r]_{\Delta} \Sigma' \Gamma \rrbracket \subseteq \text{dom } \llbracket [S_r]_{\Delta} \Sigma_0 \Gamma \rrbracket = \text{dom } \Delta_r$ . We split the elements of  $seq$  into two sets: the  $A$  set contains the elements of the form  $f \overline{a_i^n} @ \overline{r_j^m}$ , and the  $NA$  set contains the remaining ones. Let us denote by  $p$  the number of elements of  $A$ , so we get:

$$A = \{f \overline{a_{i,1}^n} @ \overline{r_j^m}, f \overline{a_{i,2}^n} @ \overline{r_j^m}, \dots, f \overline{a_{i,p}^n} @ \overline{r_j^m}\}$$

The list of region arguments  $\overline{r_j^m}$  is the same as in the function definition of  $f$ , as we assume the absence of polymorphic recursion. Since  $A \uplus NA$  contains every element of the sequence, we get the following equality in  $D$ :

$$\begin{aligned} \llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket &= \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket + \sum_{be \in A} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket \\ &= \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket + \sum_{k=1}^p \llbracket [f \overline{a_{i,k}^n} @ \overline{r_j^m}]_{\Delta} \Sigma' \Gamma \rrbracket \quad (\text{in } D) \end{aligned} \quad (7.13)$$

in which, for every  $k \in \{1..p\}$

$$\llbracket [f \overline{a_{i,k}^n} @ \overline{r_j^m}]_{\Delta} \Sigma' \Gamma \rrbracket = \lambda \overline{x_i}. \left[ G_k \rightarrow \Delta_0 \overline{a_{i,k} | \overline{x_j^n}} \right] \quad (7.14)$$

$$\text{where } G_k = \lambda \overline{x_i}. \left( \Delta_0 \overline{a_{i,k} | \overline{x_j^n}} \neq \perp \right)$$

since  $\text{unify}(\Gamma, f, \overline{r_j}) = \text{id}_{R_r}$ , due to the absence of polymorphic recursion. Moreover,  $G_k$  holds in those points belonging to  $\text{dom } (\Delta_0 \overline{a_{i,k} | \overline{x_j^n}})$  and only in those points. So, we can rewrite (7.14)

as follows:

$$\llbracket f \overline{a_{i,k}^n} @ \overline{r_j^m} \rrbracket_{\Delta} \Sigma' \Gamma = \lambda \overline{x_i} . \Delta_0 \overline{a_{i,k} | \overline{x_j}^n}$$

Let us define, for every  $k \in \{1..p\}$ :

$$\begin{aligned} \Delta_{b,k} &\stackrel{\text{def}}{=} \lambda \overline{x_i}^n . \Delta_b \overline{a_{i,k} | \overline{x_j}^n} \\ \Delta_{r,k} &\stackrel{\text{def}}{=} \lambda \overline{x_i}^n . \Delta_r \overline{a_{i,k} | \overline{x_j}^n} \\ nb_k &\stackrel{\text{def}}{=} \lambda \overline{x_i}^n . nb \overline{a_{i,k} | \overline{x_j}^n} \\ nr_k &\stackrel{\text{def}}{=} \lambda \overline{x_i}^n . nr \overline{a_{i,k} | \overline{x_j}^n} \end{aligned}$$

where  $\Delta_b, \Delta_r, nb$  and  $nr$  are as defined in *computeDelta*. We get for each  $k \in \{1..p\}$ :

$$\llbracket f \overline{a_{i,k}^n} @ \overline{r_j^m} \rrbracket_{\Delta} \Sigma' \Gamma = ([\Delta_{b,k}] * nb_k + [\Delta_{r,k}] * nr_k) \sqcup [\Delta_{b,k}] \quad (\text{in } D)$$

By substituting this into (7.13) we get:

$$\llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket = \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket + \sum_{k=1}^p ([\Delta_{b,k}] * nb_k + [\Delta_{r,k}] * nr_k) \sqcup [\Delta_{b,k}] \quad (\text{in } D) \quad (7.15)$$

For each  $k \in \{1..p\}$  We denote by  $\Delta_{f,k}$  the result of  $([\Delta_{b,k}] * nb_k + [\Delta_{r,k}] * nr_k) \sqcup [\Delta_{b,k}]$ . We prove the following fact:

$$\forall k \in \{1..p\}. \Delta_{f,k} \sqsubseteq [\Delta_b] * nb_k + [\Delta_r] * nr_k \quad (\text{in } D) \quad (7.16)$$

Let us consider a  $k \in \{1..p\}$  and define  $R_k \stackrel{\text{def}}{=} \{\overline{x_i} \in D \mid [\Delta_{b,k}] \overline{x_i} \geq ([\Delta_{b,k}] * nb_k + [\Delta_{r,k}] * nr_k) \overline{x_i}\}$ . In  $R_k$  we obtain:

$$\Delta_{f,k} = [\Delta_{b,k}] \sqsubseteq [\Delta_b] \sqsubseteq [\Delta_b] * nb_k + [\Delta_r] * nr_k \quad (\text{in } R_k)$$

The first  $\sqsubseteq$  holds because  $\Delta_b$  is parameter-decreasing, and the second  $\sqsubseteq$  is due to the fact that  $nb_k \sqsupseteq 1$ ,  $nr_k \sqsupseteq 0$ , and  $[\Delta_r] \neq \perp$ . In  $D \setminus R_k$  we get:

$$\Delta_{f,k} = [\Delta_{b,k}] * nb_k + [\Delta_{r,k}] * nr_k \sqsubseteq [\Delta_b] * nb_k + [\Delta_r] * nr_k \quad (\text{in } D \setminus R_k)$$

Again, this holds because both  $\Delta_b$  and  $\Delta_r$  are parameter-decreasing, so (7.16) holds for every  $k \in \{1..p\}$ . By substituting (7.16) into (7.15) we obtain:

$$\llbracket [seq]_{\Delta} \Sigma' \Gamma \rrbracket \sqsubseteq \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket + \sum_{k=1}^p ([\Delta_b] * nb_k + [\Delta_r] * nr_k) \quad (\text{in } D)$$

It is easy to prove that

$$\Delta_r \sqsupseteq \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma_0 \Gamma \rrbracket + \sum_{be \in A} \llbracket [be]_{\Delta} \Sigma_0 \Gamma \rrbracket = \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma_0 \Gamma \rrbracket + [ ]_f = \sum_{be \in NA} \llbracket [be]_{\Delta} \Sigma' \Gamma \rrbracket$$

in  $D$ . The first equality holds since  $\Sigma_0(f)$  is defined for every input size, so the guard occurring in

the abstract interpretation rule for function application always holds. Hence:

$$\begin{aligned}
\llbracket [seq] \rrbracket_{\Delta} \Sigma' \Gamma &\sqsubseteq \llbracket \Delta_r \rrbracket + \sum_{k=1}^p (\llbracket \Delta_b \rrbracket * nb_k + \llbracket \Delta_r \rrbracket * nr_k) \\
&= \llbracket \Delta_r \rrbracket + \llbracket \Delta_b \rrbracket * \sum_{k=1}^p nb_k + \llbracket \Delta_r \rrbracket * \sum_{k=1}^p nr_k \\
&= \llbracket \Delta_b \rrbracket * \sum_{k=1}^p nb_k + \llbracket \Delta_r \rrbracket * \left( 1 + \sum_{k=1}^p nr_k \right) \quad (\text{in } D)
\end{aligned}$$

Finally, by admissibility of  $nb$  and  $nr$ :

$$\begin{aligned}
\llbracket [seq] \rrbracket_{\Delta} \Sigma' \Gamma &\sqsubseteq \llbracket \Delta_b \rrbracket * nb + \llbracket \Delta_r \rrbracket * nr \\
&\sqsubseteq \Delta_0 \quad (\text{in } D)
\end{aligned}$$

□

This result not only proves the correctness of our initial approximation  $\Delta_0$ . It also allows to iterate the abstract interpretation in order to reach more precise bounds by considering the following chain,

$$\Delta_0 \supseteq \mathcal{I}_f(\Delta_0) \supseteq \mathcal{I}_f^2(\Delta_0) \supseteq \cdots \supseteq \mathcal{I}_f^n(\Delta_0) \supseteq \cdots$$

which, if eventually stabilizes, it does in a fixed point. Since the initial approximation to  $\mu$  depends on the input  $\Delta$  given (see next section), it is advisable to spent some time iterating  $\mathcal{I}_f$  in order to achieve better results.

**Example 7.44.** Assume the  $\Delta_0$  of Example 7.40. By applying  $\mathcal{I}_{append}$  on this abstract heap we obtain:

$$\mathcal{I}_{append}(\Delta_0) \text{ } xs \text{ } ys \text{ } \rho_2 = [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1] \sqcup [xs \geq 3 \rightarrow xs - 1]$$

which is strictly smaller than  $\Delta_0$  when  $xs \in [2, 3)$ . Another iteration yields the following result:

$$\mathcal{I}_{append}^2(\Delta_0) \text{ } xs \text{ } ys \text{ } \rho_2 = [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1] \sqcup [xs \geq 3 \rightarrow 2] \sqcup [xs \geq 4 \rightarrow xs - 1]$$

In general, the  $i$ -th iteration results in the following abstract heap:

$$\mathcal{I}_{append}^i(\Delta_0) = \lambda xs \text{ } ys \text{ } \rho_2 \mapsto \begin{cases} \perp & xs < 1 \\ \llbracket xs - 1 \rrbracket & 1 \leq xs < i + 2 \\ xs - 1 & i + 2 \leq xs \end{cases}$$

□

Notice that, in the previous example, the abstract heaps of every iteration are equal if we consider their domains only from a given threshold value. In some applications, it may be enough to obtain an expression  $xs - 1$  as an upper-bound to the costs of *append*, even though it is not a fixed point of the corresponding iteration operator. This motivates the following definition:

**Definition 7.45.** Two abstract heaps  $\Delta_1, \Delta_2 \in \mathbb{D}$  are said to be asymptotically equivalent (denoted  $\Delta_1 \approx \Delta_2$ ) if there exists some  $\bar{x}_{i,0}$  such that, for every  $\bar{x}_i$  such that  $x_i \geq x_{i,0}$  for all  $i \in \{1..n\}$ ,  $\Delta_1(\bar{x}_i) = \Delta_2(\bar{x}_i)$ . The definition of two cost functions  $\zeta_1, \zeta_2 \in \mathbb{F}$  being asymptotically equivalent is analogous.

$$\text{computeMu } (f \bar{x}_i @ \bar{r}_j = e_f) \Sigma \Gamma \Delta \text{ len} = \left( \Delta_{\text{self}} * (\text{len} - 1) + |\Delta_{\text{bef}}| + \sqcup \{ \mu_{\text{bef}}, \mu_{\text{aft}}, \mu_b \} \right) \sqcup \mu_b$$

$$\begin{aligned} \text{where } S &= \text{seqs } e_f \\ (S_b, S_r) &= \text{splitExp}_f S \\ (S_{\text{bef}}, S_{\text{aft}}) &= \text{splitBA}_f^* S_r \\ \Delta_{\text{bef}}^* &= \llbracket S_{\text{bef}} \rrbracket_{\Delta} (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \\ \Delta_{\text{bef}} &= \left\lfloor \Delta_{\text{bef}}^* \right\rfloor \\ \Delta_{\text{self}} &= \Delta_{\text{bef}}^* \rho_{\text{self}} \\ \mu_{\text{bef}} &= \llbracket S_{\text{bef}} \rrbracket_{\mu} (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \\ \mu_{\text{aft}} &= \llbracket S_{\text{aft}} \rrbracket_{\mu} (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \\ \mu_b &= \llbracket S_b \rrbracket_{\mu} (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \end{aligned}$$

Figure 7.10: Algorithm for computing  $\mu_0$ .

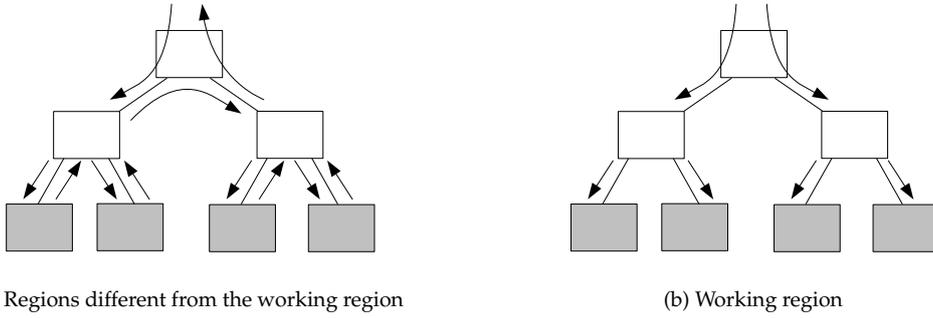


Figure 7.11: Growth of the charges done to different regions as the execution progresses.

In our example above, it holds that  $\Delta_0 \approx \mathcal{I}_{\text{append}}(\Delta_0)$ . In this case, we say that  $\Delta_0$  is an *asymptotic fixed point* of  $\mathcal{I}_{\text{append}}$ .

#### 7.5.4 Algorithm for computing $\mu_0$

The algorithm for computing a first approximation  $\mu_0$  to the heap needs of a function is shown in Figure 7.10. Let us consider the call tree of a given function call  $f$ . We make distinction between the charges done to the working regions of the calls in this tree, and the charges done to the remaining regions. The latter are cumulative, in the sense that the cells created in these regions are not removed while the execution of the root call to  $f$  progresses. The arrows in Figure 7.11a represent the directions of the execution flow, in which charges to these regions grow. With respect to the charges done to the working

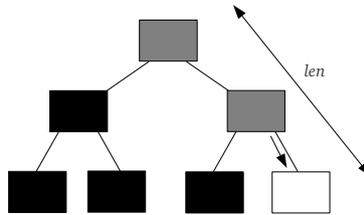


Figure 7.12: Execution point before the control flow reaches the last recursive call. Black nodes represent those calls whose execution is finished. Gray nodes represent those calls whose execution has started, but not finished.

regions of the calls of the activation tree, these only grow from the root call to the base cases, as Figure 7.11b shows. In this case we no longer have arrows pointing upwards in the tree, because all the cells created in the working region are removed when its corresponding function call finishes. Therefore, the only directions in which we know for sure that these charges grow, are the paths from the root call to its recursive children.

Now assume that, during this call to  $f$ , the execution flow has reached the point before executing the last base call (Figure 7.12). In the algorithm,  $\Delta_{self} * (len - 1)$  represents the charges to the working regions of the gray-marked calls<sup>3</sup>, whereas  $|\Delta_{bef}|$  stands for the charges done to the remaining regions during the execution of the black- and gray-marked calls. The combination of these two charges gives us  $\Delta_{self} * (len - 1) + |\Delta_{bef}|$ . Taking this value as a base level, we have to take the following charges into account:

1. Maximum level of occupied memory before the execution flow reaches the last base call, on account of the charges done in memory before the first recursive call, and between the subsequent recursive calls ( $\mu_{bef}$ ).
2. Memory needs of the last base case, which is going to be executed ( $\mu_b$ ). This corresponds to the white-marked call of Figure 7.12.
3. Memory needs of the part of the recursive cases which is still to be executed ( $\mu_{aft}$ ). This corresponds to that part of the gray-marked cells whose execution is pending.

Since none of them is necessarily greater than the other, we take the least upper bound of the three. In the same way as in our  $\Delta_0$ , we add  $\mu_b$  in the least upper bound in order to deal with  $\perp$  in the remaining components.

Since the  $\llbracket \cdot \rrbracket_\mu$  interpretation depends on the  $\llbracket \cdot \rrbracket_\Delta$  interpretation, the *computeMu* function receives an abstract heap  $\Delta$ , which is bound to  $f$  in the signature environment  $\Sigma$ .

**Example 7.46.** The algorithm applied to *append* yields the following partial results:

$$\begin{aligned} |\Delta_{bef}| &= \lambda xs ys. [xs \geq 3 \rightarrow xs - 2] \sqcup [xs \geq 2 \rightarrow 0] \\ \Delta_{self} &= \lambda xs ys. [xs \geq 2 \rightarrow 0] \\ \mu_{bef} &= \lambda xs ys. [xs \geq 2 \rightarrow 0] \\ \mu_{aft} &= \lambda xs ys. [xs \geq 2 \rightarrow 1] \\ \mu_b &= \lambda xs ys. [xs \geq 1 \rightarrow 0] \end{aligned}$$

which result in the following initial bound  $\mu_0$ :

$$\mu_0 = \lambda xs ys. [xs \geq 1 \rightarrow 0] \sqcup [xs \geq 2 \rightarrow 1] \sqcup [xs \geq 3 \rightarrow xs - 1]$$

By applying the  $\mathcal{M}$  operator to this  $\mu_0$ , we get an asymptotically equal upper bound. □

Again, the reductivity of the  $\mu_0$  computed in this way depends on some admissibility conditions of the *len* function, which bounds the length of longest call chain.

**Definition 7.47** (Admissible *len*). A function *len* for computing the maximal length of call chains is admissible with respect to a definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  iff for every  $\bar{x}_i^n \in \mathbb{R}^n$  the following conditions hold:

<sup>3</sup>We assume the worst-case execution in which the longest call chain is the one who leads to the last base call.

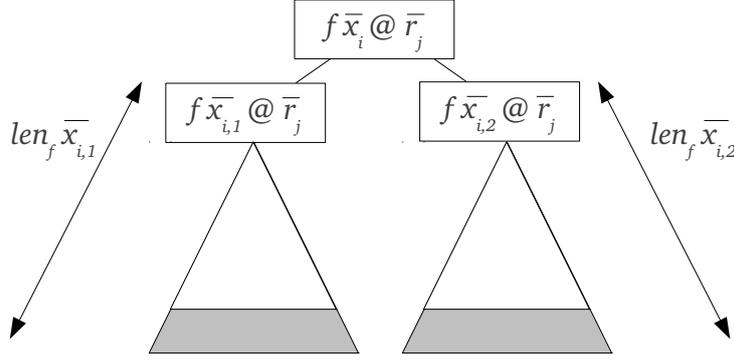


Figure 7.13: Activation tree of a call  $f \bar{x}_i @ \bar{r}_j$ . The triangles represent the activation trees of each child call.

1.  $len \bar{x}_i^n \geq 1$
2.  $\forall seq \in seqs e_f. 1 + \sqcup \left\{ len \overline{a_i} \bar{x}_i^n \mid f \bar{a}_i^n @ \bar{r}_j^m \in seq \right\} \leq len \bar{x}_i^n$
3. If  $(S_b, S_r) = splitExp_f (seqs e_f)$ , for every  $seq \in S_r$ ,  $len \bar{x}_i \geq 2$  whenever  $guard(seq) \bar{x}_i$  holds.

The first admissibility condition is fairly reasonable. The second one is analogous to its counterparts in  $nb$  and  $nr$ . Assume the situation given in Figure 7.13. An upper bound to the height of the whole tree is  $1 + \sqcup \{len \bar{x}_{i,1}, len \bar{x}_{i,2}\}$ . The second condition states that the approximated height  $len \bar{x}_i$  must be greater or equal than this bound. Finally, the third condition states that, in those values  $\bar{x}_i$  which may possibly lead to a recursive call,  $len$  must be greater than two (accounting for the caller and the callee).

Assuming these admissibility conditions, we can prove the reductivity of  $\mu_0$ :

**Theorem 7.48.** *Let us define  $\mu_0 = computeMu (f \bar{x}_i @ \bar{r}_j = e_f) \Sigma \Gamma \Delta len$ . If  $len$  is admissible, the  $\Delta_{self}$ ,  $\mu_{bef}$ ,  $\mu_{aft}$ ,  $\mu_b$  occurring in  $computeMu$  are parameter-decreasing, and  $\llbracket seq \rrbracket_\Delta (\Sigma \uplus [f \mapsto (\Delta, 0, 0)]) \Gamma \sqsubseteq \Delta$  for every  $seq \in seqs e_f$ , then:*

$$\llbracket seq \rrbracket_\mu (\Sigma \uplus [f \mapsto (\Delta, \mu_0, 0)]) \Gamma \sqsubseteq \mu_0 \quad \text{for every } seq \in seqs e_f$$

Therefore,  $\mu_0 \in Red(\mathcal{M}_{\Delta, f})$ .

*Proof.* We denote by  $\Sigma'$  and  $\Sigma_0$  the signature environments  $\Sigma \uplus [f \mapsto (\Delta, \mu_0, 0)]$  and  $\Sigma \uplus [f \mapsto (\Delta, 0, 0)]$  respectively. It is straightforward to prove that  $dom \Sigma'(f) \subseteq dom \Sigma_0(f)$ . Let  $seq \in seqs e_f$ . If  $S_b$  and  $S_r$  are as defined in  $computeMu$ , we know that  $seq \in S_b \uplus S_r$ , so we distinguish cases:

- **Case  $seq \in S_b$ .** In other words,  $seq$  is a base sequence. In this case  $\llbracket seq \rrbracket_\mu \Sigma' \Gamma = \llbracket seq \rrbracket_\mu \Sigma_0 \Gamma$ , as  $seq$  does not contain calls to  $f$ . In this case,

$$\llbracket seq \rrbracket_\mu \Sigma' \Gamma = \llbracket seq \rrbracket_\mu \Sigma_0 \Gamma \sqsubseteq \mu_b \sqsubseteq \mu_0$$

where  $\mu_b$  is as defined in  $computeMu$ .

- **Case  $seq \in S_r$ .** That is,  $seq$  is recursive. We define  $D = dom (\llbracket seq \rrbracket_\mu \Sigma' \Gamma)$ . The Theorem follows trivially in those points  $\bar{x}_i$  not belonging to  $D$ , so we assume  $\bar{x}_i \in D$  in the following. Since our sequence is recursive, we can split it into two subsequences  $seq_{bef}$  and  $seq_{aft}$  such that

$seq_{bef} ++ seq_{aft} = seq$ ,  $seq_{bef} \in S_{bef}$  and  $seq_{aft} \in S_{aft}$ . As a consequence of Lemma 7.20, we know that

$$\begin{aligned} \text{dom} (\llbracket seq \rrbracket_{\mu} \Sigma' \Gamma) &= \text{dom} (\llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma) \cap \text{dom} (\llbracket seq_{aft} \rrbracket_{\mu} \Sigma' \Gamma) \\ &\subseteq \text{dom} (\llbracket seq_{bef} \rrbracket_{\mu} \Sigma_0 \Gamma) \cap \text{dom} (\llbracket seq_{aft} \rrbracket_{\mu} \Sigma_0 \Gamma) \end{aligned}$$

which implies  $\mu_{bef} \bar{x}_i \neq \perp$  and  $\mu_{aft} \bar{x}_i \neq \perp$  for every  $\bar{x}_i \in D$ . Moreover, and by Lemma 7.18, we get  $\Delta_{bef}^* \bar{x}_i \neq \perp$  for all  $\bar{x}_i \in D$ , which implies  $|\Delta_{bef}| \bar{x}_i \neq \perp$  and  $|\Delta_{self}| \bar{x}_i \neq \perp$  for every  $\bar{x}_i \in D$ .

Let us start with the computation of  $\llbracket seq \rrbracket_{\mu} \Sigma' \Gamma$ . By Lemma 7.20:

$$\begin{aligned} \llbracket seq \rrbracket_{\mu} \Sigma' \Gamma &= \llbracket seq_{bef} ++ seq_{aft} \rrbracket_{\mu} \Sigma' \Gamma \\ &= \sqcup \{ \llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma, \llbracket seq_{bef} \rrbracket_{\Delta} \Sigma' \Gamma + \llbracket seq_{aft} \rrbracket_{\mu} \Sigma' \Gamma \} \end{aligned}$$

Since the result of  $\llbracket seq_{bef} \rrbracket_{\Delta}$  does not depend on the  $\mu$  value given by  $\Sigma'(f)$ , and there are not calls to  $f$  in  $seq_{aft}$ , we can substitute  $\Sigma_0$  for  $\Sigma'$  in the previous equation in order to get:

$$\begin{aligned} \llbracket seq \rrbracket_{\mu} \Sigma' \Gamma &= \sqcup \{ \llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma, \llbracket seq_{bef} \rrbracket_{\Delta} \Sigma_0 \Gamma + \llbracket seq_{aft} \rrbracket_{\mu} \Sigma_0 \Gamma \} \\ &\subseteq \sqcup \{ \llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma, |\Delta_{bef}^*| + \mu_{aft} \} \end{aligned}$$

The Theorem holds if we can prove that each element of the  $\sqcup$  is subsumed by  $\mu_0$ . Let us start with the second one. Since  $len \sqsupseteq 2$  (in  $D$ ) we get:

$$\begin{aligned} |\Delta_{bef}^*| + \mu_{aft} &= |\Delta_{bef}| + \Delta_{self} + \mu_{aft} \\ &\subseteq |\Delta_{bef}| + \Delta_{self} * (len - 1) + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \} \\ &\subseteq \mu_0 \quad (\text{in } D) \end{aligned}$$

Now we prove that  $\llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma \subseteq \mu_0$  (in  $D$ ). By unfolding the definition of  $\llbracket seq_{bef} \rrbracket_{\mu} \Sigma' \Gamma$ , this inequality can be rewritten as follows:

$$\sqcup_{k=1}^n \left( \sum_{j=1}^{k-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| + \llbracket be_k \rrbracket_{\mu} \Sigma' \Gamma \right) \subseteq \mu_0$$

where  $n$  is the number of elements in  $seq_{bef}$ . It is enough to prove that each element of the  $\sqcup$  is dominated by  $\mu_0$ . We denote by  $\zeta_k$  the  $k$ -th element of the  $\sqcup$ . Let  $k \in \{1..n\}$ . We distinguish cases:

–  $be_k$  does not have the form  $f \bar{a}_{i,k} @ \bar{r}_j$ .

In this case  $\llbracket be_k \rrbracket_{\mu} \Sigma' \Gamma = \llbracket be_k \rrbracket_{\mu} \Sigma_0 \Gamma$ . In addition,  $\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma = \llbracket be_j \rrbracket_{\Delta} \Sigma_0 \Gamma$  for each  $j \in \{1..k-1\}$ . Hence we obtain (in  $D$ ):

$$\begin{aligned} \zeta_k &= \sum_{j=1}^{k-1} (|\llbracket be_j \rrbracket_{\Delta} \Sigma_0 \Gamma|) + \llbracket be_k \rrbracket_{\mu} \Sigma_0 \Gamma \\ &\subseteq \mu_{bef} \\ &\subseteq \Delta_{self} * (len - 1) + |\Delta_{bef}| + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \} \\ &\subseteq \mu_0 \quad (\text{in } D) \end{aligned}$$

since  $len \sqsupseteq 1$  (in  $D$ ) and  $|\Delta_{bef}|, \Delta_{self}$  are distinct from  $\perp$  in  $D$ .

–  $be_k$  has the form  $f \bar{a}_{i,k} @ \bar{r}_j$ .

Let  $P$  be the set of  $k \in \{1..n\}$  such that  $be_k$  is a function call to  $f$ . We define, for each  $k \in P$ :

$$\begin{aligned}
\mu_{f,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \mu_0 \overline{|a_{i,k}| \bar{x}_i^n} \\
\Delta_{bef,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \Delta_{bef} \overline{|a_{i,k}| \bar{x}_i^n} \\
\Delta_{self,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \Delta_{self} \overline{|a_{i,k}| \bar{x}_i^n} \\
\Delta_{f,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \Delta_0 \overline{|a_{i,k}| \bar{x}_i^n} \\
\mu_{bef,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \mu_{bef} \overline{|a_{i,k}| \bar{x}_i^n} \\
\mu_{aft,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \mu_{aft} \overline{|a_{i,k}| \bar{x}_i^n} \\
\mu_{b,k} &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . \mu_b \overline{|a_{i,k}| \bar{x}_i^n} \\
len_k &\stackrel{\text{def}}{=} \lambda \bar{x}_i^n . len \overline{|a_{i,k}| \bar{x}_i^n}
\end{aligned}$$

We get (in  $D$ )

$$\begin{aligned}
\zeta_k &= \sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma' \Gamma| + \mu_{f,k} \\
&= \sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma' \Gamma| + \left( (\Delta_{self,k} * (len_k - 1) + |\Delta_{bef,k}| + \sqcup \{ \mu_{bef,k}, \mu_{aft,k}, \mu_{b,k} \}) \sqcup \mu_{b,k} \right)
\end{aligned}$$

Since  $\Delta_{self}$ ,  $\mu_{bef}$ ,  $\mu_{aft}$  and  $\mu_b$  are parameter-decreasing we get:

$$\zeta_k \sqsubseteq \sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma' \Gamma| + \left( (\Delta_{self} * (len_k - 1) + |\Delta_{bef,k}| + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \}) \sqcup \mu_b \right)$$

Let  $R_k$  be the set of elements  $\bar{x}_i \in D$  such that  $(\Delta_{self} * (len_k - 1) + |\Delta_{bef,k}| + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \}) \bar{x}_i \leq \mu_b \bar{x}_i$ . We obtain, in  $R_k$

$$\zeta_k \sqsubseteq \sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma' \Gamma| + \mu_b \quad (\text{in } R_k) \quad (7.17)$$

Again, for each  $j \in \{1..k-1\}$  it holds that  $[\![be_j]\!]_{\Delta} \Sigma' \Gamma = [\![be_j]\!]_{\Delta} \Sigma_0 \Gamma$ . Thus we get:

$$\sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma' \Gamma| = \sum_{j=1}^{k-1} |[\![be_j]\!]_{\Delta} \Sigma_0 \Gamma| \sqsubseteq \sum_{j=1}^n |[\![be_j]\!]_{\Delta} \Sigma_0 \Gamma| \sqsubseteq |\Delta_{bef}^*| \quad (\text{in } D) \quad (7.18)$$

provided  $([\![be_j]\!]_{\Delta} \Sigma_0 \Gamma) \bar{x}_i \neq \perp$  for every  $\bar{x}_i \in D$  and  $j \in \{k..n\}$ . However, if we had some  $\bar{x}_i \in D$  such that  $([\![be_j]\!]_{\Delta} \Sigma_0 \Gamma) \bar{x}_i = \perp$  for some  $j \in \{k..n\}$  we would have  $([\![seq]\!]_{\Delta} \Sigma_0 \Gamma) \bar{x}_i = \perp$  contradicting the fact that  $\bar{x}_i \in D$  since  $D = \text{dom } [\![seq]\!]_{\mu} \Sigma' \Gamma = \text{dom } [\![seq]\!]_{\Delta} \Sigma' \Gamma \subseteq \text{dom } [\![seq]\!]_{\Delta} \Sigma_0 \Gamma$ . Hence we can rewrite the inequality (7.17) above by using (7.18) as follows (in  $R_k$ ):

$$\begin{aligned}
\zeta_k &\sqsubseteq |\Delta_{bef}^*| + \mu_b \\
&= |\Delta_{bef}| + |\Delta_{self}| + \mu_b \\
&\sqsubseteq |\Delta_{bef}| + |\Delta_{self}| * (len - 1) + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \} \\
&\sqsubseteq \mu_0 \quad (\text{in } R_k)
\end{aligned}$$

where the third step holds by admissibility of  $len$ . On the other hand we get, in  $D \setminus R_k$ :

$$\begin{aligned}
\zeta_k &\sqsubseteq \sum_{j=1}^{k-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| + \Delta_{self} * (len_k - 1) + |\Delta_{bef,k}| + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \\
&\sqsubseteq \sum_{j=1}^{k-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| + \Delta_{self} * \left( \bigsqcup_{p \in P} len_p - 1 \right) \\
&\quad + |\Delta_{bef,k}| + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \quad (\text{in } D \setminus R_k)
\end{aligned} \tag{7.19}$$

Now we prove that  $|\Delta_{bef,k}| \sqsubseteq |\Delta_{f,k}|$ . Let us consider a sequence  $seq' \in S_r$  of the form  $[G \rightarrow be'_1, \dots, be'_m]$ . This sequence can be split into two subsequences  $seq'_{bef}$  and  $seq'_{aft}$ , each guarded by  $G$ , as done by  $splitBA_f$ . Notice that  $\text{dom} \llbracket seq'_{bef} \rrbracket_{\Delta} \Sigma \Gamma \subseteq \text{dom} \llbracket seq'_{aft} \rrbracket_{\Delta} \Sigma \Gamma$  for every  $\Sigma$  and  $\Gamma$ , as  $seq'_{aft}$  does not contain calls to  $f$ . Therefore:

$$\left[ \llbracket seq'_{bef} \rrbracket_{\Delta} \Sigma_0 \Gamma \right] \sqsubseteq \left[ \llbracket seq'_{bef} \rrbracket_{\Delta} \Sigma_0 \Gamma \right] + \left[ \llbracket seq'_{aft} \rrbracket_{\Delta} \Sigma_0 \Gamma \right] = \llbracket seq' \rrbracket_{\Delta} \Sigma_0 \Gamma = \llbracket seq' \rrbracket_{\Delta} \Sigma' \Gamma \sqsubseteq \Delta$$

where the last step follows by assumption (reductivity of  $\Delta$ ). Since  $\Delta_{bef}$  is the least upper bound of all the  $\left[ \llbracket seq'_{bef} \rrbracket_{\Delta} \Sigma_0 \Gamma \right]$ , then  $\Delta_{bef} \sqsubseteq \Delta$ , which implies  $|\Delta_{bef}| \sqsubseteq |\Delta|$  and, in turn,  $|\Delta_{bef,k}| \sqsubseteq |\Delta_{f,k}|$ . We apply this into (7.19) so as to get:

$$\zeta_k \sqsubseteq \sum_{j=1}^{k-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| + \Delta_{self} * \left( \bigsqcup_{p \in P} len_p - 1 \right) + |\Delta_{f,k}| + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \quad (\text{in } D \setminus R_k)$$

In a similar way as with the set  $R_k$ , in  $D \setminus R_k$  it holds that

$$\sum_{j=1}^{k-1} |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| + |\Delta_{f,k}| = \sum_{j=1}^k |\llbracket be_j \rrbracket_{\Delta} \Sigma' \Gamma| = \sum_{j=1}^k |\llbracket be_j \rrbracket_{\Delta} \Sigma_0 \Gamma| \sqsubseteq \sum_{j=1}^n |\llbracket be_j \rrbracket_{\Delta} \Sigma_0 \Gamma| = |\Delta_{bef}^*|$$

hence,

$$\begin{aligned}
\zeta_k &\sqsubseteq |\Delta_{bef}^*| + \Delta_{self} * \left( \bigsqcup_{p \in P} len_p - 1 \right) + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \\
&= |\Delta_{bef}| + \Delta_{self} + \Delta_{self} * \left( \bigsqcup_{p \in P} len_p - 1 \right) + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \\
&= |\Delta_{bef}| + \Delta_{self} * \left( \bigsqcup_{p \in P} len_p \right) + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \\
&\sqsubseteq |\Delta_{bef}| + \Delta_{self} * (len - 1) + \sqcup \{\mu_{bef}, \mu_{aft}, \mu_b\} \quad (\text{in } D \setminus R_k)
\end{aligned}$$

where the last step follows by admissibility of  $len$ . Hence  $\zeta_k \sqsubseteq \mu_0$ , which proves the theorem.  $\square$

Notice that there is an additional constraint in the assumptions of this theorem: the  $\Delta$  given as parameter must satisfy the conclusions of Theorem 7.43. This assumption holds, in particular, if  $\Delta$  is either the initial approximation  $\Delta_0$  computed by the algorithm of Section 7.5.3, or the result of applying the abstract interpretation  $n > 0$  times to it,  $\mathcal{I}_f^n(\Delta_0)$ .

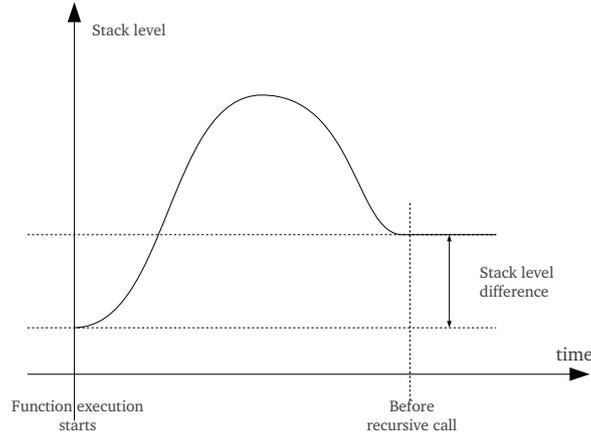


Figure 7.14: The stack level difference value denotes the maximum difference between the stack levels when the function starts executing and before its recursive call is done.

$$\begin{aligned}
SD_f (f \bar{a}_i^n @ \bar{r}_j^m) td &= n + m - td \\
SD_f be td &= \perp \quad \text{if } be \text{ is not a call to } f \\
SD_f (\mathbf{let } x_1 = e_1 \mathbf{in } e_2) td &= \sqcup \{2 + SD_f e_1 0, 1 + SD_f e_2 (td + 1)\} \\
SD_f (\mathbf{case } x \mathbf{of } \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i) td &= \sqcup_{r=1}^n (n_r + SD_f e_r (td + n_r))
\end{aligned}$$

Figure 7.15: Definition of  $SD_f$ , which computes the stack level difference.

### 7.5.5 Algorithm for computing $\sigma_0$

In order to approximate the stack costs of a function we follow an approach similar to that of  $\mu_0$ . In this case we do not have cumulative components such as the  $\Delta_{bef}$  shown before, because the behaviour of the stack, in this sense, is analogous to that of the *self* region in the heap: it grows as the execution flow descends the activation tree. We use the term *stack level* to denote the number of words existing in the stacks at a given execution time. It is useful to obtain the maximum difference between the stack levels in two execution points: when the context function  $f$  starts its execution, and when a recursive call to  $f$  is going to be done (Figure 7.14). We use the term *stack level difference* to refer to this value, which is not expressed as an element of  $\mathbb{F}$ , but as an element of  $\mathbb{N}^\perp$  instead. This is because this value does not depend on the input sizes as the rest of the components we have seen so far.

The  $SD$  function computes the stack level difference for a given expression in the context of a function definition. If this expression does not contain recursive calls, it returns  $\perp$ . If  $e_f$  is the body of the context function  $f$ , with  $n$  data parameters and  $m$  region parameters, the result of  $SD_f e_f (n + m)$  is the stack level difference of this function. A formal proof of this fact is not necessary at this moment and it will be deferred to Appendix B.

Figure 7.16 shows the algorithm for computing an initial approximation  $\sigma_0$  to the stack consumption. The intuitive idea is that, if we replicate the behaviour shown in Figure 7.14 along a number  $len$  of nested recursive calls, we get the graph shown in Figure 7.17. At the point in which the last base case is about to

$$\begin{aligned}
\text{computeSigma } (f \bar{x}_i^n @ \bar{r}_j^m = e_f) \Sigma td len &= \sqcup \{0, SD_f e_f (n + m)\} * (len - 1) + \sigma \\
\mathbf{where } \sigma &= \llbracket e_f \rrbracket_\sigma (\Sigma \uplus [f \mapsto ([ ]_f, 0, 0)]) td
\end{aligned}$$

Figure 7.16: Computation of an initial approximation  $\sigma_0$ .

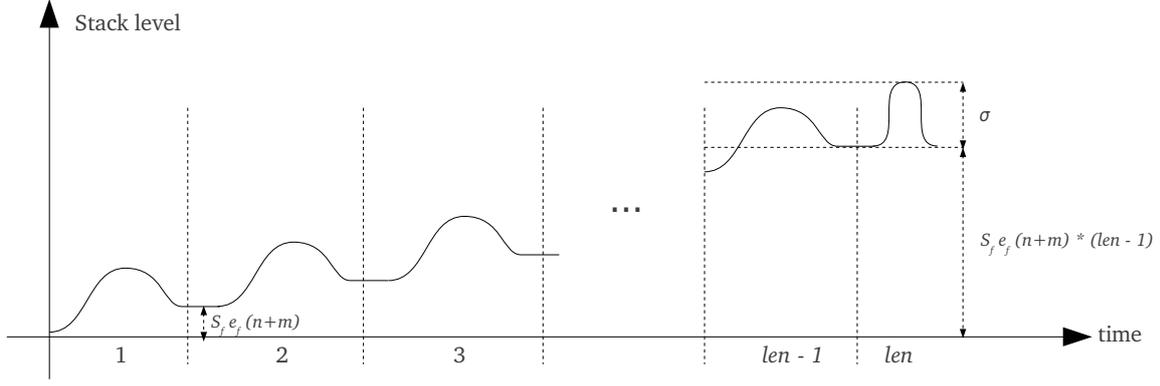


Figure 7.17: Stack level behaviour during the execution of subsequent recursive calls.

be executed, the stack level reaches  $SD e_f (n + m) * (len - 1)$  words. Then we have to proceed similarly as in the computation of  $\mu_0$ : taking the value  $SD e_f (n + m) * (len - 1)$  as a base level, we should consider the stack costs on account of the base cases and the part of the recursive cases before and after the last recursive call. If we denote these components by  $\sigma_b$ ,  $\sigma_{bef}$  and  $\sigma_{aft}$  respectively, we have to take the least upper bound  $\sqcup\{\sigma_b, \sigma_{bef}, \sigma_{aft}\}$ . However, and because the absence of cumulative components in the computation of stack needs (unlike the heap needs  $\mu$ , in which the cumulative  $\Delta$  component is involved), taking the least upper bound of these three components is equivalent to applying the  $\llbracket \cdot \rrbracket_\sigma$ -interpretation to the whole expression, without taking into account the costs of the subsequent recursive calls. That is what the  $\sigma$  component represents.

**Example 7.49.** The algorithm *computeSigma*, when applied to the *append* function, yields the following intermediate results,

$$SD_{append} (e_{append}) 3 = 7 \quad \sigma = [xs \geq 1 \rightarrow 1] \sqcup [xs \geq 2 \rightarrow 7]$$

leading to  $\sigma_0 = [xs \geq 1 \rightarrow 7xs - 6] \sqcup [xs \geq 2 \rightarrow 7xs]$ , which is an asymptotic fixed point.  $\square$

As in the previous cases, we have to prove the reductivity of the  $\sigma_0$  obtained in this way. Before this, we need two auxiliary results: the first one states the minimum value that  $SD_f$  can return for a given function  $f$ .

**Lemma 7.50.** Let  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  be a function definition. Then  $SD_f e_f td \geq n + m - td$  for all  $td$  and every subexpression  $e$  of  $e_f$  in which there is a recursive call to  $f$ .

*Proof.* By induction on the structure of  $e$ . If  $e$  is a recursive call to  $f$  the property holds trivially. We consider the remaining cases:

- **Case**  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$

If there is a call to  $f$  in  $e_2$  then:

$$SD_f e_f td \geq 1 + SD_f e_2 (td + 1) \geq 1 + n + m - (td + 1) = n + m - td$$

If there is no call to  $f$  in  $e_2$ , there must be at least one call in  $e_1$  and hence:

$$SD_f e_f td \geq 2 + SD_f e_1 0 \geq 2 + n + m - 0 \geq n + m - td$$

- **Case**  $e \equiv \text{case } x \text{ of } \overline{C_i \overline{x_{ij}^{n_i}} \rightarrow e_i^n}$

Let us assume that there is a call to  $f$  in the  $r$ -th branch of the **case**. Then:

$$SD_f e \text{ td} \geq n_r + SD_f e_r (td + n_r) \geq n_r + n + m - (td + n_r) = n + m - td$$

□

As a consequence of this Lemma, it holds that  $SD_f e (n + m) \geq 0$  for every function definition  $f \overline{x_i^n} @ \overline{r_j^m} = e_f$ . The second auxiliary lemma allows us to express the result of the  $\llbracket \cdot \rrbracket_\sigma$ -interpretation as the least upper bounds of the stack costs of the function itself (without taking recursive calls into account), and the stack costs of its recursive calls (taking the maximum stack level difference as a base level).

**Lemma 7.51.** *Let  $f$  be the context function, and  $e$  a subexpression of its body which contains  $p$  recursive calls to  $f$ . Let  $\sigma' \in \mathbb{F}$  a stack cost function. For each recursive call  $k \in \{1..p\}$  we denote by  $\sigma'_k$  the stack cost function  $\lambda \overline{x_i} . \sigma' (|a_{ki}| \overline{x_i})$ , where  $|a_{ki}|$  is the size function of the  $i$ -th argument occurring in the  $k$ -th recursive call. If we define:*

$$\begin{aligned} \sigma &= \llbracket e \rrbracket_\sigma (\Sigma \uplus [f \mapsto ([ ]_f, 0, \sigma')]) \Gamma \text{ td} \\ \sigma_r &= \llbracket e \rrbracket_\sigma (\Sigma \uplus [f \mapsto ([ ]_f, 0, 0)]) \Gamma \text{ td} \end{aligned}$$

then the following relation between  $\sigma$ ,  $\sigma_r$  and  $\sigma'$  holds:

$$\sigma \sqsubseteq \sqcup \left( \{ \sigma_r \} \cup \left\{ SD_f e \text{ td} + \sigma'_k \mid k \in \{1 \dots p\} \right\} \right)$$

*Proof.* By induction on the structure of  $e$ .

- **Case**  $e \equiv be$ , where  $be$  is not a recursive call.

In this case we get  $p = 0$  and the signature of  $f$  is not relevant. Thus  $\sigma = \sigma_r$ , from which the required result follows trivially.

- **Case**  $e \equiv f \overline{a_i^n} @ \overline{r_j^m}$

We get  $p = 1$ ,  $\sigma = \sqcup \{n + m, \sigma'_1 - td + n + m\}$  and  $\sigma_r = \sqcup \{n + m, n + m - td\} = n + m$  and hence:

$$\sigma = \sqcup \{n + m, n + m - td + \sigma'_1\} = \sqcup \{ \sigma_r, SD_f e \text{ td} + \sigma'_1 \} \sqsubseteq \sqcup \{ \sigma_r, SD_f e \text{ td} + \sigma'_1 \}$$

- **Case**  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$

We assume there are  $q \in \{0 \dots p\}$  calls to  $f$  in  $e_1$  and  $q - p$  calls to  $f$  in  $e_2$ . If we denote by  $\sigma_1$  and  $\sigma_2$  the stack costs obtained in  $e_1$  and  $e_2$ , we can apply the induction hypothesis as follows:

$$\begin{aligned} \sigma_1 &\sqsubseteq \sqcup \left( \{ \sigma_{1,r} \} \cup \{ SD_f e_1 \text{ td} + \sigma'_k \mid k \in \{1 \dots q\} \} \right) \\ \sigma_2 &\sqsubseteq \sqcup \left( \{ \sigma_{2,r} \} \cup \{ SD_f e_2 (td + 1) + \sigma'_k \mid k \in \{q + 1 \dots p\} \} \right) \end{aligned}$$

Consider the set  $D = \text{dom } \sigma_1 \cap \text{dom } \sigma_2$ . If  $\overline{x_i} \notin D$  then we get:

$$\sigma \overline{x_i} = \sqcup \{ 2 + \sigma_1 \overline{x_i}, 1 + \sigma_2 \overline{x_i} \} = \perp$$

and the lemma holds trivially. On the other hand, if  $\bar{x}_i \in D$  then  $\sigma_1 \bar{x}_i \neq \perp$  and  $\sigma_2 \bar{x}_i \neq \perp$ , which implies, by the definition of  $\sigma_{1,r}$  and  $\sigma_{2,r}$ ,  $\sigma_{1,r} \bar{x}_i \neq \perp$  and  $\sigma_{2,r} \bar{x}_i \neq \perp$ . Therefore we get, in  $D$ :

$$\begin{aligned}
& \sigma \\
= & \quad \{ \text{by definition of } \llbracket \cdot \rrbracket_\sigma \} \\
& \sqcup \{2 + \sigma_1, 1 + \sigma_2\} \\
\sqsubseteq & \quad \{ \text{by I.H.} \} \\
& \sqcup \{2 + \sqcup (\{\sigma_{1,r}\} \cup \{SD_f e_1 0 + \sigma'_k \mid k \in \{1 \dots q\}\}), \\
& \quad 1 + \sqcup (\{\sigma_{2,r}\} \cup \{SD_f e_2 (td + 1) + \sigma'_k \mid k \in \{q + 1 \dots p\}\})\} \\
= & \quad \{ \text{by properties of } \sqcup, \text{ and since } \sigma_1, \sigma_2 \neq \perp \text{ in } D \} \\
& \sqcup \{ \sqcup \{2 + \sigma_{1,r}, 1 + \sigma_{2,r}\}, \sqcup \{2 + SD_f e_1 0 + \sigma'_k \mid k \in \{1 \dots q\}\}, \\
& \quad \sqcup \{1 + SD_f e_2 (td + 1) + \sigma'_k \mid k \in \{q + 1 \dots p\}\} \} \\
\sqsubseteq & \quad \{ \text{by definition of } SD_f \text{ and } \sigma_r, \text{ and since } \sigma_{1,r}, \sigma_{2,r} \neq \perp \text{ in } D \} \\
& \sqcup \{ \sigma_r, \sqcup \{SD_f e td + \sigma'_k \mid k \in \{1 \dots q\}\}, \\
& \quad \sqcup \{SD_f e (td) + \sigma'_k \mid k \in \{q + 1 \dots p\}\} \} \\
= & \quad \{ \text{by properties of } \sqcup \} \\
& \sqcup \{ \sigma_r, \sqcup \{SD_f e td + \sigma'_k \mid k \in \{1 \dots p\}\} \} \\
= & \quad \{ \text{by properties of } \sqcup \} \\
& \sqcup (\{ \sigma_r \} \cup \{SD_f e td + \sigma'_k \mid k \in \{1 \dots p\}\})
\end{aligned}$$

- **Case  $e \equiv$  case  $x$  of  $\overline{C_i \bar{x}_i^{n_i} \rightarrow e_i^n}$**

The  $p$  recursive calls are distributed among the  $e_i$ , so we assume a sequence  $1 = p_0 \leq p_1 \leq \dots \leq p_n = p$  such that there are  $p_j - p_{j-1}$  recursive calls in the  $j$ -th branch. If we denote by  $\sigma_j$  ( $j \in \{1 \dots n\}$ ) the result of applying the abstract interpretation rules to the  $j$ -th branch, we get:

$$\begin{aligned}
& \sigma \\
= & \quad \{ \text{by definition of } \llbracket \cdot \rrbracket_\sigma \} \\
& \sqcup_{j=1}^n \{n_j + \sigma_j\} \\
\sqsubseteq & \quad \{ \text{by I.H.} \} \\
& \sqcup_{j=1}^n \{n_j + \sqcup (\{\sigma_{j,r}\} \cup \{SD_f e_j (td + n_j) + \sigma'_k \mid k \in \{p_{j-1} \dots p_j\}\})\} \\
= & \quad \{ \text{by properties of } \sqcup \} \\
& \sqcup \{ \sqcup_{j=1}^n \{n_j + \sigma_{j,r}\}, \sqcup_{j=1}^n \{n_j + SD_f e_j (td + n_j) + \sigma'_k \mid k \in \{p_{j-1} \dots p_j\}\} \} \\
\sqsubseteq & \quad \{ \text{by defs. of } SD_f \text{ and } \sigma_r \} \\
& \sqcup \{ \sigma_r, \sqcup_{j=1}^n \{SD_f e td + \sigma'_k \mid k \in \{p_{j-1} \dots p_j\}\} \} \\
& \\
= & \quad \{ \text{by properties of } \sqcup \} \\
& \sqcup (\{ \sigma_r \} \cup \{SD_f e td + \sigma'_k \mid k \in \{1 \dots p\}\})
\end{aligned}$$

□

Given these two properties, we can prove the reductivity of  $\sigma_0$  provided the *len* function involved in its computation is admissible (see Definition 7.47).

**Theorem 7.52.** *Let  $\sigma_0 = \text{computeSigma}(f \bar{x}_i @ \bar{r}_j = e_f) \Sigma td \text{ len}$ . If *len* is admissible and the  $\sigma$  occurring in the definition of *computeSigma* is parameter-decreasing, then  $\sigma_0 \in \text{Red}(\mathcal{S}_f)$ .*

*Proof.* If  $f$  is not recursive, we get:

$$\mathcal{S}_f(\sigma_0) = \llbracket e \rrbracket_{\sigma} \Sigma \uplus [f \mapsto ([ ]_f, 0, \sigma)] \Gamma (n + m) = \llbracket e \rrbracket_{\sigma} \Sigma [f \mapsto ([ ]_f, 0, 0)] \Gamma (n + m) = \sigma \sqsubseteq \sigma_0$$

and reductivity of  $\sigma_0$  holds. Now let us assume that there are  $p \geq 1$  recursive calls in the function definition. For each  $k \in \{1..p\}$  we define  $\sigma_{f,k}$ ,  $len_k$  and  $\sigma_k$  as follows.

$$\sigma_{f,k} \stackrel{\text{def}}{=} \lambda \bar{x}^n. \sigma_0 (\overline{|a_{ki}| \bar{x}_i}) \quad len_k \stackrel{\text{def}}{=} \lambda \bar{x}^n. len (\overline{|a_{ki}| \bar{x}}) \quad \sigma_k \stackrel{\text{def}}{=} \lambda \bar{x}^n. \sigma (\overline{|a_{ki}| \bar{x}})$$

where the  $|a_{ki}|$  are size functions previously inferred for each parameter in each recursive call. Then:

$$\begin{aligned} & \mathcal{S}_f(\sigma_0) \\ \sqsubseteq & \quad \{ \text{by Lemma 7.51} \} \\ & \sqcup \left( \{ \sigma \} \cup \{ SD_f e_f (n + m) + \sigma_{f,k} \mid k \in \{1 \dots p\} \} \right) \\ = & \quad \{ \text{by Lemma 7.50, } SD_f e_f (n + m) \geq 0 \} \\ & \sqcup \left( \{ \sigma \} \cup \{ \sqcup \{ 0, SD_f e_f (n + m) \} + \sigma_{f,k} \mid k \in \{1 \dots p\} \} \right) \\ = & \quad \{ \text{by definition of } \sigma_0 \} \\ & \sqcup \left( \{ \sigma \} \cup \{ \sqcup \{ 0, SD_f e_f (n + m) \} + \sqcup \{ 0, SD_f e_f (n + m) \} * (len_k - 1) + \sigma_k \mid k \in \{1 \dots p\} \} \right) \\ = & \\ & \sqcup \left( \{ \sigma \} \cup \{ \sqcup \{ 0, SD_f e_f (n + m) \} * len_k + \sigma_k \mid k \in \{1 \dots p\} \} \right) \\ \sqsubseteq & \quad \{ \text{admissibility of } len \} \\ & \sqcup \left( \{ \sigma \} \cup \{ \sqcup \{ 0, SD_f e_f (n + m) \} * (len - 1) + \sigma_k \mid k \in \{1 \dots p\} \} \right) \\ \sqsubseteq & \quad \{ \text{since } \sigma_k \sqsubseteq \sigma \text{ for each } k \in \{1 \dots p\} \} \\ & \sqcup \left( \{ \sigma \} \cup \{ \sqcup \{ 0, SD_f e_f (n + m) \} * (len - 1) + \sigma \mid k \in \{1 \dots p\} \} \right) \\ = & \quad \{ \text{by properties of } \sqcup, \text{ and because } len \sqsupseteq 1 \} \\ & \sqcup \{ 0, SD_f e_f (n + m) \} * (len - 1) + \sigma \\ = & \quad \{ \text{by definition of } \sigma_0 \} \\ & \sigma_0 \end{aligned}$$

□

## 7.5.6 Correctness in absence of admissibility conditions

Let us summarize what we have achieved so far: given a function definition  $f$ , we have defined three algorithms for computing an initial signature  $(\Delta_0, \mu_0, \sigma_0)$ , which is a correct approximation of the memory needs of  $f$ . Moreover, if we insert this signature into a signature environment  $\Sigma$  and apply the abstract interpretation functions with this updated environment, we get a triple  $(\Delta_1, \mu_1, \sigma_1)$ , whose components can be equal or more precise than the initial  $(\Delta_0, \mu_0, \sigma_0)$ . Hence, the initial signature is reductive w.r.t. the abstract interpretation. In case this new signature is strictly more precise than the original one, we can apply the abstract interpretation again and get another triple  $(\Delta_2, \mu_2, \sigma_2)$  which can be equal or more precise than  $(\Delta_1, \mu_1, \sigma_1)$ , and so on.

A key point to notice is that we have only shown the reductivity property of the initial signature, but not (directly) its correctness, since the latter follows from the reductivity property (Corollary 7.34). However, reductivity only holds under some admissibility conditions on the externally given  $nb$ ,  $nr$ ,

len functions. If one of these admissibility conditions does not hold, the initial signature may not be reductive. In this case we cannot apply the implication  $[\mathbf{RD}] \Rightarrow [\mathbf{CR}]$ , since the antecedent does not hold. However, it can be proven that, even in those cases in which these admissibility conditions do not hold, the initial signature  $(\Delta_0, \mu_0, \sigma_0)$  is still correct. The proof of this fact is rather technical, and its details are left to Appendix B.

Given the above, let us assume that  $(\Delta_0, \mu_0, \sigma_0)$  is correct, but not reductive. If we define,  $\Delta_1 = \mathcal{I}_f(\Delta_0)$ ,  $\mu_1 = \mathcal{M}_{\Delta_0, f}(\mu_0)$  and  $\sigma_1 = \mathcal{S}_f(\sigma_0)$ , we know, by Theorem 7.26, that the new signature  $(\Delta_1, \mu_1, \sigma_1)$  is correct, but not necessarily more precise than  $(\Delta_0, \mu_0, \sigma_0)$ . If  $\Delta_1 \sqsupset \Delta_0$  then we can safely discard  $\Delta_1$ , since our initial approximation  $\Delta_0$  is more precise. It could also be the case that  $\Delta_1$  and  $\Delta_0$  are not comparable, in which case we compute  $\Delta'_1 = \sqcap\{\Delta_1, \Delta_0\}$ , since it is more precise than both  $\Delta_1$  and  $\Delta_0$ , but still correct. The same reasoning applies to the  $\mu_0$  and  $\sigma_0$  components.

Therefore, in absence of the reductivity property we can define the following modified iteration operators as follows:

$$\begin{aligned}\mathcal{I}'_f(\Delta) &= \sqcap\{\Delta, \mathcal{I}_f(\Delta)\} \\ \mathcal{M}'_{\Delta, f}(\mu) &= \sqcap\{\mu, \mathcal{M}_{\Delta, f}(\mu)\} \\ \mathcal{S}'_f(\sigma) &= \sqcap\{\sigma, \mathcal{S}_f(\sigma)\}\end{aligned}$$

for every  $\Delta \in \mathbb{D}$ ,  $\mu \in \mathbb{F}$  and  $\sigma \in \mathbb{F}$ . It is easy to show that, if  $(\Delta_0, \mu_0, \sigma_0)$  is our initial signature,

$$\Delta_0 \in \text{Red}(\mathcal{I}'_f) \quad \wedge \quad \mu_0 \in \text{Red}(\mathcal{M}'_{\Delta_0, f}) \quad \wedge \quad \sigma_0 \in \text{Red}(\mathcal{S}'_f)$$

and, for every  $n \geq 0$ ,  $((\mathcal{I}'_f)^n(\Delta_0), (\mathcal{M}'_{\Delta_0, f})^n(\mu_0), (\mathcal{S}'_f)^n(\sigma_0))$  is a correct signature.

### 7.5.7 Correctness in absence of parameter-decrease conditions.

In addition to the above mentioned admissibility conditions, the reductivity of the initial bounds also depend on the fact that their components are parameter-decreasing, as stated in Definition 7.39. In absence of these conditions, we can ensure neither the reductivity nor the correctness of these initial bounds. Therefore, it is useful to establish some sufficient conditions under which the parameter-decrease property is guaranteed. Given a function definition, the first condition states that if the sizes of the parameters do not increase from the root call to the recursive ones, every abstract heap or space cost function is parameter-decreasing with respect to that definition.

**Proposition 7.53.** *Assume a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . If for every recursive call  $f \bar{a}_i^n @ \bar{r}_j^m$  in its body it holds that  $|a_k| \bar{x}_i^n \sqsubseteq x_k$  for every  $k \in \{1..n\}$ , then every  $\Delta \in \mathbb{D}$  and  $\xi \in \mathbb{F}$  is parameter-decreasing with respect to  $f$ .*

*Proof.* It follows trivially from the fact that the elements of  $\mathbb{D}$  and  $\mathbb{F}$  are monotone functions.  $\square$

It is not usual to obtain function definitions for which this condition does not hold. In particular, those functions having an accumulator parameter whose size increases from the root call to its recursive calls (such as *revAuxD*, in Example 5.2) do not satisfy this condition. Notice, however, that in these kind of functions, the costs do not depend on the sizes of these accumulator functions. Hence we can set out the following weaker sufficient condition:

**Proposition 7.54.** *Assume a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . If for every recursive call  $f \bar{a}_i^n @ \bar{r}_j^m$  in its body it holds that  $|a_k| \bar{x}_i^n \sqsubseteq x_k$  for every  $k \in P \subseteq \{1..n\}$ . For every  $\Delta \in \mathbb{D}$  (resp.  $\xi \in \mathbb{F}$ ), if  $\Delta$  (resp.  $\xi$ ) does not*

depend on the parameters belonging to  $\{1..n\} \setminus P$ , then  $\Delta$  (resp.  $\xi$ ) is parameter-decreasing with respect to  $f$ .

*Proof.* It follows trivially from monotonicity of  $\Delta$  and  $\xi$ . □

This criterion is satisfied by every example function definition in this thesis. In absence of the parameter-decrease property, we can still adapt our *computeDelta*, *computeMu* and *computeSigma* algorithms so as to get correct initial approximations. We can follow an approach similar to that of [4]: the key idea is to compute an invariant  $\Psi$  bounding the feasible sizes of the parameters as a function of the arguments given to the root call, and then maximize the partial components appearing in each algorithm (for instance,  $\Delta_b$  and  $\Delta_r$  in *computeDelta*). If the invariant  $\Psi$  is given by a set of linear constraints, we can use linear programming techniques for this maximization. This adaptation is subject of future work.

## 7.6 Case studies

In this section we apply our space analysis to several examples. Some of the functions have already been introduced in previous chapters. Since, at this moment, our algorithm does not deal with explicit destruction, **case!** expressions are handled by our implementation as non-destructive **case** expressions. The upper bounds to the heap costs of these non-destructive versions are correct approximations to the costs of their destructive counterparts, whereas stack costs are not affected by explicit destruction.

All the algorithms explained in this chapter are currently implemented in *Maple*. The compiler's front-end generates a representation of the abstract syntax tree corresponding to the program being analysed, and *Maple* computes the initial symbolic approximations and perform the necessary simplifications. This system is also used for computing the asymptotic expressions to the obtained bounds.

In order to avoid excessive subscripting, we shall use  $\mathcal{I}$ ,  $\mathcal{M}$  and  $\mathcal{S}$  to denote the iteration operators, without specifying the name of the function being inferred. The latter can be deduced from the context.

**Example 7.55** (Insertion sort). We will compute a bound to the memory needs of *insertD* and *insortD* from Example 3.12 on page 91. Let us start with the *insertD* function. We use  $x$  and  $ys$  to denote its input parameters. We start from the following information regarding its activation tree,

$$\begin{aligned} nb &= \lambda x \ ys.1 \\ nr &= \lambda x \ ys.ys - 1 \\ len &= \lambda x \ ys.ys \end{aligned}$$

and the following size information,

$$|yy| = \lambda x \ ys.ys - 1$$

assuming  $yy$  is the name of the variable with which the recursive call to *insertD* is done. It is easy to see that the admissibility conditions on  $nb$ ,  $nr$  and  $len$  hold, as well as the parameter-decreasingness conditions.

The *computeDelta* algorithm returns the abstract heap  $\Delta_0 = \lambda x \ ys.[\rho_1 \mapsto \xi(x, ys)]$ , where  $\xi$  is defined as follows:

$$\zeta(x, ys) = \sqcup \left\{ \begin{array}{l} \overbrace{(ys - 1) * [ys \geq 2 \rightarrow 1]}^{n_r} + \overbrace{1 * \sqcup\{[ys \geq 1 \rightarrow 2], [ys \geq 2 \rightarrow 0], [ys \geq 2 \rightarrow 2]\}}^{\Delta_r} \\ \underbrace{\sqcup\{[ys \geq 1 \rightarrow 2], [ys \geq 2 \rightarrow 0], [ys \geq 2 \rightarrow 2]\}}_{\Delta_b} \end{array} \right\}$$

This result can be converted into a piecewise function, so as to obtain:

$$\Delta_0 = \lambda x \ ys. \left[ \rho_1 \mapsto \begin{cases} \perp & ys < 1 \\ 2 & 1 \leq ys < 2 \\ ys + 1 & 2 \leq ys \end{cases} \right]$$

An iteration of the abstract interpretation leads to the following result,

$$\Delta_1 \stackrel{\text{def}}{=} \mathcal{I}(\Delta_0) = \lambda x \ ys. \left[ \rho_1 \mapsto \begin{cases} \perp & ys < 1 \\ 2 & 1 \leq ys < 2 \\ 3 & 2 \leq ys < 3 \\ ys + 1 & 3 \leq ys \end{cases} \right]$$

which is equal to  $\Delta_0$ , except in those points belonging to the interval  $[2, 3)$ , in which the latter result is more precise. However,  $\Delta_0$  is an asymptotic fixed point. If we use this abstract heap as an input to the *computeMu* function we get the following initial approximation:

$$\mu_0 = \lambda x \ ys. \begin{cases} \perp & ys < 1 \\ 2 & 1 \leq ys < 2 \\ 4 & 2 \leq ys < 3 \\ ys + 2 & 3 \leq ys \end{cases}$$

By applying the iterator  $\mathcal{M}_{\Delta_0}$  to this result, we obtain,

$$\mu_1 \stackrel{\text{def}}{=} \mathcal{M}_{\Delta_0}(\mu_0) = \lambda x \ ys. \begin{cases} \perp & ys < 1 \\ 2 & 1 \leq ys < 2 \\ 3 & 2 \leq ys < 3 \\ 4 & 3 \leq ys < 4 \\ ys + 1 & 4 \leq ys \end{cases}$$

which is strictly lower than  $\mu_0$ . A second iteration shows that  $\mu_1$  is an asymptotic fixed point of  $\mathcal{M}_{\Delta_0}$ :

$$\mu_2 \stackrel{\text{def}}{=} \mathcal{M}_{\Delta_0}^2(\mu_0) = \lambda x \text{ } ys. \begin{cases} \perp & ys < 1 \\ 2 & 1 \leq ys < 2 \\ 3 & 2 \leq ys < 3 \\ 4 & 3 \leq ys < 4 \\ 5 & 4 \leq ys < 5 \\ ys + 1 & 5 \leq ys \end{cases}$$

With respect to the stack costs, we obtain

$$\sigma_0 = \lambda x \text{ } ys. \begin{cases} \perp & ys < 1 \\ 9ys - 5 & 1 \leq ys < 2 \\ 9ys & 2 \leq ys \end{cases}$$

which is an asymptotic fixed point, since an application of the iterator  $\mathcal{S}$  leads to the following function:

$$\sigma_1 \stackrel{\text{def}}{=} \mathcal{S}(\sigma_0) = \begin{cases} \perp & ys < 1 \\ 9ys - 5 & 1 \leq ys < 3 \\ 9ys & 3 \leq ys \end{cases}$$

Notice that the iteration of  $\mathcal{S}$  never reaches its actual fixed point, which is  $\lambda x \text{ } ys.[ys \geq 1 \rightarrow [9ys - 5]]$ .

With respect to the *inssortD* definition, and given the following functions:

$$\begin{aligned} nb &= \lambda xs.1 \\ nr &= \lambda xs.xs - 1 \\ len &= \lambda xs.xs \end{aligned}$$

The abstract heap  $\Delta_0$  returned by *computeDelta* is the following:

$$\Delta_0 \approx \lambda xs.[\rho_2 \mapsto xs^2 - xs + 1]$$

The successive iterations of the abstract interpretation lead to the following results,

$$\begin{aligned} \Delta_1 &= \mathcal{I}(\Delta_0) \approx \lambda xs.[\rho_2 \mapsto xs^2 - 2xs + 3] \\ \Delta_2 &= \mathcal{I}^2(\Delta_0) \approx \lambda xs.[\rho_2 \mapsto xs^2 - 3xs + 6] \\ \Delta_3 &= \mathcal{I}^3(\Delta_0) \approx \lambda xs.[\rho_2 \mapsto xs^2 - 4xs + 10] \\ \Delta_4 &= \mathcal{I}^4(\Delta_0) \approx \lambda xs.[\rho_2 \mapsto xs^2 - 5xs + 15] \end{aligned}$$

which are depicted in Figure 7.18. None of them is an asymptotic fixed point.

Assume we give  $\Delta_4$  as an input to the *computeMu* algorithm. We get,

$$\mu_0 \approx \lambda xs.xs^2 - 6xs + 21$$

from which, by iterating with  $\mathcal{M}_{\Delta_4}$  we obtain an asymptotically equivalent bound  $\mu_1$ . Finally, an initial

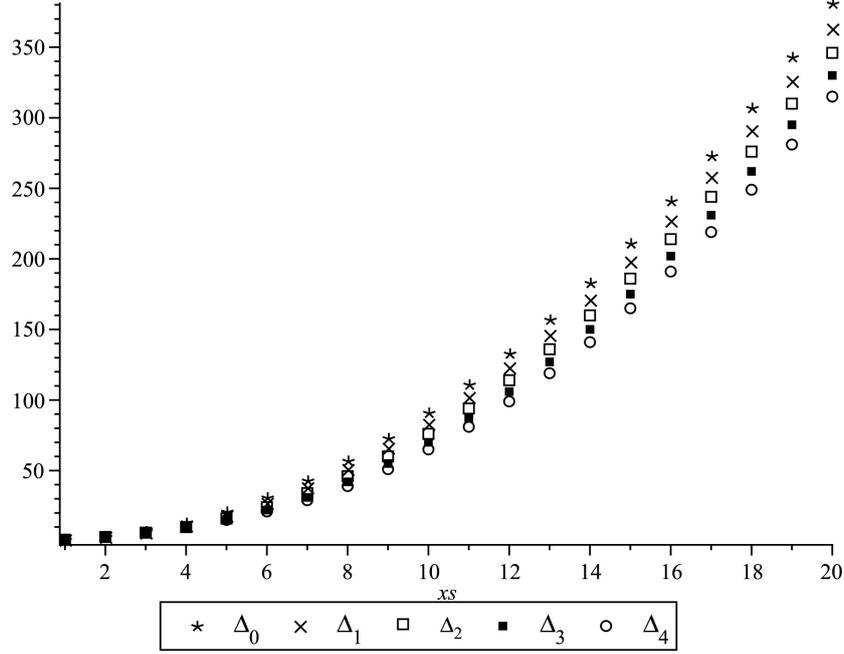


Figure 7.18: Graphical representation of  $\Delta_i xs \rho_2$  resulting from the  $i$ -th iteration.

approximation to the stack costs is given by the following expression:

$$\sigma_0 \approx \lambda xs.15xs - 14$$

and the successive iterations of  $\mathcal{S}$  give place to the following functions.

$$\sigma_1 = \mathcal{S}(\sigma_0) \approx \lambda xs.15xs - 23$$

$$\sigma_2 = \mathcal{S}(\sigma_1) \approx \lambda xs.15xs - 32$$

none of which is an asymptotic fixed point. They are shown in Figure 7.19. □

**Example 7.56.** In Figures 7.20, 7.21 and 7.22 we show the results of our memory consumption analysis for most of the examples appearing in this thesis. We omit the  $\lambda$ -prefixes for a better readability. The values  $\Delta_0$ ,  $\mu_0$  and  $\sigma_0$  represent the initial upper bounds obtained by the *computeDelta*, *computeMu* and *computeSigma* functions. In case these bounds are not fixed points of their corresponding  $\mathcal{I}$ ,  $\mathcal{M}$  and  $\mathcal{S}$ , we represent the results of successive applications of these operators. When  $i > 0$ ,  $\Delta_i$  denotes the abstract heap  $\mathcal{I}^i(\Delta_0)$ . The same applies with the  $\mu_i$  and  $\sigma_i$ . A (\*) mark besides an abstract heap or cost function indicates that it is an asymptotic fixed point of its corresponding operator.

Some functions have been slightly adapted to meet the requirements of the analysis. In particular, polymorphic recursion on regions has been disabled. That is why, in *pascal* function we obtain a quadratic cost function *pascal* instead of linear, as stated in Example 4.20. With regard to the *fib* function, we use a variant of type  $Int \rightarrow HInt$ , instead of the function appearing in Example 4.2, of type  $HInt \rightarrow HInt$ . This is done in order to accommodate the input values to the size model of our analysis. A value of type *HInt* always has size one, whereas the size of a non-negative *Int* is its value.

In some cases, we would get an infinite descending chain  $\mu_0 \sqsupset \mu_1 \sqsupset \dots \sqsupset \mu_i \sqsupset \dots$  of functions, as in the  $\mu$  component of *appendC*. However, this does not mean that the limit of this chain when  $i$  tends to  $+\infty$  is the zero-constant function, since the results shown here are only asymptotic bounds, and the

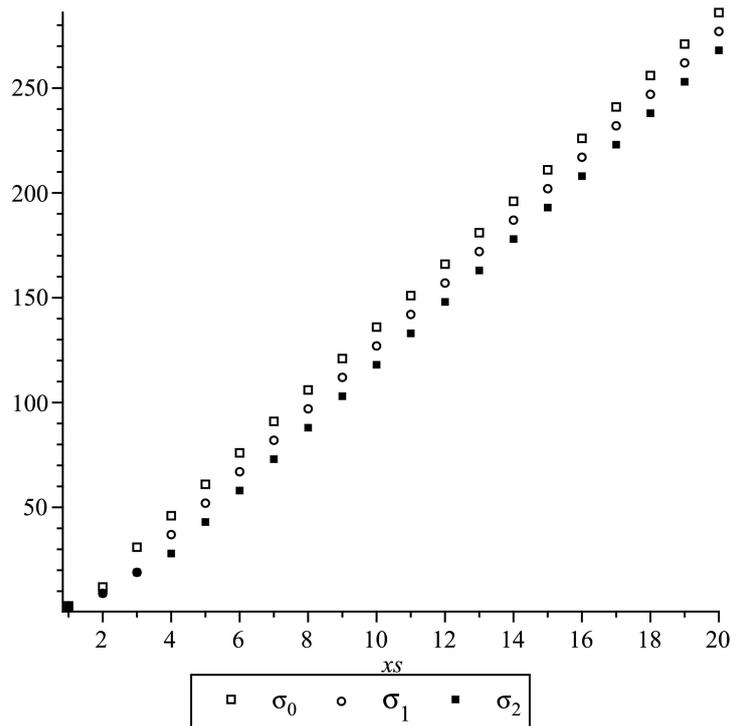


Figure 7.19: Graphical representation of  $\sigma_i xs$  resulting from the  $i$ -th iteration.

limit does not have to coincide asymptotically with the  $\mu_i$ . In the case of *appendC*, the sequence  $\{\mu_i\}_{i \in \mathbb{N}}$  converges pointwise to  $\lambda xs \ ys.xs + ys - 1$ , which is a fixed point of the  $\mathcal{M}$  operator. With the stack costs of *qsort* we get a similar situation.

Normally, the initial bounds are overapproximations of the actual fixed points. In the case of *msort*, whose worst-case space complexity is in  $\mathcal{O}(xs \log xs)$ , we get a quadratic bound. Notice, however, that the  $xs^2$  coefficient keeps decreasing at each iteration.

The *reverse'* function of Figure 7.22 implements a naive algorithm for reversing the elements of a list.

$$\begin{aligned} \text{reverse}' [] &= [] \\ \text{reverse}' (x : xs) &= \text{append} (\text{reverse}' xs) [x] \end{aligned}$$

This algorithm has quadratic heap space complexity, in contrast to the *revAux* function, of linear heap space complexity. These differences become apparent in the results shown in Figure 7.22.  $\square$

Recall from Section 2.8 that the SVM allows us to obtain constant stack costs for tail-recursive functions. Our algorithm *computeSigma* is aware of this, since it can be proven that for every tail-recursive function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$  the result of  $SD_f e_f (n + m)$  is always zero. As a consequence, we get  $\sigma_0 = \sigma$ , where  $\sigma$  is defined as in *computeSigma*. If the latter does not depend on the input sizes, neither does  $\sigma_0$ .

**Example 7.57** (Tail recursive selection sort). Let us consider the implementation of the *selection sort* algorithm shown in Figure 7.23, where every function builds its result in an accumulator parameter. As a consequence, all the functions are tail recursive, and their stack costs are constant (i.e. do not depend on the size of the input). Besides this, the stack upper bounds inferred by *computeSigma* are also

---

*append xs ys @ r* (pg. 24) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto xs - 1]^{(*)} \quad \mu_0 \approx xs - 1^{(*)} \quad \sigma_0 \approx 7xs^{(*)}$$


---

*appendC xs ys @ r* (pg. 26) with  $\Gamma(r) = \rho_3$

$$\Delta_0 \approx [\rho_3 \mapsto xs + ys - 1]^{(*)} \quad \begin{array}{l} \mu_0 \approx 2ys + xs - 2 \\ \mu_1 \approx 2ys + xs - 3 \\ \mu_2 \approx 2ys + xs - 4 \end{array} \quad \sigma_0 \approx 7xs^{(*)}$$


---

*length xs* (pg. 21)

$$\Delta_0 \approx [ ]^{(*)} \quad \mu_0 \approx 0 \quad \sigma_0 \approx 5xs^{(*)}$$


---

*split n xs @ r<sub>1</sub> r<sub>2</sub> r<sub>3</sub>* (pg. 171) with  $\Gamma(r_1) = \rho_1, \Gamma(r_2) = \rho_2, \Gamma(r_3) = \rho_3$

$$\Delta_0 \approx \left[ \begin{array}{l} \rho_1 \mapsto 0 \\ \rho_2 \mapsto \min(n, xs + 1) + 1 \\ \rho_3 \mapsto \min(n, xs + 1) + 1 \end{array} \right]^{(*)} \quad \begin{array}{l} \mu_0 \approx 6 + 2 \min(xs, n - 1) \\ \mu_1 \approx 5 + 2 \min(xs, n - 1)^{(*)} \end{array} \quad \sigma_0 \approx 9 \min(n + 1, xs + 2) + 1^{(*)}$$


---

*merge xs ys @ r* (pg. 171) with  $\Gamma(r) = \rho_1$

$$\Delta_0 \approx [\rho_1 \mapsto 2xs + 2ys - 3] \quad \mu_0 \approx 2xs + 2ys - 3 \quad \sigma_0 \approx 11xs + 11ys - 10$$


---

*msort xs @ r<sub>1</sub> r<sub>2</sub>* (pg. 171) with  $\Gamma(r_1) = \rho_1, \Gamma(r_2) = \rho_2$

$$\Delta_0 \approx \left[ \begin{array}{l} \rho_1 \mapsto \frac{1}{2}xs^2 + \frac{1}{2}xs - 3 \\ \rho_2 \mapsto 2xs^2 - 3xs \end{array} \right] \quad \begin{array}{l} \mu_0 \approx \frac{1}{2}xs \log(xs - 1) + 2xs + \\ \frac{1}{2} \log(xs - 1) + \frac{15}{8}xs^2 - \frac{55}{8} \end{array} \quad \begin{array}{l} \sigma_0 \approx 14 \log(xs - 1) + \\ 11xs + 15 \end{array}$$

$$\Delta_1 \approx \left[ \begin{array}{l} \rho_1 \mapsto \frac{1}{4}xs^2 + \frac{3}{2}xs - \frac{15}{4} \\ \rho_2 \mapsto xs^2 + xs - 3 \end{array} \right] \quad \mu_1 \approx \frac{5}{4}xs^2 + 3xs - \frac{25}{4}^{(*)} \quad \sigma_1 \approx 11xs + 1^{(*)}$$

$$\Delta_2 \approx \left[ \begin{array}{l} \rho_1 \mapsto \frac{1}{8}xs^2 + \frac{9}{4}xs - \frac{35}{8} \\ \rho_2 \mapsto \frac{1}{2}xs^2 + 4xs - \frac{11}{2} \end{array} \right]$$

$$\Delta_3 \approx \left[ \begin{array}{l} \rho_1 \mapsto \frac{1}{16}xs^2 + \frac{23}{8}xs - \frac{79}{16} \\ \rho_2 \mapsto \frac{1}{4}xs^2 + \frac{13}{2}xs - \frac{31}{4} \end{array} \right]$$


---

*partition y xs @ r<sub>2</sub> r<sub>3</sub> r<sub>4</sub>* (pg. 152) with  $\Gamma(r_2) = \rho_2, \Gamma(r_3) = \rho_3, \Gamma(r_4) = \rho_4$

$$\Delta_0 \approx \left[ \begin{array}{l} \rho_2 \mapsto xs \\ \rho_3 \mapsto xs \\ \rho_4 \mapsto xs \end{array} \right]^{(*)} \quad \begin{array}{l} \mu_0 \approx 3xs \\ \mu_1 \approx 3xs - 1^{(*)} \end{array} \quad \sigma_0 \approx 9xs - 2^{(*)}$$


---

Figure 7.20: Results of the space analysis when applied to some example functions on lists.  $(\Delta_0, \mu_0, \sigma_0)$  are the initial bounds, and  $(\Delta_i, \mu_i, \sigma_i)$  denote the result of the  $i$ -th iteration. A  $(*)$  mark indicates that the corresponding bound is an asymptotic fixed point.

---

*qsort xs @ r<sub>1</sub> r<sub>2</sub>* (pg. 129) with  $\Gamma(r_1) = \rho_1$  and  $\Gamma(r_2) = \rho_2$

$$\begin{array}{lll} \Delta_0 \approx \left[ \begin{array}{l} \rho_1 \mapsto 2xs^2 - 4xs + 2 \\ \rho_2 \mapsto xs^2 - xs + 1 \end{array} \right] & \mu_0 \approx 7xs^2 - 37xs + 91 & \sigma_0 \approx 20xs - 9 \\ \Delta_1 \approx \left[ \begin{array}{l} \rho_1 \mapsto 2xs^2 - 6xs + 6 \\ \rho_2 \mapsto xs^2 - 2xs + 3 \end{array} \right] & \mu_1 \approx 7xs^2 - 48xs + 132 & \sigma_1 \approx 20xs - 19 \\ \Delta_2 \approx \left[ \begin{array}{l} \rho_1 \mapsto 2xs^2 - 8xs + 12 \\ \rho_2 \mapsto xs^2 - 3xs + 6 \end{array} \right] & \mu_2 \approx 7xs^2 - 59xs + 184 & \sigma_2 \approx 20xs - 29 \\ \Delta_3 \approx \left[ \begin{array}{l} \rho_1 \mapsto 2xs^2 - 10xs + 20 \\ \rho_2 \mapsto xs^2 - 4xs + 10 \end{array} \right] & \mu_3 \approx 7xs^2 - 70xs + 247 & \sigma_3 \approx 20xs - 39 \end{array}$$


---

*sumList xs @ r* (pg. 153) with  $\Gamma(r) = \rho_2$

$$\begin{array}{lll} \Delta_0 \approx [\rho_2 \mapsto xs]^{(*)} & \mu_0 \approx xs + 1 & \sigma_0 \approx 9xs - 9^{(*)} \\ & \mu_1 \approx xs^{(*)} & \end{array}$$


---

*pascal n @ r* (pg. 153) with  $\Gamma(r) = \rho_1$

$$\begin{array}{lll} \Delta_0 \approx [\rho_1 \mapsto n^2 + 3n + 2] & \mu_0 \approx n^2 - 2n + 17^{(*)} & \sigma_0 \approx 15n + 13 \\ \Delta_1 \approx [\rho_1 \mapsto n^2 + 2n + 3] & & \sigma_1 \approx 15n + 4 \\ \Delta_2 \approx [\rho_1 \mapsto n^2 + n + 5] & & \sigma_2 \approx 15n - 5 \\ \Delta_3 \approx [\rho_1 \mapsto n^2 + 8] & & \sigma_3 \approx 15n - 14 \end{array}$$


---

*combNumbers m n* (pg. 153)

$$\begin{array}{lll} \Delta_0 \approx [ ]^{(*)} & \mu_0 \approx m^2 - m + 12^{(*)} & \sigma_0 \approx 15m + 17^{(*)} \end{array}$$


---

*fib' n @ r* (variation of *fib*, pg. 130) with  $\Gamma(r) = \rho_1$

$$\begin{array}{lll} \Delta_0 \approx [\rho_1 \mapsto 2^n - 1] & \mu_0 \approx 4 \cdot 2^{n-5} + 14 \cdot 2^{n-6} + 18 \cdot 2^{n-7} + 10 \cdot 2^{n-8} + 2 \cdot 2^{n-9} - 3 & \sigma_0 \approx 5n + 7 \\ \Delta_1 \approx [\rho_1 \mapsto 2^{n-1} + 2^{n-2} - 1] & \mu_1 \approx 2 \cdot 2^{n-5} + 6 \cdot 2^{n-6} + 10 \cdot 2^{n-7} + 16 \cdot 2^{n-8} + 18 \cdot 2^{n-9} + 10 \cdot 2^{n-10} + 2 \cdot 2^{n-11} - 4 & \sigma_1 \approx 5n + 6 \\ \Delta_2 \approx [\rho_1 \mapsto 2^{n-1} + 2^{n-4} - 1] & \mu_2 \approx 2 \cdot 2^{n-5} + 8 \cdot 2^{n-6} + 12 \cdot 2^{n-7} + 8 \cdot 2^{n-8} + 2 \cdot 2^{n-9} - 1^{(*)} & \sigma_2 \approx 5n + 5 \\ \Delta_3 \approx [\rho_1 \mapsto 2^{n-3} + 3 \cdot 2^{n-4} + 3 \cdot 2^{n-5} + 2^{n-6} - 1] & & \sigma_3 \approx 5n + 4 \end{array}$$


---

Figure 7.21: Results of the space analysis when applied to some example numeric functions and functions on lists.  $(\Delta_0, \mu_0, \sigma_0)$  are the initial bounds, and  $(\Delta_i, \mu_i, \sigma_i)$  denote the result of the  $i$ -th iteration. A  $(\star)$  mark indicates that the corresponding bound is an asymptotic fixed point.

---

*unshuffle xs @ r<sub>2</sub> r<sub>3</sub>* (pg. 205) with  $\Gamma(r_2) = \rho_2$  and  $\Gamma(r_3) = \rho_3$

$$\Delta_0 \approx \left[ \begin{array}{l} \rho_2 \mapsto xs + 1 \\ \rho_3 \mapsto xs \end{array} \right]^{(*)} \quad \mu_0 \approx 2xs + 2 \quad \sigma_0 \approx 7xs + 2^{(*)}$$

$$\mu_1 \approx 2xs + 1^{(*)}$$


---

*reverse' xs @ r* (variation of *reverse* on pg. 165) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto xs^2 - xs + 1] \quad \mu_0 \approx xs^2 - 5xs + 15^{(*)} \quad \sigma_0 \approx 13xs - 12$$

$$\Delta_1 \approx [\rho_2 \mapsto xs^2 - 2xs + 3] \quad \sigma_1 \approx 13xs - 19$$

$$\Delta_2 \approx [\rho_2 \mapsto xs^2 - 2xs + 6] \quad \sigma_2 \approx 13xs - 26$$

$$\Delta_3 \approx [\rho_2 \mapsto xs^2 - 2xs + 10] \quad \sigma_3 \approx 13xs - 33$$


---

*revAux xs ys @ r* (pg. 165) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto xs - 1]^{(*)} \quad \mu_0 \approx xs^{(*)} \quad \sigma_0 \approx 6^{(*)}$$


---

*reverse xs @ r* (pg. 165) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto xs]^{(*)} \quad \mu_0 \approx xs + 1^{(*)} \quad \sigma_0 \approx 7^{(*)}$$


---

*insertT x t @ r* (pg. 156) with  $\Gamma(r) = \rho_1$

$$\Delta_0 \approx \left[ \rho_1 \mapsto \frac{1}{2}t + \frac{5}{2} \right]^{(*)} \quad \mu_0 \approx \frac{1}{2}t + \frac{9}{2} \quad \sigma_0 \approx \frac{11}{2}t + \frac{11}{2}^{(*)}$$

$$\mu_1 \approx \frac{1}{2}t + \frac{7}{2}$$

$$\mu_2 \approx \frac{1}{2}t + \frac{5}{2}^{(*)}$$


---

*mkTree xs @ r* (pg. 156) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto xs^2] \quad \mu_0 \approx xs^2 - 4xs + 14^{(*)} \quad \sigma_0 \approx 17xs - 16$$

$$\Delta_1 \approx [\rho_2 \mapsto xs^2 - xs + 2] \quad \sigma_1 \approx 17xs - 27$$

$$\Delta_2 \approx [\rho_2 \mapsto xs^2 - 2xs + 5] \quad \sigma_2 \approx 17xs - 38$$

$$\Delta_3 \approx [\rho_2 \mapsto xs^2 - 3xs + 9] \quad \sigma_3 \approx 17xs - 49$$


---

*inorder t @ r* (pg. 156) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto t^2 - \frac{7}{2}t + \frac{7}{2}] \quad \mu_0 \approx 2t^2 - 22t + 79 \quad \sigma_0 \approx \frac{39}{2}t - \frac{79}{2}$$

$$\Delta_1 \approx [\rho_2 \mapsto t^2 - \frac{11}{2}t + \frac{19}{2}] \quad \mu_1 \approx 2t^2 - 30t + 131 \quad \sigma_1 \approx \frac{39}{2}t - \frac{143}{2}$$

$$\Delta_2 \approx [\rho_2 \mapsto t^2 - \frac{15}{2}t + \frac{39}{2}] \quad \mu_2 \approx 2t^2 - 38t + 199 \quad \sigma_2 \approx \frac{39}{2}t - \frac{207}{2}$$

$$\Delta_3 \approx [\rho_2 \mapsto t^2 - \frac{19}{2}t + \frac{67}{2}] \quad \mu_3 \approx 2t^2 - 46t + 283 \quad \sigma_3 \approx \frac{39}{2}t - \frac{271}{2}$$


---

*treesort xs @ r* (pg. 156) with  $\Gamma(r) = \rho_2$

$$\Delta_0 \approx [\rho_2 \mapsto 4xs^2 - 19xs + 28]^{(*)} \quad \mu_0 \approx 9xs^2 - 103xs + 340^{(*)} \quad \sigma_0 \approx 39xs - 123^{(*)}$$


---

Figure 7.22: Results of the space analysis when applied to some example functions on lists and trees.  $(\Delta_0, \mu_0, \sigma_0)$  are the initial bounds, and  $(\Delta_i, \mu_i, \sigma_i)$  denote the result of the  $i$ -th iteration. A  $(*)$  mark indicates that the corresponding bound is an asymptotic fixed point.

```

min :: Int → Int → Int
min x y | x ≤ y = x
        | x > y = y

minimumAc :: Int → [Int] → Int
minimumAc ac [] = ac
minimumAc ac (x : xs) = minimumAc (ac 'min' x) xs

removeAc :: Int → [Int] → [Int] → [Int]
removeAc x [] ac = ac
removeAc x (y : ys) ac
  | x == y = revAux ys ac
  | x / = y = removeAc x ys (y : ac)

selectSortAc :: [Int] → [Int] → [Int]
selectSortAc [] ac = ac
selectSortAc xs ac = selectOrdAc (removeAc y xs []) (y : ac)
  where y = minimumAc (-∞) xs

```

Figure 7.23: Tail-recursive selection sort algorithm

constant. In particular, we obtain:

$$\begin{aligned}
\sigma_{min} &\approx \lambda x y.5 \\
\sigma_{minimumAc} &\approx \lambda ac xs.11 \\
\sigma_{removeAc} &\approx \lambda x ys ac.9 \\
\sigma_{selectSortAc} &\approx \lambda xs ac.17
\end{aligned}$$

□

## 7.7 Inference in presence of explicit destruction and polymorphic recursion

The techniques shown in this chapter yield correct upper bounds provided the input function does not have destructive pattern matching, and assuming the absence of region-polymorphic recursion. In this section we briefly sketch how to improve the algorithms of this chapter in order to deal with these language facilities. Notice, however, that none of these extensions is implemented nor formally specified. They are just given here as a proof of the feasibility of these techniques for inferring full-fledged *Safe* definitions.

If we consider region-polymorphic recursive definitions the abstract interpretation described in Section 7.3 would require no changes, since its proof of correctness does not distinguish between recursive and non-recursive function applications. Both of these are handled exactly in the same way. Polymorphic recursion does affect the computation of the initial  $\Delta_0$  and  $\mu_0$  explained in Section 7.5. Assume a function definition with  $m$  region parameters, and that the type of the  $i$ -th region parameter is  $\rho_i$ , for each  $i \in \{1..m\}$ . The *computeDelta* algorithm assumes that this mapping between region parameters and RTVs remains constant through the subsequent recursive calls, but this is not true in the case of recursive definitions: the  $i$ -th region parameter may be mapped to a different  $\rho_j$  ( $j \neq i$ ) in some recursive calls, or it could be mapped to  $\rho_{self}$ . As a consequence, we have a finite number  $\Gamma_1, \dots, \Gamma_n$  of typing environments typing the subsequent recursive calls. For every recursive call, the correspondence between

region variables and RTVs is given by one of these environments. Assume we are able to find some  $nr_{(i)}$  ( $i \in \{1..n\}$ ) such that:

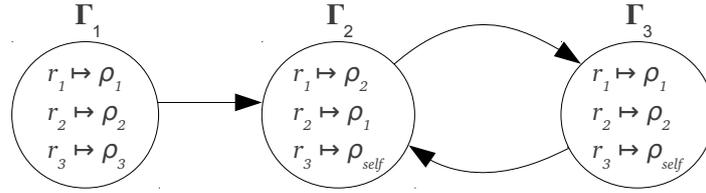
$$nr \sqsubseteq nr_{(1)} + nr_{(2)} + \dots + nr_{(n)}$$

and each  $nr_{(i)}$  bounds the number of recursive calls associated with the environment  $\Gamma_i$ . Then we would be able to consider each  $nr_{(i)}$  separately, multiplied by the charges done by the recursive subsequences under the environment  $\Gamma_i$ , in the style of the following example:

**Example 7.58.** Given the following region-annotated definition:

$$\begin{aligned} f &:: Int \rightarrow \rho_1 \rightarrow \rho_2 \rightarrow \rho_3 \rightarrow ([Int]@_{\rho_1}, [Int]@_{\rho_2})@_{\rho_3} \\ f \ 0 @ r_1 r_2 r_3 &= ([ ]@_{r_1}, [ ]@_{r_2})@_{r_3} \\ f \ n @ r_1 r_2 r_3 &= (n : xs, n : ys)@_{r_3} \\ &\textbf{where } (xs, ys) = f \ (n - 1) @ r_2 r_1 self \end{aligned}$$

Assume  $\Gamma = [r_1 : \rho_1, r_2 : \rho_2, r_3 : \rho_3]$ . In the first recursive call, the type of first parameter becomes  $\rho_2$  and the type of the second one becomes  $\rho_1$ , whereas no region is mapped to  $\rho_3$ . In the second recursive call, the type of the first parameter is  $\rho_1$  again, and the type of the second parameter is  $\rho_2$ . In the third recursive call we have the same mapping as in the first one. These changes in the mappings between parameter positions and RTVs can be depicted as follows:



If  $nr = \lambda n.n$  is an upper bound to the number of recursive calls, one of these calls is done with the typing environment  $\Gamma_1$  above, at most  $\lceil \frac{n-1}{2} \rceil$  of these calls correspond to  $\Gamma_2$ , and at most  $\lceil \frac{n-1}{2} \rceil$  are done with  $\Gamma_3$ . Therefore, we have three different functions for modeling the number of recursive calls:

$$nr_{(1)} = \lambda n.n \quad nr_{(2)} = \lambda n. \left\lceil \frac{n-1}{2} \right\rceil \quad nr_{(3)} = \lambda n. \left\lceil \frac{n-1}{2} \right\rceil$$

By applying the abstract interpretation rules with the recursive part of the expression assuming each  $\Gamma_i$ , we obtain the following abstract heaps:

$$\Delta_{(1)}^r = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \\ \rho_3 & \mapsto & 1 \end{bmatrix} \quad \Delta_{(2)}^r = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \end{bmatrix} \quad \Delta_{(3)}^r = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \end{bmatrix}$$

We proceed in a similar way with the base sequences of the function's body:

$$\Delta_{(1)}^b = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \\ \rho_3 & \mapsto & 1 \end{bmatrix} \quad \Delta_{(2)}^b = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \end{bmatrix} \quad \Delta_{(3)}^b = \begin{bmatrix} \rho_1 & \mapsto & 1 \\ \rho_2 & \mapsto & 1 \end{bmatrix}$$

Hence we get:

$$\begin{aligned}
\Delta_0 n \rho_1 &= nr_{(1)} n * \Delta_{(1)}^r n \rho_1 + nr_{(2)} n * \Delta_{(2)}^r n \rho_2 + nr_{(3)} n * \Delta_{(3)} n \rho_1 \\
&\quad + nb n * (\Delta_{(1)}^b n \rho_1 \sqcup \Delta_{(2)}^b n \rho_2 \sqcup \Delta_{(3)}^b n \rho_1) \\
&= 1 + \left\lceil \frac{n-1}{2} \right\rceil + \left\lceil \frac{n-1}{2} \right\rceil + 1 \\
\Delta_0 n \rho_2 &= nr_{(1)} n * \Delta_{(1)}^r n \rho_2 + nr_{(2)} n * \Delta_{(2)}^r n \rho_1 + nr_{(3)} n * \Delta_{(3)} n \rho_2 \\
&\quad + nb n * (\Delta_{(1)}^b n \rho_2 \sqcup \Delta_{(2)}^b n \rho_1 \sqcup \Delta_{(3)}^b n \rho_2) \\
&= 1 + \left\lceil \frac{n-1}{2} \right\rceil + \left\lceil \frac{n-1}{2} \right\rceil + 1 \\
\Delta_0 n \rho_3 &= nr_{(1)} n * \Delta_{(1)}^r n \rho_3 + nb n * \Delta_{(1)}^b n \rho_2 \\
&= 1
\end{aligned}$$

□

With regard to destructive pattern matching facility, we need to know which variable is affected by a given **case!**. This variable, which appears explicitly in the discriminant of the **case!**, is lost after flattening. So, our first modification is to keep these variables somewhere in the sequences. Assume we adapt our *seqs* function such that it returns sequences of the form  $[G \rightarrow be_1, \dots, be_n \mid D]$ , where  $D$  is a set containing the variables occurring in the discriminant of a destructive pattern matching. The  $\llbracket \cdot \rrbracket_\Delta$  interpretation of sequences would be extended as follows:

$$\llbracket [G \rightarrow be_1, \dots, be_n \mid D] \rrbracket_\Delta \Sigma \Gamma = \left[ G \rightarrow \sqcup \left\{ \llbracket [] \rrbracket_f, \sum_{i=1}^n (\llbracket be_i \rrbracket_\Delta \Sigma \Gamma) + \sum_{x \in D} [region(x) \mapsto -1] \right\} \right]$$

where  $region(x)$  denotes the outermost RTV of the type of  $x$ . This information can be obtained from the typing environment<sup>4</sup>. The least upper bound with the empty abstract heap is done to avoid overall negative charges. With this new definition it can be proven that the  $\llbracket \cdot \rrbracket_\Delta$  interpretation is correct, and yields a function in  $\mathbb{ID}$ . However, the *computeDelta* algorithm is more problematic: it could return an initial bound  $\Delta_0$  not belonging to  $\mathbb{ID}$ , since the monotonicity property might not hold. However, in those examples where  $\Delta_b$  and  $\Delta_r$  (being these as defined in *computeDelta*) belong to  $\mathbb{ID}$ , so does the resulting  $\Delta_0$ . The same applies to  $\mu_0$  and  $\sigma_0$ : if their respective components belong to  $\mathbb{IF}$ , so do the results  $\mu_0$  and  $\sigma_0$ .

**Example 7.59.** Assume the *revAuxD* function of Example 5.2.

$$\begin{aligned}
revAuxD &:: [\alpha]@p_1 \rightarrow [\alpha]@p_2 \rightarrow p_2 \rightarrow [\alpha]@p_2 \\
revAuxD \ xs \ ys \ @ \ r &= \mathbf{case!} \ xs \ \mathbf{of} \\
&\quad [] \rightarrow ys \\
&\quad (x : xx) \rightarrow \mathbf{let} \ x_1 = (x : ys)@r \ \mathbf{in} \ revAuxD \ xx \ x_1 \ @ \ r
\end{aligned}$$

We get the following sequences:

$$\begin{aligned}
seq_1 &= [xs \geq 1 \rightarrow ys \mid xs] \\
seq_2 &= [xs \geq 2 \rightarrow (x : ys)@r, revAuxD \ xx \ x_1 \mid xs]
\end{aligned}$$

<sup>4</sup>It is worth noting that, with this new definition, the set  $R_f$  occurring in the definition of  $\mathbb{ID}$  would have to be extended with the RTVs occurring in the types of the input parameters, since there could be negative charges in those RTVs.

The first one is a base sequence, from which we get following abstract heap:

$$\lfloor \Delta_b \rfloor = \lambda xs \ ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 0]]$$

The second sequence is recursive, and by assuming  $\Sigma(\text{revAuxD}) = ([\ ]_f, 0, 0)$  we obtain:

$$\lfloor \Delta_r \rfloor = \lambda xs \ ys. [xs \geq 2 \rightarrow [\rho_2 \mapsto 0]]$$

Both of these heaps belong to  $\mathbb{F}$ . Since  $nr = \lambda xs \ ys. xs - 1$  and  $nb = \lambda xs \ ys = 1$ , the *computeDelta* algorithm yields the following result:

$$\Delta_0 = \lambda xs \ ys. [xs \geq 1 \rightarrow [\rho_2 \mapsto 0]]$$

□

## 7.8 Conclusions and related work

We have introduced an abstract interpretation-based analysis for computing memory bounds, which takes heap and stack consumption into account. Our approach consists in finding some initial correct approximations, and applying an abstract interpretation function repeatedly in order to increase their accuracy. The strengths of our approach can be summarized as follows:

1. It scales well to large programs, as each *Safe* function can be separately inferred. The relevant information about the called functions is recorded in the signature environment.
2. It supports arbitrary algebraic data types, provided they do not present mutual recursion.
3. We get upper bounds for the maximum amount of *live* memory, as the inference algorithms take into account the deallocation of dead regions made at function termination.
4. It can accommodate several complexity classes, provided these are monotone with respect to the input sizes.
5. It is, to our knowledge, the first approach in which the upper bounds can be improved just by iterating the inference algorithm.

The latter point should be interpreted cautiously, since a larger amount of iterations implies more accuracy, but also more complexity in the resulting expressions (at least with our current implementation). Notice, however, that we still can get simple *asymptotic* bounds. In those cases where the programmer is only interested in asymptotic bounds, these can be given as input to the next iteration, and the result is also (asymptotically) another asymptotic bound, much simpler to compute than if the iteration is done with a non-asymptotic initial bound.

Notice that, in some cases, it is possible to get an infinite, strictly decreasing sequence of upper bounds without reaching a fixed point. It is an interesting subject of future work to study the convergence of such sequences.

A weak point that still requires more work is the restriction we have imposed to our functions: they must be non-negative and monotone. That is why destructive pattern matching has been omitted from our analysis in a first phase. In Section 7.7 we have given some sufficient conditions in which cell

deallocation can be taken into account without breaking the monotonicity restriction. Notice, however, that these are not necessary conditions: there are some examples not satisfying them, but still resulting in monotone bounds. For instance, the *inssortD* function of Example 7.55 gives a constant space bound in presence of explicit destruction. In spite of this, restricting the abstract domain to monotone functions still excludes some interesting function definitions, such as those that destroy more memory than they consume, or those whose output size decreases as the input size increases.

The first approaches to space consumption analysis were restricted to infer linear memory bounds. Hughes and Pareto developed in [61] a type system and a type-checking algorithm which guarantees safe memory upper bounds in a region-based first-order functional language. Unfortunately, the approach requires the programmer to provide detailed consumption annotations. The first fully automatic technique is due to Hofmann and Jost. In [58] they present a type system and a type inference algorithm which, in case of success, guarantees linear heap upper bounds for a first-order functional language, and it does not require programmer annotations. This type system has been extended in [64] so as to support higher-order functions.

Beyond linear bounds, the pioneer research on memory consumption is that carried out under the AHA project (*Amortised analysis of Heap space Usage*) [43], aimed at inferring amortised costs for heap space. In [105], Shkaravska et al. introduce a variant of sized types, in which the size annotations can be polynomials of any degree. They address two novel problems: polynomials are not necessarily monotonic and they are *exact* bounds, as opposed to approximate upper bounds. These bounds are inferred with a combination of testing and polynomial interpolation-based techniques. In [107] they extend their work to give approximate upper bounds on the output sizes, thus broadening the class of analysable programs. In this case, the size relations are expressed via non-deterministic conditional rewriting systems, from which a closed form is extracted by using polynomial interpolation. A strength of this approach is that, since the inference is testing-based, the function being inferred can be considered as a black box. This allows them to use the same inference techniques in different applications, such as, for example, inferring loop-bounds for Java programs [106]. Unfortunately, polynomial interpolation-based techniques do not necessarily lead to a correct upper bound, and some external mechanism is needed for checking that the result of the analysis is sound. Our analysis always gives correct bounds if the externally given call-tree and size information is correct.

The COSTA system (*COST and Termination Analyzer for Java Bytecode*) [5, 6] implements a fully mechanical approach to resource analysis for Java bytecode programs. It is based on the classical method of Wegbreit [119]. It consists in the generation of a recurrence relation which captures the cost of the program being analysed, and the computation of a closed form by using a built-in recurrence solver PUBS [3, 4]. Their results go far beyond linear bounds: the system can infer polynomial, logarithmic, and exponential bounds. COSTA also allows a restricted form of non-monotonicity, provided it occurs in the context of linear expressions. The computation of our initial  $\Delta_0$  shares some similarities with the way in which PUBS solves recurrence relations. The main difference is that PUBS computes the *nb* and *nr* functions giving the number of calls in a recurrence, whereas these functions are given externally in our system. However, PUBS' approach of computing both *nr* and *nb* from the *len* function may yield imprecise over-approximations in some cases (e.g. Quicksort). A drawback, in comparison to our system, is that COSTA does not support non-linear size relations, even if these were given externally. The reason behind this is that these relations are the guards of the recurrence being generated, and PUBS assumes these guards to be conjunctions of linear constraints.

More recently, the COSTA team has extended their system in order to deal with different models of garbage collection [7]. The new results are very promising. In their work, they claim that their liveness-

based model can accommodate the region-based memory management approach of [27], although this integration is neither described nor formally specified.

Another promising technique for inferring polynomial bounds is due to Hoffmann and Hofmann [56], which extends the work of [58]. In the univariate case, their system is able to infer bounds, expressed as non-negative linear combinations of binomial coefficients. These combinations subsume the class of polynomials with non-negative coefficients, while allowing some polynomials with negative coefficients, such as  $x^2 - x$ . However, it does not cover some other natural-valued polynomial bounds, such as  $3x^2 - 6x + 7$ . In a more recent work [55], Hoffmann et al. extend their analysis to the inference of multivariate functions. Unlike our system, they do not handle regions nor explicit destruction, although they claim that the latter can be added with no difficulty. Another drawback is that, to our knowledge, their approach does not deal with arbitrary data types.

Both the original type system of [58], and its extension to polynomial bounds [56] are closely related to the potential method used in the context of amortised analysis [110, 32]. This approach provides an implicit notion of *input size* (the number of elements in a list, usually), but there is no explicit dependence between input sizes and costs, since the latter are given by the potential assigned to each element of the input and output DSs. On the contrary, our space analysis, as well as the COSTA system, are both based on explicit sizes and symbolic manipulation. It is arguable whether one is better than the other. An advantage of amortised analysis is that it allows to express more precisely the memory costs, when they do not depend exclusively on the input size. Moreover, amortised analysis yields more precise results when considering several functions executed in sequence, since it accounts for the overall costs as a whole, instead of just adding the worst-case costs of each function separately. On the contrary, the approaches based on symbolic manipulation can be extended more easily, for instance, by including cost expressions from several complexity classes, as it is done in *Safe*.

## Chapter 8

# Conclusions and future work

### 8.1 Conclusions

In this thesis we have designed and implemented several analyses for proving pointer-safety and bounded memory consumption of programs written in a first-order functional language combining region-based memory management with explicit deallocation. Let us summarize the goals of this work, and assess the extent to which they have been achieved.

1. **Develop an efficient, fully automatic, type-based, static analysis for ensuring pointer safety** [98, 84, 85, 81].

We have defined a type system which attaches region type variables and marks to standard Hindley-Milner algebraic types. Region type variables reflect all the regions in which a DS may live. They allow the compiler to track the DSs that will be located in the working region at runtime, and ensure that neither the input parameters nor the result of the function are located in this region. Marks indicate whether a given DS is destroyed via destructive pattern matching. An auxiliary analysis is needed for approximating the sharing relations between the different variables. Those relations determine which variables may get corrupted when some of them are disposed of.

We have provided an inference algorithm for this destruction-aware type system. It consists of two separate parts:

- (a) A region inference algorithm, which annotates the program with region variables, and infers the region type variables in the typing derivation. The fact that our language is first-order, and that our memory model is simpler than in [116], makes our inference algorithm simpler and more efficient than Tofte and Birkedal's algorithm [112].
- (b) A mark inference algorithm, which annotates the types with marks, and ensures that destructive pattern matching is done in a safe way.

We have tested the inference algorithms with several case studies. A number of algorithms manipulating data structures with constant heap space cost have been successfully typed by the algorithm. In those which have not, we have identified the auxiliary sharing analysis as the main source of accuracy loss.

The type system has been proven correct with respect to the semantics of *Safe*, whereas the inference algorithms have been shown to be correct with respect to the type system. In addition, the

mark inference algorithm is complete with respect to the latter.

## 2. Certify pointer-safety in an automatic way [36].

We have shown how the compiler generates an Isabelle/HOL script establishing that the program being certified is well-typed. Together with a set of previously proved theorems guaranteeing pointer-safety of well-typed programs, this certifies that the object program is pointer-safe.

## 3. Develop a memory cost model for *Safe* programs [82].

From the big-step operational semantics of the language, and in a series of successive refinements, we have derived an imperative abstract machine (called SVM) in which *Safe* programs are run. The translation between *Safe* and the instructions of the SVM has been formalized. This translation has allowed us to set up a connection between each *Safe* expression and its heap and stack memory consumption in the SVM. This connection has been formalised by means of some resource annotations, which have been attached to the big-step operational semantics. These annotations make up a cost model for *Safe*. Finally, these annotations have been proved correct with respect to the actual consumption of *Safe* programs in the SVM.

## 4. Develop an abstract interpretation-based analysis for inferring heap and stack memory bounds [83].

We have developed an analysis which, given a *Safe* function, returns an expression bounding its stack and heap memory consumption as a function on the sizes of the input parameters. Our abstract domain is the set of monotone functions with the usual  $\sqsubseteq$  ordering.

As a first step, we have defined an abstract interpretation function describing the heap and stack needs of each expression. This interpretation has been proved correct with respect to the cost model developed previously. After this, we have introduced algorithms addressing the inference problem for recursive functions. We have proved that the result of these algorithms is reductive under certain conditions. This implies that we can iterate the abstract interpretation in order to improve the results of these algorithms with each iteration.

In a first stage we have considered the inference of recursive bounds in absence of polymorphic recursion and explicit destruction, but we have shown the feasibility of the integration of these features in our inference algorithms, provided that the result belongs to the abstract domain.

The main advantages of our approach are its flexibility for including functions from a broad set of complexity classes, in spite that, at this moment, non-monotonic functions are not considered. A remarkable feature is that the obtained upper bounds can be improved by just iterating the inference algorithm.

## 8.2 Future work

In this section we describe some possible extensions and improvements that are subject of future work. Regarding our destruction-aware type system, we have the following issues to address:

- **Improved sharing analysis.**

Some of the case studies of Section 5.3 show the need for a more accurate sharing analysis than the one currently implemented [98], so that the programmer does not have to adjust manually its results in these cases. Our current analysis distinguishes between the recursive and non-recursive

part of a data structure, and keeps track of those variables that may point to each of these parts. However, when approximating the sharing relations in the context of function applications, this distinction has to be lost, in order to keep the analysis sound. A possibility that may increase the accuracy of our analysis is to give more information on which particular descendant of a data structure is being shared by a variable.

- **Principal types.**

In Section 4.7 we have proved the absence of principal types in the  $\vdash_{Reg}$  rules, when considering some pathological cases, such as the function that copies its parameter and returns it as a result. The reason behind this is that the copy operator must be applied to a variable with an *algebraic* type, but it can be *any* algebraic type. The syntax of our types allows us to specify concrete algebraic types, such as  $[\alpha]@_{\rho_1}$  or *Tree Int* @  $\rho_2$ , and also polymorphic type variables, which may be instantiated to any type (basic or algebraic). However, there is no mechanism in our type syntax for denoting an arbitrary algebraic data structure whose outermost variable is fixed. A solution to this problem involves adding a new sort of types that allows to specify such algebraic structures. It is subject of short-term future work to integrate this extension in the type system.

- **Mutually recursive data types.**

The current implementation of *Safe* does not allow the specification of mutually recursive data types. Extending the type system with this broader class of types would have major implications in the sharing analysis, but also in the semantics of the language itself. In absence of mutual recursion, we can easily refer to the recursive spine of a DS just by considering the recursive positions of the data constructors that make up that spine. As a consequence, the recursive children of a given cell are located just one pointer away from this cell. This does not hold if two data types are mutually recursive, as there might be several cells staying in the middle of the path between a given cell and its recursive children. Under these circumstances, our notion of recursive positions would become more sophisticated.

- **Data types with nested recursion.**

*Safe* currently allows recursion in data types, but only at the outermost level. Support for nested recursion involves the same implications as having mutually recursive data types. In particular, nested recursion can be translated into a set of mutually recursive data structures.

- **Nested destruction.**

Although our in-danger types are able to track those DS whose non-recursive part may be corrupted, they give very little information on *which* part. For instance, assume a variable  $x$  with type  $[[\alpha]@_{\rho_1}]#_{\rho_2}$ . The fact that  $x$  gets an in-danger type implies that:

- The outermost list pointed to by  $x$  may be destroyed, and/or
- Some of the innermost lists may be destroyed, and/or
- Some element in some innermost list may be destroyed.

By extending the type system with nested condemned types, we can increase the accuracy of the type system in determining the level of nesting to which the DS pointed to by  $x$  can be destroyed. For instance,  $[[\alpha]@_{\rho_1}]!_{\rho_2}$  only allows the destruction of the outermost list, whereas  $[[\alpha]!_{\rho_1}]!_{\rho_2}$  allows the programmer to destroy the innermost lists, while ensuring that none of their elements

of type  $\alpha$  is destroyed. In order to carry out this improvement, we have to increase the accuracy of our auxiliary sharing analysis, so that it distinguishes between the different layers of a DS. This is more precise than distinguishing only the outermost recursive spine from the rest of the DS.

With respect to our abstract interpretation-based space consumption analysis, we highlight the following aspects to be improved:

- **Support for different kinds of resources.**

In this thesis we have focused on bounding memory consumption, but the research field of resource analysis is broader; it addresses more kinds of resources. Most of the ideas shown in Chapter 7 can be applied to different classes of resources. A desirable property of a resource bounds analysis is to have some kind of parametrizability on the resource being measured. As an example, the COSTA system [5] provides support for several fixed notions of resources: number of instructions executed, memory consumption in the presence of different models of garbage collection, and number of calls to a given method.

The work of Aspinall et al. [9] introduces a generic concept of *resource algebras*, providing a general framework for embedding different kinds of resources. A generalization of our abstract interpretation function for arbitrary resource algebras should be straightforward. However, the computation of the initial bounds requires special attention. The algorithm for computing the initial  $\Delta_0$  (see Section 7.5.3) can be applied on those kind of resources which are *compositionally additive*, that is, those resources in which the cost of a compound expression is the sum of the costs of its basic sub-expressions. The algorithm for the computation of the initial  $\sigma_0$  (Section 7.5.5) can be applied on those resources whose consumption may increase and/or decrease along the execution of the program, and we are interested in a maximum consumption level. The algorithm for computing  $\mu_0$  also finds a maximum consumption level, but also entails an additive component. A more general resource framework should set certain conditions on resource algebras for determining which algorithm can be used for computing the initial approximation of the bounds to a given resource consumption.

- **Support for different size models.**

In the same way it is useful to consider different kinds of resources, it is also convenient to consider different kinds of size models. With the term *size* we mean being bound in cost expressions, and that gives an idea on how easy or difficult is to process the input. In this sense, the size of a DS is not necessarily related to the amount of heap memory needed by the DS. For instance, in our system, the size of a natural number is its value, although natural numbers do not take up space in the heap. *Safe* has a fixed notion of size of a DS: the number of cells in the recursive spine. This notion is also fixed in the COSTA system: the size of a DS is the length of the longest reachable chain from its root. The size model may decisively influence the complexity order of the resulting bound. For instance, consider a function for generating a copy of a binary tree. In *Safe* this function is reported to have  $\mathcal{O}(t)$  space complexity, whereas, according to COSTA, it has  $\mathcal{O}(2^t)$  space complexity. The actual difference is the meaning of the  $t$  variable: in *Safe* it stands for the number of nodes in the tree, whereas in COSTA it represents the height of the tree.

A good starting point is the work of Tamalet et al. [109], which presents a sized type system with arbitrary algebraic data types. Sizes are defined by assigning a weight to each data constructor. The amortised analysis-based approach of Jost et al. [64, 63] can also assign different potentials to each constructor.

- **Non-monotonic upper bounds.**

As it was shown in Section 7.2, the abstract domain in our system is the set  $\mathbb{F}$  of monotonic functions. The monotonicity condition is essential in order to ensure the correctness of the abstract interpretation, but it is very restrictive in presence of explicit deallocation, since the latter can produce negative charges in the heap. There are some interesting functions that cannot be considered with this restriction, such as those destroying a data structure without building anything. In these cases we can observe some sort of duality with respect to our current approach. Assume a function that only destroys a cell in each recursive call, and does nothing else. Regarding heap consumption, the worst-case corresponds to the *minimum* number of recursive calls, as, in that case, less cells are destroyed, which implies less free memory. Hence, when analysing a function whose costs are monotonically decreasing in each call, we have to consider *lower bounds* on the size of the call-tree. Those cost functions which are neither monotonically increasing nor decreasing would have to be considered in a piecewise basis.

- **Convergence of a sequence of cost functions.**

Our approach allows us to get a decreasing sequence of upper bounds by iterating the abstract interpretation function. In some case studies we have obtained an infinite strictly decreasing sequence of functions, without reaching a fixed point. In many cases we have been able to graphically represent the behaviour of this sequence, and we have been able to (manually) determine the cost function to which this sequence converges, which turned out to be a fixed point of the abstract interpretation. It would be interesting to study the conditions under which we can ensure the convergence of those sequences, and the resulting function that we would obtain when the number of iterations tends to infinity.

Finally, and with respect to the *Safe* language itself, there are two extensions that are worth considering:

- **Higher-order functions.**

The most challenging (but rewarding, as well) extension which is subject of future work is the inclusion of higher-order functions in *Safe*. This would require a change in almost every aspect of the language: syntax, semantics, virtual machine, region and mark inference, sharing analysis, space consumption analysis, and certification.

The adaptation of some of the already existing analyses to a higher-order language can be carried out with relatively low difficulty. This includes the region inference algorithm, and the space consumption analysis. With respect to the former, the MLKit compiler [116, 113] already supports region inference for higher-order functions. Since we will maintain the decision of having a single region per call, we expect that our algorithm will remain simpler than that appearing in [112]. With respect to the space consumption analysis for higher-order functions, the cost expressions of functional parameters must appear symbolically in the resulting bound. Only when concrete functions are passed as parameters (and their memory costs have been inferred), these symbols can be replaced by particular cost functions. A different approach, based on amortised analysis, is that of [64], which extends the work of [58] in order to include support for higher-order functions.

- **Alternative target languages.**

The current implementation of *Safe* generates Java bytecode as a result. A strong point of the Java virtual machine is that it is available in many software and hardware platforms. However, it is somewhat restrictive with regard to memory management. As a consequence of this, the explicit

destruction of a cell is handled by linking that cell to a free list. This cell is reused by the runtime system when a subsequent allocation takes place. Moreover, the statically typed nature of the JVM makes the process of translation awkward. It is convenient to target the bytecode generation to other virtual machines and compilation frameworks which allow more flexibility in memory management. A notable example of this is the LLVM (*Low Level Virtual Machine*) compilation framework [71], which provides a language-independent instruction set. Programs written in this language can be further translated into machine code.

## Appendix A

# Type constraints solving by unification

The region inference algorithm described in Chapter 4 involves the computation of a solution to a previously generated set of constraints between types. In this appendix we put forward an algorithm for computing this solution. This algorithm is an extension of [12], which, in turn, is based on the original Robinson’s unification algorithm [101]. This algorithm is extended to support *isData* constraints and  $\approx_k$  equations, which are weaker forms of standard unification (see 4.4.1 for details)

We provide here a description of our algorithm as a state machine. A *configuration* is a pair of the form  $(E, \langle \theta, \overline{\varphi}_i^p \rangle)$ , where  $E$  is the set of constraints to be solved and  $\langle \theta, \overline{\varphi}_i^p \rangle$  is the partial solution accumulated from the initial configuration. We assume that the number of elements  $p$  in the list  $\overline{\varphi}_i^p$  is fixed, and the corresponding superscript will be omitted when it is not relevant. We also define a transition relation  $\Longrightarrow$  between configurations, as the minimal relation satisfying the rules in Figure A.1. The reflexive-transitive closure of this relation is denoted by  $\Longrightarrow^*$ .

Given a set  $E$  of constraints to be solved, our algorithm starts from the initial configuration  $E, \langle id, \overline{[]^p} \rangle$  and applies successively the rules in Figure A.1 in order to reach a configuration in which the set of constraints is empty. The pair  $\langle \theta, \overline{\varphi}_i \rangle$  contained within this final configuration is a solution to  $E$ , that is, it satisfies the conditions stated in Definition 4.5. If the state machine gets stuck in a configuration in which the set of constraints to be solved is nonempty, the algorithm returns an error.

The  $\overline{\varphi}_i$  given as a result is a list of functions on **RegType**  $\rightarrow$  **RegType**. However, during the algorithm, some  $\varphi_i$  may map a given RTV to several different RTVs. As a consequence, the  $\overline{\varphi}_i$  occurring in the intermediate states of the  $\Longrightarrow^*$  derivation are *relations* on **RegType**, rather than functions. Hence we shall consider our  $\varphi_i$  as a relation on **RegType**. The notation  $\varphi_i(\rho) \ni \rho'$  specifies that the pair  $(\rho, \rho')$  belongs to  $\varphi_i$ . We extend the notation  $\varphi_k(\rho) \ni \rho'$  to arbitrary types as follows:

$$\begin{aligned}
 \varphi_k(B) & \ni B \\
 \varphi_k(\alpha) & \ni \alpha \\
 \varphi_k(T \overline{s}_i^n @ \overline{\rho}_j^m) & \ni T \overline{s}_i^n @ \overline{\rho}_j^m & \text{if } \forall i \in \{1..n\}. \varphi_k(s_i) \ni s'_i \wedge \forall j \in \{1..m\}. \varphi_k(\rho_j) \ni \rho'_j \\
 \varphi_k(\overline{s}_i^n \rightarrow \overline{\rho}_j^m \rightarrow s) & \ni \overline{s}_i^n \rightarrow \overline{\rho}_j^m \rightarrow s' & \text{if } \forall i \in \{1..n\}. \varphi_k(s_i) \ni s'_i \wedge \forall j \in \{1..m\}. \varphi_k(\rho_j) \ni \rho'_j \\
 & \wedge \varphi_k(s) \ni s'
 \end{aligned}$$

Notice that, also in this case, there may be more than one  $s'$  such that  $\varphi_k(s) \ni s'$ . Our algorithm ensures that the list of relations  $\overline{\varphi}_i$  obtained in the final state is actually a list of functions. Only then we can write  $\varphi_i(\rho) = \rho'$ . Moreover, if  $\varphi_k$  is a function on regions, so will be its extension to arbitrary types.

Name	Equations	Substitutions	Cond.
[ID]	$\{\tau = \tau\} \uplus E,$ $\implies E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[VAR]	$\{\alpha = s\} \uplus E,$ $\implies [\alpha \mapsto s](E),$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle [\alpha \mapsto s] \circ \theta, \overline{\varphi_i^p} \rangle$	(1)
[REG]	$\{\rho_1 = \rho_2\} \uplus E$ $\implies [\rho_1 \mapsto \rho_2](\bigcup_{i=1}^p E_i \cup E)$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle [\rho_1 \mapsto \rho_2] \circ \theta, \overline{[\rho_1 \mapsto \rho_2](\varphi_i)^p} \rangle$	(2)
[FUN]	$\{\overline{s_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s = \overline{s_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s'\} \uplus E,$ $\implies \{\overline{s_i^n}, \overline{\rho_j^m}, s = s'\} \cup E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[CONS]	$\{T \overline{s_i^n} @ \overline{\rho_j^m} = T \overline{s_i^n} @ \overline{\rho_j^m}\} \uplus E,$ $\implies \{\overline{s_i^n}, \overline{\rho_j^m}\} \cup E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[ISD1]	$\{T \overline{s_i^n} @ \overline{\rho_j^m} = \text{isData}(\alpha, \rho)\} \uplus E,$ $\implies \{\alpha = T \overline{s_i^n} @ \overline{\rho_j^{m-1}} \rho', \rho_m = \rho\} \cup E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	(3)
[ISD2]	$\{\alpha = \text{isData}(T \overline{s_i^n} @ \overline{\rho_j^m}, \rho)\} \uplus E,$ $\implies \{\alpha = T \overline{s_i^n} @ \overline{\rho_j^{m-1}} \rho\} \cup E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[ISD3]	$\{T \overline{s_i^n} @ \overline{\rho_j^m} = \text{isData}(T \overline{s_i^n} @ \overline{\rho_j^m}, \rho)\} \uplus E,$ $\implies \{T \overline{s_i^n} @ \overline{\rho_j^{m-1}} \rho = T \overline{s_i^n} @ \overline{\rho_j^m}\} \cup E,$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[QFUN]	$\{\overline{s_i^n} \rightarrow s \approx_k \overline{s_i^n} \rightarrow s'\} \uplus E$ $\implies \{\overline{s_i^n} \approx_k \overline{s_i^n}, s \approx_k s'\} \cup E$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[QCONS]	$\{T \overline{s_i^n} @ \overline{\rho_j^m} \approx_k T \overline{s_i^n} @ \overline{\rho_j^m}\} \uplus E$ $\implies \{\overline{s_i^n} \approx_k \overline{s_i^n}, \overline{\rho_j^m} \approx_k \overline{\rho_j^m}\} \cup E$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	
[QVAR-L]	$\{\alpha \approx_k s\} \uplus E$ $\implies \{s' \approx_k s\} \cup [\alpha \mapsto s'](E)$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle [\alpha \mapsto s'] \circ \theta, \overline{\varphi_i^p} \rangle$	(4)
[QVAR-R]	$\{s \approx_k \alpha\} \uplus E$ $\implies \{s \approx_k s'\} \cup [\alpha \mapsto s'](E)$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle [\alpha \mapsto s'] \circ \theta, \overline{\varphi_i^p} \rangle$	(4)
[QREG1]	$\{\rho_1 \approx_k \rho_2\} \uplus E$ $\implies \{\rho_2 = \rho'\} \cup E$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	(5)
[QREG2]	$\{\rho_1 \approx_k \rho_2\} \uplus E$ $\implies E$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \varphi_1 \cdots \varphi_k \uplus [\rho_1 \mapsto \rho_2] \cdots \varphi_p \rangle$	(6)
[QEND]	$\{\overline{\alpha_i^n} \approx_k \overline{\alpha_i^n}\}$ $\implies \{\alpha_i = \alpha_i^n\}$	$\langle \theta, \overline{\varphi_i^p} \rangle$ $\langle \theta, \overline{\varphi_i^p} \rangle$	

Figure A.1: Rules defining the  $\implies$  relation between configurations. For each rule, the topmost line stands for the initial configuration and the line below it denotes the final configuration. Some rules may only be applied under certain conditions, which are described in Figure A.2.

- (1)  $\alpha \notin \text{fv}(s)$
- (2)  $\rho_1 \not\equiv \rho_2 \wedge \forall i \in \{1..p\}. E_i = \begin{cases} \{\rho = \rho'\} & \text{if } (\rho_1, \rho), (\rho_2, \rho') \in \varphi_i \text{ for some } \rho, \rho' \text{ such that } \rho \not\equiv \rho' \\ \emptyset & \text{otherwise} \end{cases}$
- (3)  $\text{fresh}(\rho')$
- (4)  $s \notin \mathbf{TypeVar}, \alpha \notin \text{fv}(s), s' = \text{freshReg}(s)$
- (5)  $(\rho_1, \rho') \in \varphi_k$
- (6)  $\not\equiv \rho'.(\rho_1, \rho') \in \varphi_k$

Figure A.2: Conditions under which the unification rules of Figure A.1 may be applied.

Since we have generalized the notion of our  $\overline{\varphi}_i^p$  functions to relations, it is necessary to modify the Definition 4.5 as follows:

**Definition A.1.** A pair  $\langle \theta, \overline{\varphi}_i \rangle$  is a solution to a set of constraints  $E$  (denoted  $\langle \theta, \overline{\varphi}_i \rangle \models E$ ) if and only if the following conditions hold:

1. For each  $s_1 = s_2 \in E$ , then  $\theta(s_1) \equiv \theta(s_2)$ . The same applies to equations of the form  $sf_1 = sf_2$ .
2. For each  $s = \text{isData}(s', \rho) \in E$  then  $\theta(s) \equiv T \overline{s}_i^n @ \overline{\rho}_j^m(\theta(\rho))$  and  $\theta(s') \equiv T \overline{s}_i^n @ \overline{\rho}_j^m \rho'$  for some  $\overline{s}_i^n, \overline{\rho}_j^m$  and  $\rho'$ .
3. For each  $s_1 \approx_k s_2$ , then  $\varphi_k(\theta(s_1)) \ni \theta(s_2)$ . The same applies to equations of the form  $sf_1 \approx_k sf_2$ .

If the  $\overline{\varphi}_i$  are functions, this definition is equivalent to Definition 4.5. In order to ensure that the  $\overline{\varphi}_i$  resulting from the algorithm are functions, we have to prove that the following property is an invariant preserved by the  $\implies$  relation:

*If a  $\varphi_k$  relates the same  $\rho$  to different RTVs  $\rho_a, \rho_b$ , the latter will be eventually unified.*

The following definition formalizes these intuitions:

**Definition A.2.** A machine configuration  $E, \langle \theta, \overline{\varphi}_i^p \rangle$  is said to be *good* if, for every  $k \in \{1..p\}$  and  $\rho, \rho_a, \rho_b \in \mathbf{RegType}$  such that  $\varphi_k(\rho) \supseteq \{\rho_a, \rho_b\}$  and  $\rho_a \not\equiv \rho_b$ , then there exists a list of region type variables  $\overline{\rho}_i^n$  (with  $n \geq 2$ ) such that:

$$\rho_a \equiv \rho_1 \wedge \rho_b \equiv \rho_n \wedge \forall j \in \{1..n-1\}. \rho_j = \rho_{j+1} \in E$$

We abbreviate this as follows:

$$\{\rho_a \equiv \rho_1 = \rho_2 = \dots = \rho_n \equiv \rho_b\} \subseteq E$$

where  $\equiv$  denotes syntactic equality, and  $=$  denotes belonging to  $E$ .

The existence of a chain in  $E$  between two different RTVs may be considered as an equivalence relation on  $\mathbf{RegType}$ . Two RTVs belong to the same equivalence class if and only if there exists a chain of equations connecting them in  $E$ . As an example, let us assume the following definition of  $\varphi_k$  for some  $k$ :

$$\varphi_k = \{(\rho_1, \rho_3), (\rho_1, \rho_5), (\rho_1, \rho_8), (\rho_2, \rho_4), (\rho_2, \rho_7)\} \quad (\text{A.1})$$

The configuration  $E, \langle \theta, \overline{\varphi}_i \rangle$  is good if  $E = \{\rho_3 = \rho_5, \rho_5 = \rho_8, \rho_4 = \rho_7\}$ . There exists a chain of equations in  $E$  between  $\rho_3$  and  $\rho_8$ , namely  $\{\rho_3 = \rho_5 = \rho_8\} \in E$ . We have two equivalence classes:  $\{\rho_3, \rho_5, \rho_8\}$  and  $\{\rho_4, \rho_7\}$ .

We now explain the rules of Figure A.1. Rules [ID], [VAR], [FUN] and [CONS] are standard and deserve no explanation. The = rules are symmetric, so the equations of the form  $s = \alpha$  are dealt with by [VAR]. Rule [REG] is applied when unifying two region types. In addition to generating the corresponding substitution, more equations may be generated in order to preserve the goodness property of the initial configuration. For example, given the definition of  $\varphi_k$  in (A.1), let us assume that the equation  $\rho_1 = \rho_2$  is being processed. We apply the substitution  $[\rho_1 \mapsto \rho_2]$  to this relation, and we obtain:

$$[\rho_1 \mapsto \rho_2](\varphi_k) = \{(\rho_2, \rho_3), (\rho_2, \rho_5), (\rho_2, \rho_8), (\rho_2, \rho_4), (\rho_2, \rho_7)\}$$

In order to preserve the invariant, every RTV in the set  $\{\rho_3, \rho_5, \rho_8\}$  has to be unified with the RTVs of the set  $\{\rho_4, \rho_7\}$ , as the goodness property demands. This is done by selecting a representative of each equivalence class and generating an equation between these two representatives. In our example, we could choose  $\rho_3 = \rho_4$ . With this new equation the two equivalence classes would be merged into a single equivalence class. If we denote by  $E'$  the set  $E \cup \{\rho_3 = \rho_4\}$ , we can always find a chain in  $E'$  between two given variables of this class. For instance, between  $\rho_5$  and  $\rho_7$ :  $\{\rho_5 = \rho_3 = \rho_4 = \rho_7\} \subseteq E'$ .

Rules [ISD1], [ISD2], [ISD3] deal with *isData* constraints. These are relaxed forms of standard unification, since they allow different outermost regions in each side of the equation. Notice that there is no rule for processing equations of the form  $\alpha_1 = \text{isData}(\alpha_2, \rho)$ , which are not addressed until one of the type variables is replaced by an algebraic type. If this substitution never takes place, the algorithm fails.

The rest of the rules involve  $\approx$ -equations. In this case, the right-hand side of the equations must be a region instance of the left-hand side. The corresponding  $\varphi_k$  may be updated when processing equations of the form  $\rho_1 \approx_k \rho_2$ . This equation would add the pair  $(\rho_1, \rho_2)$  to  $\varphi_k$  if  $\rho_1$  is not related to any RTV in  $\varphi_k$  (rule [QREG2]). Otherwise we would have to unify the  $\rho_2$  with the RTV already bound to  $\rho_1$ , so that the goodness property is preserved (rule [QREG1]). Finally, the rule [QEND] ensures that the equations of the form  $\alpha_1 \approx_k \alpha_2$  are the last ones to be processed, and they will be processed as standard unification equations ( $\alpha_1 = \alpha_2$ ). By delaying the processing of these equations we achieve more general types, since one of the involved variables may be instantiated to a more concrete type, so that the variable occurring in the other side of the  $\approx_k$ -equation may get the same type with different RTVs. On the contrary, if we unify both variables with an equation of the form  $\alpha_1 = \alpha_2$ , they will be forced to have the same region types, if they are instantiated later.

Now we prove that this property is preserved by the  $\implies$  rules of Figure A.1.

**Lemma A.3.** *Given a transition  $E, \langle \theta, \overline{\varphi_i} \rangle \implies E', \langle \theta', \overline{\varphi'_i} \rangle$ , if the initial configuration is good, then so is the final configuration.*

*Proof.* By case distinction on the  $\implies$  rule applied. Most cases are trivial, except the following ones:

- **Case [ID]**

In this case  $E = E' \uplus \{\rho = \rho\}$  for some  $\rho$  and  $\overline{\varphi_i} = \overline{\varphi'_i}$ . So, if  $\varphi'_k(\rho') \ni \{\rho_a, \rho_b\}$  with  $\rho_a \neq \rho_b$  for some  $\rho'$ , then  $\varphi_k(\rho') \ni \{\rho_a, \rho_b\}$ . Since the initial configuration is good, we have the following chain:

$$\{\rho_a \equiv \rho_1 = \rho_2 = \dots = \rho_{i-1} = \rho_i = \rho_{i+1} = \dots = \rho_n \equiv \rho_b\} \subseteq E$$

Assume that there exists an  $i$  such that  $\rho_i \equiv \rho$  and  $\rho_{i+1} \equiv \rho$ . We can shorten the chain as follows:

$$\{\rho_a \equiv \rho_1 = \rho_2 = \dots = \rho_{i-1} = \rho_{i+1} = \dots = \rho_n \equiv \rho_b\} \subseteq E$$

If we repeat this until we obtain a chain in  $E$  not containing the  $\rho = \rho$  equation, the result is a chain in  $E'$ , which proves the lemma.

- **Case [REG]**

Assume  $k$  such that  $\varphi'_k(\rho') \supseteq \{\rho'_a, \rho'_b\}$ , where  $\rho'_a \neq \rho'_b$ . There must exist some RTVs  $\rho_x, \rho_y, \rho_a, \rho_b$  such that  $\rho_a \neq \rho_b$  and:

$$\begin{aligned} [\rho_1 \mapsto \rho_2](\rho_x) &= \rho' \\ [\rho_1 \mapsto \rho_2](\rho_y) &= \rho' \\ [\rho_1 \mapsto \rho_2](\rho_a) &= \rho'_a \\ [\rho_1 \mapsto \rho_2](\rho_b) &= \rho'_b \end{aligned}$$

and  $\varphi_k(\rho_x) \ni \rho_a, \varphi_k(\rho_y) \ni \rho_b$ . We distinguish cases:

1.  $\rho_x \equiv \rho_y$

Since the initial configuration is good, there exists a chain:

$$\{\rho_a \equiv \rho'_1 = \dots = \rho'_n \equiv \rho_b\} \subseteq E$$

By applying the  $[\rho_1 \mapsto \rho_2]$  substitution to each element, we get:

$$\{[\rho_1 \mapsto \rho_2](\rho_a) \equiv [\rho_1 \mapsto \rho_2](\rho'_1) = \dots = [\rho_1 \mapsto \rho_2](\rho'_n) \equiv [\rho_1 \mapsto \rho_2](\rho_b)\} \subseteq [\rho_1 \mapsto \rho_2](E)$$

which is equivalent to:

$$\{\rho'_a \equiv [\rho_1 \mapsto \rho_2](\rho'_1) = \dots = [\rho_1 \mapsto \rho_2](\rho'_n) \equiv \rho'_b\} \subseteq [\rho_1 \mapsto \rho_2](\bigcup_{i=1}^n E_i \cup E)$$

which proves the required result.

2.  $\rho_x \neq \rho_y$

In this case one of the  $\rho_x, \rho_y$  must be equal to  $\rho_1$ . Otherwise we would get:

$$\rho_x \equiv [\rho_1 \mapsto \rho_2](\rho_x) \equiv \rho' \equiv [\rho_1 \mapsto \rho_2](\rho_y) \equiv \rho_y$$

which contradicts the fact that  $\rho_x$  and  $\rho_y$  are distinct. Without loss of generality, let us assume that  $\rho_x \equiv \rho_1$  (the case  $\rho_x \equiv \rho_2$  is symmetrical). Then  $\rho' \equiv \rho_2$ , and  $\rho_y$  must be either  $\rho_1$  or  $\rho_2$  (otherwise  $\rho' \equiv \rho_2 \neq [\rho_1 \mapsto \rho_2](\rho_y)$ , which leads to a contradiction). Since  $\rho_y$  is distinct from  $\rho_x$  (which is equal to  $\rho_1$ ), the only possibility is  $\rho_y \equiv \rho_2$ . Summarizing, we get:

$$\varphi_k(\rho_x) \ni \rho_a, \text{ which is equivalent to } \varphi_k(\rho_1) \ni \rho_a$$

$$\varphi_k(\rho_y) \ni \rho_b, \text{ which is equivalent to } \varphi_k(\rho_2) \ni \rho_b$$

By one of the premises of the [REG] rule, there exist some  $\rho_c, \rho_d$  with  $\varphi_k(\rho_1) \ni \rho_c$  and  $\varphi_k(\rho_2) \ni \rho_d$  such that the equation  $[\rho_1 \mapsto \rho_2](\rho_c = \rho_d)$  belongs to  $E'$ . Notice that these  $\rho_c, \rho_d$  may or may not be equal to the region types  $\rho_a, \rho_b$ . Thus there are four possibilities:

–  $\rho_c \equiv \rho_a$  and  $\rho_d \equiv \rho_b$ . We build the chain:

$$\{\rho'_a \equiv [\rho_1 \mapsto \rho_2](\rho_a) = [\rho_1 \mapsto \rho_2](\rho_b) \equiv \rho'_b\} \subseteq E'$$

–  $\rho_c \not\equiv \rho_a$  and  $\rho_d \equiv \rho_b$ . In this case  $\varphi_k(\rho_1) \supseteq \{\rho_a, \rho_c\}$ . Since the initial state is good, there exists a chain:

$$\{\rho_a \equiv \rho_1^{(0)} = \dots = \rho_{n_1}^{(0)} \equiv \rho_c\} \subseteq E$$

to which the substitution  $[\rho_1 \mapsto \rho_2]$  can be applied:

$$\{\rho'_a \equiv [\rho_1 \mapsto \rho_2](\rho_1^{(0)}) = \dots = [\rho_1 \mapsto \rho_2](\rho_{n_1}^{(0)}) \equiv [\rho_1 \mapsto \rho_2]\rho_c\} \subseteq [\rho_1 \mapsto \rho_2](E)$$

By using the equation  $[\rho_1 \mapsto \rho_2](\rho_c = \rho_d)$ , we get the required result:

$$\{\rho'_a \equiv [\rho_1 \mapsto \rho_2](\rho_1^{(0)}) = \dots = [\rho_1 \mapsto \rho_2](\rho_c) = [\rho_1 \mapsto \rho_2](\rho_d)\} \subseteq E' \quad (\text{A.2})$$

and the lemma follows since  $[\rho_1 \mapsto \rho_2](\rho_d) = [\rho_1 \mapsto \rho_2](\rho_b) = \rho'_b$ .

–  $\rho_c \equiv \rho_a$  and  $\rho_d \not\equiv \rho_b$ . Similarly to the previous case, we obtain  $\varphi_k(\rho_2) \supseteq \{\rho_b, \rho_d\}$  and the following chain:

$$\{\rho_d \equiv \rho_2^{(1)} = \dots = \rho_{n_2}^{(1)} \equiv \rho_b\} \subseteq E$$

By applying the above mentioned substitution and equation, we get:

$$\{[\rho_1 \mapsto \rho_2](\rho_c) = [\rho_1 \mapsto \rho_2](\rho_d) = \dots = [\rho_1 \mapsto \rho_2](\rho_{n_2}^{(1)}) \equiv \rho'_b\} \subseteq E' \quad (\text{A.3})$$

and the lemma follows since  $[\rho_1 \mapsto \rho_2](\rho_c) = [\rho_1 \mapsto \rho_2](\rho_a) = \rho'_a$ .

–  $\rho_c \not\equiv \rho_a$  and  $\rho_b \not\equiv \rho_d$ . We obtain the desired result by joining the chains (A.2) and (A.3). □

The next step is to prove that a  $\implies^*$  derivation builds incrementally a solution to the system of constraints given in the initial configuration. Before this, we need the following auxiliary results, whose proof is a direct consequence of Definition A.1.

**Lemma A.4.** *The  $\models$  relation satisfies the following properties:*

1.  $\langle \theta, \overline{\varphi_i} \rangle \models \theta'(E) \Leftrightarrow \langle \theta \circ \theta', \overline{\varphi_i} \rangle \models E$
2.  $\langle \theta, \overline{\varphi_i} \rangle \models E \Rightarrow \langle \theta, \overline{\varphi'_i \cup \varphi_i} \rangle \models E$

*Proof.* Both facts follow trivially from the definition of  $\models$ . □

**Lemma A.5.** *If  $E, \langle \theta_0, \overline{\varphi_{i,0}^p} \rangle \implies E', \langle \theta, \overline{\varphi_i^p} \rangle$  then, for some  $\theta'$ :*

1.  $\theta = \theta' \circ \theta_0$
2.  $\theta'(\varphi_{i,0}) \subseteq \varphi_i$  for all  $i \in \{1..p\}$ .

*Proof.* By simple inspection of the  $\implies$  rules. □

**Lemma A.6.** *Given the following transition:*

$$E, \langle \theta_0, \overline{\varphi_{i,0}} \rangle \implies E', \langle \theta \circ \theta_0, \overline{\varphi_i} \rangle$$

If  $\langle \theta', \overline{\varphi'_i} \rangle$  is a solution to  $E'$ , then  $\langle \theta' \circ \theta, \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle$  is a solution to  $E$ .

*Proof.* By case distinction on the  $\implies$  rule applied. We show only the most relevant cases. The remaining ones are either straightforward or similar.

- **Case [QVAR-L]**

Let us assume  $s' = \text{freshReg}(s)$ . By hypothesis,  $\langle \theta', \overline{\varphi'_i} \rangle \models \{s' \approx_k s\} \cup [\alpha \mapsto s'](E)$ . Therefore:

$$\begin{aligned}
& \varphi'_k(\theta'(s')) \ni \theta'(s) \wedge \langle \theta', \overline{\varphi'_i} \rangle \models [\alpha \mapsto s'](E) \\
\Rightarrow & \quad \{ \text{since } \alpha \notin \text{fv}(s) \} \\
& \varphi'_k((\theta' \circ [\alpha \mapsto s'])(\alpha)) \ni (\theta' \circ [\alpha \mapsto s'])(s) \wedge \langle \theta', \overline{\varphi'_i} \rangle \models [\alpha \mapsto s'](E) \\
\Rightarrow & \quad \{ \text{by Lemma A.4 (1)} \} \\
& \varphi'_k((\theta' \circ [\alpha \mapsto s'])(\alpha)) \equiv (\theta' \circ [\alpha \mapsto s'])(s) \wedge \langle \theta' \circ [\alpha \mapsto s'], \overline{\varphi'_i} \rangle \models E \\
\Rightarrow & \quad \{ \text{by definition of } \models \} \\
& \langle \theta' \circ [\alpha \mapsto s'], \overline{\varphi'_i} \rangle \models \{\alpha \approx_k s\} \cup E \\
\Rightarrow & \quad \{ \text{by Lemma A.4 (2)} \} \\
& \langle \theta' \circ [\alpha \mapsto s'], \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle \models \{\alpha \approx_k s\} \cup E
\end{aligned}$$

- **Case [QREG1]**

Again, we start from  $\langle \theta', \varphi'_i \rangle \models \{\rho_2 = \rho'\} \cup E$  and the precondition  $\varphi_k(\rho_1) = \varphi_{k,0}(\rho_1) \ni \rho'$  of the [QREG1] rule.

$$\begin{aligned}
& \theta'(\rho_2) \equiv \theta'(\rho') \wedge \langle \theta', \overline{\varphi'_i} \rangle \models E \wedge \varphi_k(\rho_1) \ni \rho' \\
\Rightarrow & \quad \{ \text{by properties of substitutions} \} \\
& \theta'(\rho_2) \equiv \theta'(\rho') \wedge \langle \theta', \overline{\varphi'_i} \rangle \models E \wedge (\theta'(\varphi_k))(\theta'(\rho_1)) \ni \theta'(\rho') \\
\Rightarrow & \quad \{ \text{by the first equality} \} \\
& \langle \theta', \overline{\varphi'_i} \rangle \models E \wedge (\theta'(\varphi_k))(\theta'(\rho_1)) \ni \theta'(\rho_2) \\
\Rightarrow & \quad \{ \text{by definition of } \models \} \\
& \langle \theta', \overline{\varphi'_i} \rangle \models E \wedge \langle \theta', \overline{\theta'(\varphi_i)} \rangle \models \rho_1 \approx_k \rho_2 \\
\Rightarrow & \quad \{ \text{by Lemma A.4 (2) and definition of } \models \} \\
& \langle \theta', \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle \models \{\rho_1 \approx_k \rho_2\} \cup E
\end{aligned}$$

□

The following lemmata generalize these results to  $\implies^*$  derivations.

**Lemma A.7.** Assume the derivation  $E, \langle \theta, \overline{\varphi_i} \rangle \implies^* E', \langle \theta', \overline{\varphi'_i} \rangle$ . If the initial configuration is good, then so is the final configuration.

*Proof.* By induction on the length of the  $\implies^*$  derivation, and by Lemma A.3. □

**Lemma A.8.** If  $E, \langle \theta_0, \overline{\varphi_{i,0}^p} \rangle \implies^* E', \langle \theta, \overline{\varphi_i^p} \rangle$  then, for some  $\theta_1$ :

1.  $\theta = \theta_1 \circ \theta_0$
2.  $\theta_1(\varphi_{i,0}) \subseteq \varphi_i$  for all  $i \in \{1..p\}$ .

*Proof.* By induction on the length of the  $\implies^*$  derivation. If the length is zero we can prove the lemma by choosing  $\theta_1 = \text{id}$ . If the  $\implies^*$  derivation is of length greater than zero, we can split it as follows:

$$E, \langle \theta_0, \overline{\varphi_{i,0}^p} \rangle \implies E_1, \langle \theta_1, \overline{\varphi_{i,1}^p} \rangle \implies^* E', \langle \theta, \overline{\varphi_i^p} \rangle$$

By induction hypothesis we obtain:

$$\theta = \theta' \circ \theta_1 \quad \wedge \quad \overline{\theta'(\varphi_{i,1})} \subseteq \overline{\varphi_i^p} \quad \text{for some } \theta'$$

and, by applying Lemma A.5 in the first step:

$$\theta_1 = \theta'' \circ \theta_0 \quad \wedge \quad \overline{\theta''(\varphi_{i,0})} \subseteq \overline{\varphi_{i,1}^p} \quad \text{for some } \theta''$$

Hence we get:

$$\theta = (\theta' \circ \theta'') \circ \theta_1 \quad \wedge \quad \overline{(\theta' \circ \theta'')(\varphi_{i,1})} \subseteq \overline{\theta'(\varphi_{i,1})} \subseteq \overline{\varphi_i^p}$$

which proves the lemma. □

**Lemma A.9.** *Given the following derivation:*

$$E, \langle \theta_0, \overline{\varphi_{i,0}} \rangle \Longrightarrow^* E', \langle \theta \circ \theta_0, \overline{\varphi_i} \rangle$$

If  $\langle \theta', \overline{\varphi'_i} \rangle$  is a solution to  $E'$ , then  $\langle \theta' \circ \theta, \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle$  is a solution to  $E$ .

*Proof.* By induction on the length  $n$  of the  $\Longrightarrow^*$  derivation.

- **Case  $n = 0$**

We get  $\theta = id$  and, by hypothesis,  $\langle \theta', \overline{\varphi'_i} \rangle \models E'$ , which implies  $\langle \theta', \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle \models E$  by Lemma A.4 (2).

- **Case  $n > 0$**

Let us split the derivation as follows:

$$E, \langle \theta_0, \overline{\varphi_{i,0}} \rangle \Longrightarrow E_1, \langle \theta_1 \circ \theta_0, \overline{\varphi_{i,1}} \rangle \Longrightarrow^* E', \langle \underbrace{\theta_2 \circ \theta_1}_{\theta} \circ \theta_0, \overline{\varphi_i} \rangle$$

We obtain:

$$\begin{aligned} & \langle \theta', \overline{\varphi'_i} \rangle \models E' \\ \Rightarrow & \quad \{ \text{by i.h.} \} \\ & \langle \theta' \circ \theta_2, \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle \models E_1 \\ \Rightarrow & \quad \{ \text{by applying Lemma A.6 on the first step} \} \\ & \langle \theta' \circ \theta_2 \circ \theta_1, \overline{\varphi'_i \cup \theta'(\varphi_i) \cup (\theta' \circ \theta_2)(\varphi_{i,1})} \rangle \models E \\ \Rightarrow & \quad \{ \text{by Lemma A.8, } \forall i \in \{1..p\}. \theta_2(\varphi_{i,1}) \subseteq \varphi_i \} \\ & \langle \theta' \circ \theta_2 \circ \theta_1, \overline{\varphi'_i \cup \theta'(\varphi_i)} \rangle \models E \end{aligned}$$

□

Finally, we prove that if we start applying the  $\Longrightarrow^*$  rules on a set of constraints  $E$  with the empty solution and we reach a configuration with no equations to be solved, the final result is a solution to  $E$ , not only in the sense of Definition A.1, but also in the sense of Definition 4.5.

**Corollary A.10** (Soundness). *Assume a set  $E$  of constraints and  $\theta, \overline{\varphi_i^p}$  such that the following judgment can be derived from the rules of Figure A.1:*

$$E, \langle id, \overline{[]^p} \rangle \Longrightarrow^* \emptyset, \langle \theta, \overline{\varphi_i^p} \rangle$$

Then:

1. For each  $i \in \{1..p\}$ ,  $\varphi_i$  is a function from RTVs to RTVs.
2.  $\langle \theta, \overline{\varphi_i^p} \rangle$  is a solution to  $E$ .

*Proof.* The first configuration is trivially good. By Lemma A.7, the final configuration is also good. Now let us prove (1) by contradiction: assume that there exists some  $\varphi_k$  not being a function. Then there exist  $\rho, \rho_a, \rho_b$  such that  $\varphi(\rho) \ni \{\rho_a, \rho_b\}$  and  $\rho_a \neq \rho_b$ . Since the final configuration is good, there must exist a chain of the form:

$$\{\rho_a \equiv \rho_1 = \dots = \rho_m \equiv \rho_b\} \subseteq \emptyset$$

Which is contradictory, since the chain must have at least one equation. Therefore our assumption was false, and every  $\overline{\varphi_i}$  is a function. The result (2) is a particular instance of Lemma A.9.  $\square$

**Lemma A.11** (Termination). *Given a finite configuration  $E, \langle \theta, \overline{\varphi_i} \rangle$ , there does not exist an infinite  $\Longrightarrow^*$  derivation:*

$$E, \langle \theta, \overline{\varphi_i} \rangle \Longrightarrow E_1, \langle \theta_1, \overline{\varphi_{1,i}} \rangle \Longrightarrow \dots \Longrightarrow E_n, \langle \theta_n, \overline{\varphi_{n,i}} \rangle \Longrightarrow \dots$$

*Proof.* Let us define the size function  $\|\cdot\| : \mathbf{Type} \rightarrow \mathbb{N}$  as follows:

$$\begin{aligned} \|\alpha\| &= \|\rho\| = \|B\| = 1 \\ \|T \overline{s_i^n} @ \overline{\rho_j^m}\| &= \|T \overline{s_i^n} !@ \overline{\rho_j^m}\| = \|T \overline{s_i^n} \#@ \overline{\rho_j^m}\| = 1 + \sum_{i=1}^n \|s_i\| + \sum_{j=1}^m \|\rho_j\| \\ \|\overline{t_i^n} \rightarrow \overline{\rho_j^m} \rightarrow s\| &= 1 + \sum_{i=1}^n \|t_i\| + \sum_{j=1}^m \|\rho_j\| \end{aligned}$$

The domain of this function can be easily extended to constraints:

$$\begin{aligned} \|\tau_1 = \tau_2\| &= \|\tau_1\| + \|\tau_2\| \\ \|s_1 = isData(s_2, \rho)\| &= \|s_1\| + \|s_2\| + \|\rho\| \\ \|\tau_1 \approx_k \tau_2\| &= \|\tau_1\| + \|\tau_2\| \end{aligned}$$

Let  $\Phi$  be a function mapping a configuration  $E, \langle \theta, \overline{\varphi_i} \rangle$  to a tuple  $(N_\alpha, N_{isData}, N_\rho, s, N_{\approx})$  such that:

- $N_\alpha$  is the number of *distinct* type variables occurring in  $E$ .
- $N_{isData}$  is the number of *isData* constraints in  $E$ .
- $N_\rho$  is the number of *distinct* RTVs occurring in  $E$  and in the  $\overline{\varphi_i}$ .
- $s$  is the sum of the sizes of the constraints in  $E$ .
- $N_{\approx}$  is the number of  $\approx_k$ -constraints in  $E$ .

Then, if  $E, \langle \theta, \overline{\varphi_i} \rangle \Longrightarrow E', \langle \theta', \overline{\varphi'_i} \rangle$ , Figure A.3 shows that  $\Phi(E, \langle \theta, \overline{\varphi_i} \rangle) > \Phi(E', \langle \theta', \overline{\varphi'_i} \rangle)$ , where  $>$  is the lexicographical order in  $\mathbb{N}^5$ . If we apply  $\Phi$  to every configuration in a infinite  $\Longrightarrow^*$  derivation we would obtain a infinite  $>$  chain in  $\mathbb{N}^5$ , which is not possible since  $(\mathbb{N}^5, \leq)$  is a well-founded set.  $\square$

<b>Rule</b>	$N_\alpha$	$N_{isData}$	$N_\rho$	$s$	$N_\approx$
[ID]	$\leq$	$=$	$\leq$	$<$	$=$
[VAR]	$<$	$=$	$=$	$\geq$	$=$
[REG]	$=$	$=$	$<$	$\geq$	$=$
[FUN]	$=$	$=$	$=$	$<$	$=$
[CONS]	$=$	$=$	$=$	$<$	$=$
[ISD1]	$=$	$<$	$>$	$>$	$=$
[ISD2]	$=$	$<$	$=$	$<$	$=$
[ISD3]	$=$	$<$	$=$	$<$	$=$
[QFUN]	$=$	$=$	$=$	$<$	$>$
[QCONS]	$=$	$=$	$=$	$<$	$>$
[QVAR-L]	$<$	$=$	$>$	$\geq$	$=$
[QVAR-R]	$<$	$=$	$>$	$\geq$	$=$
[QREG1]	$=$	$=$	$=$	$=$	$<$
[QREG2]	$=$	$=$	$=$	$=$	$<$
[QEND]	$=$	$=$	$=$	$=$	$<$

Figure A.3: Variation of each component in the  $\Phi$  function when applying each unification rule. A  $<$  sign in a cell means that the value of the corresponding component in the size of the final configuration is strictly lower than in the size of the initial one. Similarly with  $\leq$ ,  $=$ ,  $\geq$  and  $>$ . Notice that, considering the lexicographical order  $(N_\alpha, N_{isData}, N_\rho, s, N_\approx)$ , the  $\implies$  relation always gives a configuration of lower size than the size of the initial one.

## Appendix B

# Correctness of the initial bounds

In Section 7.5 we introduced three algorithms for computing an initial signature  $(\Delta, \mu, \sigma)$  in order to bound the stack and memory needs of a given recursive function definition. In that section we were concerned about the reductivity of each of the components of the signature, but we did not state explicitly that these components are correct bounds to the actual runtime figures  $(\delta, m, s)$ . This statement was intentionally left out, since we had shown in Section 7.5.1 the fact of an upper bound being reductive was a sufficient condition to its correctness.

Nevertheless, reductivity is not a *necessary* condition to obtain correct bounds. It was pointed out in Section 7.5.6 that reductivity can only be guaranteed under some admissibility conditions on the externally given bounds to the size of the call tree ( $nb$ ,  $nr$ , and  $len$ ). These conditions were formalised in Definitions 7.41, 7.42 (page 247), and 7.47 (page 252). When these conditions do not hold, the results of the algorithms computing the initial bounds may not be reductive, but they still would be useful if we are able to show their correctness. This shows the need for an *explicit* proof of the correctness properties of the results of *computeDelta*, *computeMu*, and *computeSigma*, without basing our proof on reductivity arguments.

In this Appendix we aim to show the following fact: If  $(\Delta_0, \mu_0, \sigma_0)$  are initial approximations to the memory needs of  $f$ , they have been computed with the algorithms of Section 7.5, and the  $nb$ ,  $nr$ , and  $len$  functions are *correct* approximations to the corresponding runtime figures of the call tree, then  $(\Delta_0, \mu_0, \sigma_0)$  is a correct signature with respect to  $f$ . The proof of this statement is very technical, since it requires several extensions to the resource-aware big-step semantics defined in Figure B.1.

*Remark.* Throughout this section we assume the absence of explicit destruction via **case!** expressions. We also assume, by convention, that  $\max \emptyset = 0$ .

### B.1 Before/after semantics with call tree counters

In this section we will make precise the idea of the  $nb$ ,  $nr$ , and  $len$  functions being correct approximations to the actual parameters of the call tree deployed at runtime. This is done by extending our semantic judgements with a triple  $(N_b, N_r, L)$  of natural numbers representing the number of base calls, recursive calls, and the maximum depth of the call tree. Obviously, it makes little sense to talk about recursive calls, if we do not specify *which* function do these numbers refer to. Therefore, the name of this function will be attached to the arrow of these judgements, as in  $\Downarrow_f$ .

Besides this, it turns out to be useful to distinguish between the charges done in memory before the last recursive call in the current context, and those done after this call. For this reason, we shall split

each of the  $\delta$  and  $m$  components of our resource vector into two subcomponents: their *before* part, ( $\delta_b$  and  $m_b$ ), and their *after* part ( $\delta_a$  and  $m_a$ ). The former takes into account the expressions being executed before (and including) the last recursive call in the current call context, and the latter takes into account the expressions being executed after (and excluding) that recursive call. By convention, we assume that, if the evaluation of an expression does not contain any recursive call, all its charges are stored in the after part, and the before part is left with no charges.

Given the above, our extended semantics defines the derivation of judgements of the following form:

$$E \vdash h, k, e \Downarrow_f h', k, v, (N_b, N_r, L), (\delta_b / \delta_a, m_b / m_a)$$

We omit the  $s$  component of the resource vector, since no distinction between the before and after part is necessary for that component, as we shall see later. The semantic rules are shown in Figure B.1. The  $[Basic_{NC}]$  rule is used when the expression being evaluated does not contain recursive calls to  $f$ . In this case the evaluation can be described by our usual  $\Downarrow$  rules of Figure 2.26. We follow our convention mentioned above, and we consider their charges in the after part. The opposite situation is that of rule  $[AppR_{NC}]$ , where all the charges are added to the before component. The number of recursive calls  $N_r$ , and the length of the call-tree  $L$  are incremented accordingly. When evaluating a **let** expression we have to distinguish whether a recursive call to  $f$  is done during the evaluation of  $e_2$ . This can be done with the help of the  $L_2$  component of the call tree counter returned by the evaluation of  $e_2$ . If there are no recursive calls in  $e_2$  we obtain  $L_2 = 1$  (i.e. the depth of the call tree is 1), and we apply  $[Let1_{NC}]$ . In this case, the before part corresponds to the before part of the evaluation of  $e_1$ , whereas the after part comprises the after part of  $e_1$  and the whole evaluation of  $e_2$ . The  $[Let2_{NC}]$  rule is used when there are recursive calls in  $e_2$  (that is,  $L_2 > 1$ ). The before part includes the evaluation of  $e_1$  and the before part of  $e_2$ , whereas the after part only contains the after part of  $e_2$ . In this case, the formalization of the  $(N_r, N_b, L)$  requires further case distinction, and we define in Figure B.2 a separate operator  $\oplus$  for this purpose. The  $[Case_{NC}]$  rule just propagates the before/after information, and the call tree counters given by the branch being executed.

From these semantic rules it is easy to show that the new components introduced are just counters, and they do not influence the evaluation of the expression. As a consequence, if we can execute an expression  $e$  under the  $\Downarrow$  semantics for a given environment  $E$  and heap  $h$  with  $k$  regions, we can do the same under the  $\Downarrow_f$  semantics (for any  $f$ ), and we will get the same normal form  $v$  and final heap  $h'$ . A little less obvious is the relation between the  $(\delta_b / \delta_a, m_b / m_a)$  components of the  $\Downarrow_f$  semantics, and the original resource vector  $(\delta, m, s)$  given by the corresponding  $\Downarrow$ -judgement. The whole  $\delta$  component is equivalent to the (region-wise) addition of the before and after parts. With respect to the  $m$  component, an intuitive idea of its relation with the  $m_b$  and  $m_a$  is given by Figure 2.30. The following lemma states these results formally.

**Lemma B.1.** *Let us assume an evaluation  $E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$  under a signature environment  $\Sigma$ . Given a function  $f \in \text{dom } \Sigma$ , there exists a single tuple  $(N_b, N_r, L)$  and a single tuple  $(\delta_b / \delta_a, m_b / m_a)$  such that  $E \vdash h, k, e \Downarrow_f h', k, v, (N_b, N_r, L), (\delta_b / \delta_a, m_b / m_a)$ . Moreover, it holds that:*

1.  $N_b \geq 1, N_r \geq 0$ , and  $L \geq 1$ .
2. If  $L = 1$ , then  $\delta_b = []_k$  and  $m_b = 0$ .
3.  $\delta = \delta_b + \delta_a$ .
4.  $m = \max\{m_b, |\delta_b| + m_a\}$ .

$$\begin{array}{c}
\frac{E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s) \quad e \text{ is of the form } c, x, a_1 \oplus a_2, x @ r, C \bar{a}_i @ r \text{ or } g \bar{a}_i @ \bar{r}_j \text{ with } g \neq f}{E \vdash h, k, e \Downarrow_f h', k, v, (1, 0, 1), ([ ]_k / \delta, 0 / m)} \text{ [Basic}_{\text{NC}}] \\
\\
\frac{(f \bar{y}_i^n @ \bar{r}_j^l = e_f) \in \Sigma \quad \overline{[y_i \mapsto E(a_i)]^n, [r_j \mapsto E(r_j)]^l, \text{self} \mapsto k+1]} \vdash h, k+1, e_f \Downarrow_f h', k+1, v, (N_b, N_r, L), (\delta_b / \delta_a, m_b / m_a)}{E \vdash h, k, f \bar{a}_i^n @ \bar{r}_j^l \Downarrow_f h' |_k, k, v, (N_b, N_r + 1, L + 1), ((\delta_b + \delta_a) |_k / [ ]_k, \max\{m_b, |\delta_b| + m_a\} / 0)} \text{ [AppR}_{\text{NC}}] \\
\\
\frac{L_2 = 1 \quad E \vdash h, k, e_1 \Downarrow_f h', k, v_1, (N_{b,1}, N_{r,1}, L_1), (\delta_{b,1} / \delta_{a,1}, m_{b,1} / m_{a,1}) \quad E \cup [x_1 \mapsto v_1] \vdash h', k, e_2 \Downarrow_f h'', k, v, (N_{b,2}, N_{r,2}, L_2), (\delta_{b,2} / \delta_{a,2}, m_{b,2} / m_{a,2})}{E \vdash h, k, \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \Downarrow_f h'', k, v, (N_{b,1}, N_{r,1}, \max\{L_1, L_2\}), (\delta_{b,1} / \delta_{a,1} + \delta_{a,2}, m_{b,1} / \max\{m_{a,1}, |\delta_{a,1}| + m_{a,2}\})} \text{ [Let1}_{\text{NC}}] \\
\\
\frac{L_2 \neq 1 \quad m_b = \max\{m_{b,1}, |\delta_{b,1}| + m_{a,1}, |\delta_{b,1}| + |\delta_{a,1}| + m_{b,2}\} \quad (N_b, N_r, L) = (N_{b,1}, N_{r,1}, L_1) \otimes (N_{b,2}, N_{r,2}, L_2) \quad E \vdash h, k, e_1 \Downarrow_f h', k, v_1, (N_{b,1}, N_{r,1}, L_1), (\delta_{b,1} / \delta_{a,1}, m_{b,1} / m_{a,1}) \quad E \cup [x_1 \mapsto v_1] \vdash h', k, e_2 \Downarrow_f h'', k, v, (N_{b,2}, N_{r,2}, L_2), (\delta_{b,2} / \delta_{a,2}, m_{b,2} / m_{a,2})}{E \vdash h, k, \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \Downarrow_f h'', k, v, (N_b, N_r, L), (\delta_{b,1} + \delta_{a,1} + \delta_{b,2} / \delta_{a,2}, m_b / m_{a,2})} \text{ [Let2}_{\text{NC}}] \\
\\
\frac{C = C_r \quad E(x) = p \quad E \uplus [\bar{x}_{r_j} \mapsto \bar{v}_j^{n_r}] \vdash h, k, e_r \Downarrow_f h', k, v, (N_b, N_r, L), (\delta_b / \delta_a, m_b / m_a)}{E \vdash h[p \mapsto (j, C \bar{v}_i^{n_r})], k, \mathbf{case} \ x \ \mathbf{of} \ \bar{C}_i \bar{x}_{ij}^{n_i} \rightarrow e_i \Downarrow_f h', k, v, (N_b, N_r, L), (\delta_b / \delta_a, m_b / m_a)} \text{ [Case]}
\end{array}$$

Figure B.1: Big-step semantics enriched with components for counting the number of recursive calls, and for distinguishing the charges done before and after the last recursive call.

$$(N_{b,1}, N_{r,1}, L_1) \otimes (N_{b,2}, N_{r,2}, L_2) = (N_b, N_r, \max\{L_1, L_2\})$$

$$\text{where } (N_b, N_r) = \begin{cases} (N_{b,2}, N_{r,2}) & \text{if } N_{r,1} = 0 \wedge N_{r,2} \neq 0 \\ (N_{b,1}, N_{r,1}) & \text{if } N_{r,2} = 0 \\ (N_{b,1} + N_{b,2}, N_{r,1} + N_{r,2} - 1) & \text{otherwise} \end{cases}$$

Figure B.2: Definition of  $\otimes$  operator on  $(N_b, N_r, L)$  tuples.

*Proof.* The existence and uniqueness of the respective tuples can be easily shown by induction on the size of the  $\Downarrow$  derivation. They are direct consequence of the semantic rules being deterministic. Hence, let us prove (1), (2), (3), and (4) by induction on the  $\Downarrow$  derivation. Let us distinguish cases according to the last  $\Downarrow$ -rule applied. Cases [Lit], [Var], [PrimOp], [Copy], [Cons], and [App] when the function being applied is different from  $f$ , are trivial. In the case of rules [Case], and [App] when the function being applied is  $f$ , the result follows trivially from the induction hypothesis. The only subtlety when proving (3) is related to the restriction of the domain of  $\delta$  occurring in [App], but we get:

$$\delta|_k = (\delta_a + \delta_b)|_k = \delta_a|_k + \delta_b|_k$$

The [Let] case is more involved. Assume the tuples  $(N_{b,1}, N_{r,1}, L_1)$  and  $(\delta_{b,1}/\delta_{a,1}, m_{b,1}/m_{a,1})$ , corresponding to the  $\Downarrow$ -evaluation of  $e_1$ , and the tuples  $(N_{b,2}, N_{r,2}, L_2)$  and  $(\delta_{b,2}/\delta_{a,2}, m_{b,2}/m_{a,2})$  regarding the  $\Downarrow$ -evaluation of  $e_2$ . By induction hypothesis, we get:

$$N_{b,i} \geq 1 \quad N_{r,i} \geq 0 \quad L_i \geq 1 \quad \text{for } i \in \{1, 2\} \quad (\text{B.1})$$

$$\delta_i = \delta_{b,i} + \delta_{a,i} \quad \text{for } i \in \{1, 2\} \quad (\text{B.2})$$

$$m_i = \max\{m_{b,i}, |\delta_{b,i}| + m_{a,i}\} \quad \text{for } i \in \{1, 2\} \quad (\text{B.3})$$

where  $\delta_i$  and  $m_i$  are the figures of the resource vectors occurring in each  $\Downarrow$ -subderivation. Now we distinguish cases:

- **Case  $L_2 = 1$ .** Then  $N_b = N_{b,1}$ ,  $N_r = N_{r,1}$ , and  $L = \max\{L_1, L_2\}$ . The fact (1) follows from equation (B.1). With regard to (2), if  $L = 1$ , then  $L_1 = 1$ , which implies  $\delta_{b,1} = []|_k$  and  $m_{b,1} = 0$ , so that property holds. Moreover, by induction hypothesis, we get  $\delta_{b,2} = []|_k$  and  $m_{b,2} = 0$ , since  $L_2 = 1$ . Therefore:

$$\begin{aligned} & \delta_b + \delta_a \\ &= \delta_{b,1} + \delta_{a,1} + \delta_{a,2} \\ &= \delta_{b,1} + \delta_{a,1} + \delta_{b,2} + \delta_{a,2} \quad \{ \text{since } \delta_{b,2} = []|_k \} \\ &= \delta_1 + \delta_2 \quad \{ \text{by (B.2)} \} \\ &= \delta \end{aligned}$$

$$\begin{aligned} & \max\{m_b, |\delta_b| + m_a\} \\ &= \max\{m_{b,1}, |\delta_{b,1}| + \max\{m_{a,1}, |\delta_{a,1}| + m_{a,2}\}\} \\ &= \max\{m_{b,1}, |\delta_{b,1}| + m_{a,1}, |\delta_{b,1}| + |\delta_{a,1}| + m_{a,2}\} \\ &= \max\{m_1, |\delta_1| + m_{a,2}\} \quad \{ \text{by (B.2) and (B.3)} \} \\ &= \max\{m_1, |\delta_1| + \max\{m_{b,2}, |\delta_{b,2}| + m_{a,2}\}\} \quad \{ \text{since } m_{a,2} \geq 0, m_{b,2} = 0 \text{ and } |\delta_{b,2}| = 0 \} \\ &= \max\{m_1, |\delta_1| + m_2\} \quad \{ \text{by (B.3)} \} \end{aligned}$$

- **Case  $L_2 > 1$ .** We get  $(N_b, N_r, L) = (N_{b,1}, N_{r,1}, L_1) \otimes (N_{b,2}, N_{r,2}, L_2)$ . From induction hypothesis

it follows that  $N_b \geq 1$  and  $L \geq 1$ . With respect to  $N_r$ , the only caution to take is the case when  $N_r = N_{r,1} + N_{r,2} - 1$ , but this only holds when both  $N_{r,1}$  and  $N_{r,2}$  are strictly positive, so  $N_r \geq 1$ , and (1) holds. Property (2) holds vacuously, since  $L_2 > 1$  implies  $L > 1$ . Now we prove (3) and (4):

$$\begin{aligned}
& \delta_b + \delta_a \\
&= \delta_{b,1} + \delta_{a,1} + \delta_{b,2} + \delta_{a,2} \\
&= \delta_1 + \delta_2 \quad \{ \text{by (B.2)} \} \\
&= \delta
\end{aligned}$$

$$\begin{aligned}
& \max\{m_b, |\delta_b| + m_a\} \\
&= \max\{m_{b,1}, |\delta_{b,1}| + m_{a,1}, |\delta_{b,1}| + |\delta_{a,1}| + m_{b,2}, |\delta_{b,1}| + |\delta_{a,1}| + |\delta_{b,2}| + m_{a,2}\} \\
&= \max\{m_1, |\delta_1| + m_{b,2}, |\delta_1| + |\delta_{b,2}| + m_{a,2}\} \quad \{ \text{by (B.2) and (B.3)} \} \\
&= \max\{m_1, |\delta_1| + \max\{m_{b,2}, |\delta_{b,2}| + m_{a,2}\}\} \\
&= \max\{m_1, |\delta_1| + m_2\}
\end{aligned}$$

□

In the following we shall use the letters  $\varphi, \psi, \chi$ , etc. to denote  $\Downarrow$ -judgements. Given one of these:

$$\varphi \equiv E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$$

we have shown that, given a function  $f$ , there exist unique numbers  $N_b, N_r, L, m_b$ , and  $m_a$ , and unique mappings  $\delta_a, \delta_b$  such that  $E \vdash h, k, e \Downarrow_f h', k, v, (N_b, N_r, L), (\delta_b/\delta_a, m_b/m_a)$  can be derived. All these components are determined by the judgement  $\varphi$ , so we can use the notation  $N_b^f(\varphi), N_r^f(\varphi), L^f(\varphi), \delta_b^f(\varphi)$ , and  $\delta_a^f(\varphi)$  to refer to these components. By abuse of notation, we use  $\delta(\varphi), m(\varphi), s(\varphi)$  for denoting, respectively, the  $\delta, m$ , and  $s$  components of the resource vector occurring in the judgement  $\varphi$ . Moreover, we use  $Exp(\varphi)$  for denoting the expression being evaluated in  $\varphi$ .

Our abstract interpretation functions  $\llbracket \cdot \rrbracket_\Delta$  and  $\llbracket \cdot \rrbracket_\mu$  defined in Section 7.3 were defined in terms of sequences of basic expressions. The following lemma establishes the relation of these sequences to the before and after parts of the evaluation of an expression. It also presents some useful properties of the call tree.

**Lemma B.2.** *Assume the following judgement:*

$$\varphi \equiv E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$$

There exists a sequence  $[G \rightarrow be_1, \dots, be_n] \in \text{seqs } e$  and some  $\overline{\varphi}_i^n, \overline{E}_i^n, \overline{h}_i^n, \overline{td}_i^n, \overline{h}'_i^n, \overline{v}_i^n, \overline{\delta}_i^n, \overline{m}_i^n, \overline{s}_i^n$  such that:

1. For every  $i \in \{1..n\}$ ,  $\varphi_i \equiv E_i \vdash h_i, k, td_i, be_i \Downarrow h'_i, k, v_i, (\delta_i, m_i, s_i)$ , and  $\varphi_i$  belongs to the derivation of  $\varphi$ .
2. Let  $I \subseteq \{1..n\}$  be the set of indices  $i$  such that  $be_i$  has the form  $f \overline{a}_i @ \overline{r}_j$  for some  $f, \overline{a}_i$  and  $\overline{r}_j$  of their respective types. Then:

$$\begin{aligned}
N_b^f(\varphi) &= \begin{cases} 1 & \text{if } I = \emptyset \\ \sum_{i \in I} N_b^f(\varphi_i) & \text{otherwise} \end{cases} \\
N_r^f(\varphi) &= \begin{cases} 0 & \text{if } I = \emptyset \\ 1 + \sum_{i \in I} (N_r^f(\varphi_i) - 1) & \text{otherwise} \end{cases}
\end{aligned}$$

$$L^f(\varphi) = \begin{cases} 1 & \text{if } I = \emptyset \\ 1 + \max_{i \in I} \{L^f(\varphi_i) - 1\} & \text{otherwise} \end{cases}$$

3. Being  $I$  defined as above, let  $p = \max I$ , that is, the index of the last recursive call (if any; otherwise take  $p = 0$ ). Then:

$$\begin{aligned} \delta_b^f(\varphi) &= \sum_{i=1}^p \delta_i & \delta_a^f(\varphi) &= \sum_{i=p+1}^n \delta_i \\ m_b^f(\varphi) &= \max_{i \in \{1..p\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} & m_a^f(\varphi) &= \max_{i \in \{p+1..n\}} \left\{ \sum_{j=p+1}^{i-1} |\delta_j| + m_i \right\} \\ \delta &= \sum_{i=1}^n \delta_i & m &= \max_{i \in \{1..n\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \end{aligned} \quad (\text{B.4})$$

*Proof.* Let us start with the last equation, that is, (B.4), by assuming that the remaining facts of the third conclusion have been proved, and by using Lemma B.1:

$$\delta = \delta_b^f(\varphi) + \delta_a^f(\varphi) = \sum_{i=1}^p \delta_i + \sum_{i=p+1}^n \delta_i = \sum_{i=1}^n \delta_i$$

$$\begin{aligned} m &= \max\{m_b^f(\varphi), |\delta_b^f(\varphi)| + m_a^f(\varphi)\} \\ &= \max \left\{ \max_{i \in \{1..p\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=1}^k \delta_i + \max_{i \in \{p+1..n\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\} \right\} \\ &= \max \left\{ \max_{i \in \{1..k\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \max_{i \in \{k+1..n\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \right\} \\ &= \max_{i \in \{1..n\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \end{aligned}$$

The rest of the lemma is proven by induction on the size of the  $\Downarrow$ - derivation. We distinguish cases depending on the structure of  $e$ .

- **Cases**  $e \equiv c, e \equiv x, e \equiv a_1 \oplus a_2, e \equiv x @ r, e \equiv C \bar{a}_i @ \bar{r},$  and  $e \equiv g \bar{a}_i @ \bar{r}_j$  with  $g \neq f$

In all these cases we get  $\text{seqs } e = \{[e]\}$ , and (1) holds trivially by taking  $\varphi_1 \equiv \varphi$ . Moreover, we can apply the  $[Basic_{NC}]$  rule, so  $N_b^f(\varphi) = 1, N_r^f(\varphi) = 0, L^f(\varphi) = 1,$  and (2) follows from the fact that  $I = \emptyset$  (we have a single expression in the sequence, which is not a call to  $f$ ). Finally, since  $p = 0,$  and, again, by the  $[Basic_{NC}]$  rule:

$$\delta_b^f(\varphi) = [ ]_k = \sum_{i=1}^0 \delta_i \quad \delta_a^f(\varphi) = \delta = \delta_1 = \sum_{i=1}^1 \delta_i$$

$$m_b^f = 0 = \max_{i \in \emptyset} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \quad m_a^f(\varphi) = m = m_1 = \max_{i \in \{1\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}$$

Therefore, (3) holds.

- **Case**  $e \equiv f \bar{a}_i @ \bar{r}_j$

Again, we get a sequence with a single element ( $n = 1$ ), so (1) holds with  $\varphi_1 \equiv \varphi$ . In this case it holds that  $I = \{1\}$ , so (2) is proved as follows:

$$N_b^f(\varphi) = N_b^f(\varphi_1) \quad N_r^f(\varphi) = 1 + (N_r^f(\varphi_1) - 1) \quad L^f(\varphi) = 1 + (L^f(\varphi_1) - 1)$$

With respect to (3), we denote by  $\psi$  the judgment corresponding to the evaluation of the body of the function  $f$ , which is just above  $\varphi$  in the derivation tree of the latter, and by  $(\delta_f, m_f, s_f)$  the resource vector resulting from its evaluation. By applying the  $[AppR_{NC}]$  rule and Lemma B.1 we get:

$$\delta_b^f(\varphi) = (\delta_b^f(\psi) + \delta_a^f(\psi))|_k = \delta_f|_k = \delta_1 = \sum_{i=1}^1 \delta_i$$

$$\delta_a^f(\varphi) = []_k = \sum_{i=1}^0 \delta_i$$

$$m_b^f = \max\{m_b^f(\psi), |\delta_b^f(\psi)| + m_a^f(\psi)\} = m_f = m_1 = \max_{i \in \{1\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}$$

$$m_a^f(\varphi) = 0 = \max_{i \in \emptyset} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}$$

- **Case  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$**

We get the following judgements belonging to  $\varphi$ :

$$\psi_A \equiv E \vdash h, k, 0, e_1 \Downarrow h_A, k, v_A, (\delta_A, m_A, s_A)$$

$$\psi_B \equiv E \cup [x_1 \mapsto v_A] \vdash h_A, k, td + 1, e_2 \Downarrow h_B, k, v, (\delta_B, m_B, s_B)$$

By applying induction hypothesis, there exist two sequences  $[G_1 \rightarrow be_1, \dots, be_n] \in \text{seqs } e_1$  and  $[G_2 \rightarrow be_1, \dots, be_m]$  satisfying the lemma. By definition of *seqs*, the sequence

$$[G_1 \wedge G_2 \rightarrow be_1, \dots, be_n, be'_1, \dots, be'_m]$$

belongs to *seqs*  $e$ , so let us prove that the conclusions of the lemma hold for this sequence. The first condition follows from the induction hypothesis. If we denote by  $\overline{\varphi}_i^n$  the judgements belonging to  $\psi_A$ , and by  $\overline{\varphi}_i^m$  the judgements of  $\psi_B$  (the existence of both lists is justified by induction hypothesis), it is obvious that both the  $\overline{\varphi}_i^n$  and  $\overline{\varphi}_i^m$  belong to the main judgement  $\varphi$ . Now we move on to (2). Before this, let us define, for each  $j \in \{n+1, \dots, n+m\}$ ,  $be_j = be'_{j-n}$ ,  $\varphi_j = \varphi'_j$ . Let  $I_A$  be the set of indices  $i$  such that  $be_i$  is a function application, and let  $I'_B$  be defined similarly, but with the  $be'_i$ . We define  $I_B = \{n+j \mid j \in I'_B\}$ , and  $I = I_A \cup I_B$ . In this way we ensure that, for every  $i \in \{1..n+m\}$ ,  $be_i$  is a function application if and only if  $i \in I$ . Now we distinguish cases in order to prove (2):

- $I_A = \emptyset, I_B = \emptyset$

This implies that both  $I_A$  and  $I_B$  (and hence  $I'_B$ ) are empty. By induction hypothesis, this means that:

$$N_b^f(\psi_A) = 1 \quad N_r^f(\psi_A) = 0 \quad L^f(\psi_A) = 1 \quad (\text{B.5})$$

$$N_b^f(\psi_B) = 1 \quad N_r^f(\psi_B) = 0 \quad L^f(\psi_B) = 1 \quad (\text{B.6})$$

Therefore, we get, by rule [Let1<sub>NC</sub>]:

$$N_b^f(\varphi) = N_b^f(\psi_A) = 1 \quad N_r^f(\varphi) = N_r^f(\psi_A) = 0 \quad L^f(\varphi) = \max\{L^f(\psi_A), L^f(\psi_B)\} = 1$$

and (2) holds.

–  $I_A \neq \emptyset, I_B = \emptyset$

Then  $I = I_A$ , but the facts (B.6) also hold in this case. Again, by [Let1<sub>NC</sub>]:

$$N_b^f(\varphi) = N_b^f(\psi_A) = \sum_{i \in I_A} N_b^f(\varphi_i) = \sum_{i \in I} N_b^f(\varphi_i)$$

$$N_r^f(\varphi) = N_r^f(\psi_A) = 1 + \sum_{i \in I_A} (N_r^f(\varphi_i) - 1) = 1 + \sum_{i \in I} (N_r^f(\varphi_i) - 1)$$

$$L^f(\varphi) = \max\{L^f(\psi_A), 1\} = L^f(\psi_A) = 1 + \max_{i \in I_A} \{L^f(\varphi_i) - 1\} = 1 + \max_{i \in I} \{L^f(\varphi_i) - 1\}$$

where, in each case, the second step is justified by induction hypothesis. Thus (2) holds.

–  $I_A = \emptyset, I_B \neq \emptyset$

In this case we apply [Let2<sub>NC</sub>] in order to obtain:

$$N_b^f(\varphi) = N_b^f(\psi_B) = \sum_{i \in I'_B} N_b^f(\varphi'_i) = \sum_{i \in I_B} N_b^f(\varphi_i) = \sum_{i \in I} N_b^f(\varphi_i)$$

and similarly with  $N_r^f(\varphi)$  and  $L^f(\varphi)$ , so (2) holds.

–  $I_A \neq \emptyset, I_B \neq \emptyset$

Again, we apply [Let2<sub>NC</sub>] and get:

$$\begin{aligned} N_b^f(\varphi) &= N_b^f(\psi_A) + N_b^f(\psi_B) \\ &= \sum_{i \in I_A} N_b^f(\varphi_i) + \sum_{i \in I'_B} N_b^f(\varphi'_i) \\ &= \sum_{i \in I_A} N_b^f(\varphi_i) + \sum_{i \in I_B} N_b^f(\varphi_i) \\ &= \sum_{i \in I} N_b^f(\varphi_i) \end{aligned}$$

$$\begin{aligned} N_r^f(\varphi) &= N_r^f(\psi_A) + N_r^f(\psi_B) - 1 \\ &= 1 + \sum_{i \in I_A} (N_r^f(\varphi_i) - 1) + 1 + \sum_{i \in I'_B} (N_r^f(\varphi'_i) - 1) - 1 \\ &= 1 + \sum_{i \in I_A} (N_r^f(\varphi_i) - 1) + 1 + \sum_{i \in I_B} (N_r^f(\varphi_i) - 1) - 1 \\ &= 1 + \sum_{i \in I} (N_r^f(\varphi_i) - 1) \end{aligned}$$

$$\begin{aligned}
L^f(\varphi) &= \max\{L^f(\psi_A), L^f(\psi_B)\} \\
&= \max\{1 + \max_{i \in I_A}\{L^f(\varphi_i) - 1\}, 1 + \max_{i \in I'_B}\{L^f(\varphi'_i) - 1\}\} \\
&= 1 + \max\{\max_{i \in I_A}\{L^f(\varphi_i) - 1\}, \max_{i \in I_B}\{L^f(\varphi_i) - 1\}\} \\
&= 1 + \max_{i \in I}\{L^f(\varphi_i) - 1\}
\end{aligned}$$

Finally, let us prove (3). For every  $i \in \{1..n+m\}$ , let  $(\delta_i, m_i, s_i)$  be the resource vector occurring in  $\varphi_i$ . For every  $i \in \{1..n\}$ , let  $(\delta'_i, m'_i, s'_i)$  be the resource vector of  $\varphi'_i$ . With  $I_A, I'_B$ , and  $I_B$  defined as above, we define  $p_A = \max I_A$  and  $p_B = \max I'_B$ , so  $n + p_B = \max I_B$ . Let us distinguish cases:

–  $I_B = \emptyset$

In this case  $p_B = 0$ , and  $L^f(\psi_B) = 1$ . Therefore,  $p = \max(I_A \cup I_B) = p_A$ . We obtain, by [Let1<sub>NC</sub>], and the induction hypothesis:

$$\delta_b^f(\varphi) = \delta_b^f(\psi_A) = \sum_{i=1}^{p_A} \delta_i = \sum_{i=1}^p \delta_i$$

$$\begin{aligned}
\delta_a^f(\varphi) &= \delta_a^f(\psi_A) + \delta_a^f(\psi_B) = \sum_{i=p_A+1}^n \delta_i + \sum_{i=p_B+1}^m \delta'_i = \sum_{i=p+1}^n \delta_i + \sum_{i=1}^m \delta'_i \\
&= \sum_{i=p+1}^n \delta_i + \sum_{i=n+1}^{n+m} \delta_i = \sum_{i=p+1}^{n+m} \delta_i
\end{aligned}$$

$$m_b^f(\varphi) = m_b^f(\psi_A) = \max_{i \in \{1..p_A\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} = \max_{i \in \{1..p\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}$$

$$\begin{aligned}
m_a^f(\varphi) &= \max\{m_a^f(\psi_A), |\delta_a^f(\psi_A)| + m_a^f(\psi_B)\} \\
&= \max \left\{ \max_{i \in \{k_A+1..n\}} \left\{ \sum_{j=k_A+1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=k_A+1}^n |\delta_i| + \max_{i \in \{k_B+1..m\}} \left\{ \sum_{j=1}^{i-1} |\delta'_j| + m'_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{k+1..n\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=k+1}^n |\delta_i| + \max_{i \in \{1..m\}} \left\{ \sum_{j=1}^{i-1} |\delta'_j| + m'_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{k+1..n\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=k+1}^n |\delta_i| + \max_{i \in \{1..m\}} \left\{ \sum_{j=n+1}^{n+i-1} |\delta_j| + m'_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{k+1..n\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=k+1}^n |\delta_i| + \max_{i \in \{n+1..n+m\}} \left\{ \sum_{j=n+1}^{i-1} |\delta_j| + m_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{k+1..n\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}, \max_{i \in \{n+1..n+m\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\} \right\} \\
&= \max_{i \in \{k+1..n+m\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}
\end{aligned}$$

–  $I_B \neq \emptyset$

We obtain, in this case  $p = \max I_B = n + p_B$ . That is, the last call to  $f$  in the whole sequence is the last call to  $f$  in the sub-sequence  $[G_2 \rightarrow be'_1, \dots, be'_n]$ . Moreover, we know that  $L^f(\psi_B) \neq 1$

in this case, so we have to apply [Let2<sub>NC</sub>], instead of [Let1<sub>NC</sub>]. We obtain:

$$\begin{aligned}
\delta_b^f(\varphi) &= \delta_b^f(\psi_A) + \delta_a^f(\psi_A) + \delta_b^f(\psi_B) = \sum_{i=1}^{p_A} \delta_i + \sum_{i=p_A+1}^n \delta_i + \sum_{i=1}^{p_B} \delta'_i \\
&= \sum_{i=1}^n \delta_i + \sum_{i=n}^{n+p_B} \delta_i = \sum_{i=1}^{n+p_B} \delta_i = \sum_{i=1}^p \delta_i \\
\delta_a^f(\varphi) &= \delta_a^f(\psi_B) = \sum_{i=p_B+1}^m \delta'_i = \sum_{i=n+p_B+1}^{n+m} \delta_i = \sum_{i=p+1}^{n+m} \delta_i
\end{aligned}$$

$$\begin{aligned}
m_b^f(\varphi) &= \max \left\{ m_b^f(\psi_A), |\delta_b^f(\psi_A)| + m_a^f(\psi_A), |\delta_b^f(\psi_A)| + |\delta_a^f(\psi_A)| + m_b^f(\psi_B) \right\} \\
&= \max \left\{ \max_{i \in \{1..k_A\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \sum_{i=1}^{k_A} |\delta_i| + \max_{i \in \{k_A+1..n\}} \left\{ \sum_{j=k_A+1}^{i-1} |\delta_j| + m_i \right\}, \right. \\
&\quad \left. \sum_{i=1}^{k_A} |\delta_i| + \sum_{i=k_A+1}^n |\delta_i| + \max_{i \in \{1..k_B\}} \left\{ \sum_{j=1}^{i-1} |\delta'_j| + m'_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{1..k_A\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \max_{i \in \{k_A+1..n\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \right. \\
&\quad \left. \sum_{i=1}^n |\delta_i| + \max_{i \in \{n+1..n+k_B\}} \left\{ \sum_{j=n+1}^{i-1} |\delta_j| + m_i \right\} \right\} \\
&= \max \left\{ \max_{i \in \{1..k_A\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \max_{i \in \{k_A+1..n\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}, \right. \\
&\quad \left. \max_{i \in \{n+1..n+k_B\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \right\} \\
&= \max_{i \in \{1..n+k_B\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\} \\
&= \max_{i \in \{1..k\}} \left\{ \sum_{j=1}^{i-1} |\delta_j| + m_i \right\}
\end{aligned}$$

$$\begin{aligned}
m_a^f(\varphi) &= m_a^f(\psi_B) = \max_{i \in \{k_B+1..m\}} \left\{ \sum_{j=k_B+1}^{i-1} |\delta'_j| + m'_i \right\} \\
&= \max_{i \in \{k_B+1..m\}} \left\{ \sum_{j=n+k_B+1}^{n+i-1} |\delta_j| + m'_i \right\} \\
&= \max_{i \in \{n+k_B+1..n+m\}} \left\{ \sum_{j=n+k_B+1}^{i-1} |\delta_j| + m_i \right\} \\
&= \max_{i \in \{k+1..n+m\}} \left\{ \sum_{j=k+1}^{i-1} |\delta_j| + m_i \right\}
\end{aligned}$$

Therefore, (3) holds in both cases.

- **Case e**  $\equiv$  **case x of**  $\overline{C_i \bar{x}_{ij}} \rightarrow e_i$

Since the values  $N_b^f(\varphi)$ ,  $N_r^f(\varphi)$ ,  $L^f(\varphi)$ ,  $\delta_b^f(\varphi)$ ,  $\delta_a^f(\varphi)$ ,  $m_b^f(\varphi)$ ,  $m_a^f(\varphi)$  of the  $\varphi$  judgement are exactly the same as their counterparts in the judgement of the  $e_i$  being executed, the lemma follows trivially from the induction hypothesis applied to the latter judgement.  $\square$

Now we are ready to give a formal definition of a function  $nb$ ,  $nr$  and  $len$  being correct.

**Definition B.3.** Let us assume a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . We say that  $nr$  (resp.  $nb$  and  $len$ ) are correct approximations of the number of recursive calls (resp. base calls and height of the call tree), iff for every  $\varphi, E_f, h, k, td, h', v, \delta, m, s, \bar{v}_i^n, \bar{l}_j^m, \bar{s}_i^n$  such that:

1.  $\varphi \equiv E_f \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$ , where  $E_f = [\bar{x}_i \mapsto \bar{v}_i^n, \bar{r}_j \mapsto \bar{l}_j^m, self \mapsto k + 1]$ .
2. For each  $i \in \{1..n\}$ ,  $s_i = size(h, v_i)$

Then it holds that  $nr \bar{s}_i^n \geq N_r^f(\varphi)$  (resp.  $nb \bar{s}_i^n \geq N_b^f(\varphi)$  and  $len \bar{s}_i^n \geq L^f(\varphi)$ ).

## B.2 Correctness of the initial $\Delta_0$

The first step for proving the correctness of *computeDelta* is to express our  $\delta$  component resulting from the evaluation of a given function in terms of the charges done by the base and recursive cases, and the number of base and recursive calls done during that evaluation. The intuition behind this idea is the same as in Section 7.5.3. In order to compute the charges done by a recursive call, we need to be able to isolate the charges done by the call itself from the charges done by its subsequent recursive calls. This can be done with the help of an additional syntactic construct **dmask**, which resets the  $\delta$  component of a given expression. Its semantics are given by the following rule:

$$\frac{E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)}{E \vdash h, k, td, \mathbf{dmask} e \Downarrow h', k, v, ([ ]_k, m, s)}$$

It is easy to see that if we replace a function definition  $f \bar{x}_i @ \bar{r}_j = e_f$  in a signature  $\Sigma$  by another masked function definition  $f \bar{x}_i @ \bar{r}_j = \mathbf{dmask} e_f$  we will be able to obtain the same judgements as with our initial signature, but we will obtain a possibly different  $\delta$  component in the resource vector. The next theorem uses this function to define the charges done by a recursive call. The notation  $\Phi(\varphi)$  denotes the set of judgements belonging to the derivation of  $\varphi$  (including  $\varphi$  itself).

**Theorem B.4.** Assume a function definition  $f \bar{x}_i @ \bar{r}_j = e_f \in \Sigma$  such that the following execution takes place:

$$\varphi \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma} h', k + 1, v, (\delta, m, s)$$

Let us define  $\Sigma' = (\Sigma \setminus f) \uplus [f \mapsto f \bar{x}_i @ \bar{r}_j = \mathbf{dmask} e_f]$ , and assume the following execution under  $\Sigma'$ ,

$$\varphi' \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma'} h, k + 1, v, (\delta', m, s)$$

which is derivable by using the  $\Downarrow$ -rules. Given the following definitions:

$$\begin{aligned} \delta_{base} &= \max \{ \delta(\psi) \mid \psi \in \Phi(\varphi), Exp(\psi) = e_f, L^f(\psi) = 1 \} \\ \delta_{rec} &= \max \{ \delta(\psi) \mid \psi \in \Phi(\varphi'), Exp(\psi) = e_f, L^f(\psi) > 1 \} \end{aligned}$$

Then we get:

$$\delta|_k \leq \delta_{base}|_k * N_b^f(\varphi) + \delta_{rec}|_k * N_r^f(\varphi) \quad (\text{B.7})$$

*Proof.* By induction on  $L^f(\varphi)$ . Let us distinguish cases:

- **Case**  $L^f(\varphi) = 1$

In this case  $\delta_{base} \geq \delta$ . Since we assume absence of destruction, we get  $\delta_{rec}(i) \geq 0$  for each  $i \in \{1..n\}$ . Moreover,  $N_b^f(\varphi) \geq 1$  and  $N_r^f(\varphi) \geq 0$ , so:

$$\delta|_k \leq \delta_{base}|_k \leq \delta_{base}|_k * N_b^f(\varphi) \leq \delta_{base}|_k * N_b^f(\varphi) + \delta_{rec}|_k * N_r^f(\varphi)$$

- **Case**  $L^f(\varphi) > 1$

By Lemma B.2, there exists a sequence  $[G \rightarrow be_1, \dots, be_n]$  in seqs  $e_f$  and some  $\overline{E}_i^n, \overline{h}_i^n, \overline{td}_i^n, \overline{h}_i^n, \overline{v}_i^n, \overline{\delta}_i^n, \overline{m}_i^n, \overline{s}_i^n$  of their respective types such that:

$$\varphi_i \equiv E_i \vdash h_i, k, td_i, be_i \Downarrow_{\Sigma} h'_i, k, v_i, (\delta_i, m_i, s_i) \text{ belongs to the derivation of } \varphi$$

For each  $i \in \{1..n\}$ , since the derivation of  $\varphi'$  is the same as that of  $\varphi$ , but with different  $\delta$  components in the resource vectors, the following judgements belong to the derivation of  $\varphi'$ ,

$$\varphi'_i \equiv E_i \vdash h_i, k, td_i, be_i \Downarrow_{\Sigma'} h'_i, k, v_i, (\delta'_i, m_i, s_i)$$

with  $\delta' = \sum_{i=1}^n \delta'_i$ . Now we define the set  $I$  of indices such that the corresponding  $be_i$  is a function application to  $f$ . For every  $i \in \{1..n\}$ , if  $i \notin I$  we can execute  $be_i$  under  $\Sigma$  to obtain a judgement similar to  $\varphi'_i$ , since the latter does not contain calls to  $f$ , and the value of  $\Sigma(f)$  is not relevant to this execution. Thus we get:

$$E_i \vdash h_i, k, td_i, be_i \Downarrow_{\Sigma} h'_i, k, v_i, (\delta'_i, m_i, s_i)$$

But, since  $\varphi_i$  holds, and the computation of the resource vector is deterministic, we get  $\delta_i = \delta'_i$ . If  $i \in I$ , we know that there is an evaluation of **dmask**  $e_f$  "above" the judgement  $\varphi'_i$ , so we know that  $\delta'_i = [\ ]_k$  in these cases. Thus we get:

$$\sum_{i \notin I} \delta_i = \sum_{i \notin I} \delta'_i = \sum_{i \notin I} \delta'_i + \sum_{i \in I} \delta'_i = \sum_{i=1}^n \delta'_i = \delta' \leq \delta_{rec} \quad (\text{B.8})$$

Now assume an index  $i \in I$ , and consider again the corresponding  $\varphi_i$ . Above this judgement, we can find the execution of the body of the function, whose judgement we denote by  $\chi_i$ :

$$\chi_i \equiv E_{i,f} \vdash h_{i,f}, k + 2, td_{i,f}, e_f \Downarrow_{\Sigma} h'_{i,f}, k + 2, v_{i,f}, (\delta_{i,f}, m_{i,f}, s_{i,f})$$

There exists a counterpart  $\chi'_i$ , defined as follows:

$$\chi_i \equiv E_{i,f} \vdash h_{i,f}, k + 2, td_{i,f}, e_f \Downarrow_{\Sigma'} h'_{i,f}, k + 2, v_{i,f}, (\delta'_{i,f}, m_{i,f}, s_{i,f})$$

By induction hypothesis, we get, for each  $i \in I$ :

$$\delta_{i,f}|_{k+1} \leq \delta_{base,i}|_{k+1} * N_b^f(\chi_i) + \delta_{rec,i}|_{k+1} * N_r^f(\chi_i)$$

where  $\delta_{base,i}$  and  $\delta_{rec,i}$  are defined as follows:

$$\begin{aligned}\delta_{base,i} &= \max \{ \delta(\psi) \mid \psi \in \Phi(\chi_i), \text{Exp}(\psi) = e_f, L^f(\psi) = 1 \} \\ \delta_{rec,i} &= \max \{ \delta(\psi) \mid \psi \in \Phi(\chi'_i), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \}\end{aligned}$$

Since  $\Phi(\chi_i) \subseteq \Phi(\varphi)$  and  $\Phi(\chi'_i) \subseteq \Phi(\varphi')$ , we get  $\delta_{base,i} \leq \delta_{base}$  and  $\delta_{rec,i} \leq \delta_{rec}$ . Hence:

$$\delta_i|_k \leq \delta_{base}|_k * N_b^f(\varphi_i) + \delta_{rec}|_k * (N_r^f(\varphi_i) - 1) \quad (\text{B.9})$$

Therefore, we can prove the required result:

$$\begin{aligned}\delta|_k &= \sum_{i=1}^n \delta_i|_k && \{ \text{by Lemma B.2} \} \\ &= \sum_{i \notin I} \delta_i|_k + \sum_{i \in I} \delta_i|_k \\ &\leq \delta_{rec}|_k + \sum_{i \in I} \delta_i|_k && \{ \text{by (B.8)} \} \\ &\leq \delta_{rec}|_k + \sum_{i \in I} (\delta_{base}|_k * N_b^f(\varphi_i) + \delta_{rec}|_k * (N_r^f(\varphi_i) - 1)) && \{ \text{by (B.9)} \} \\ &= \delta_{base}|_k * \sum_{i \in I} N_b^f(\varphi_i) + \delta_{rec}|_k * (1 + \sum_{i \in I} (N_r^f(\varphi_i) - 1)) \\ &= \delta_{base}|_k * N_b^f(\varphi) + \delta_{rec}|_k * N_r^f(\varphi) && \{ \text{by Lemma B.2} \}\end{aligned}$$

□

Notice the similarity between the expression (B.7) and that occurring in the definition of *computeDelta*. The latter can be considered an “abstract” version of the former. The correctness of the result of *computeDelta* follows from this theorem. Before this, we need some technical result on the domain of the result given by the abstract interpretation function. In particular, it states the conditions under which this result is defined.

**Lemma B.5.** *Assume an execution  $\varphi \equiv E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)$  belonging to the derivation of the following context judgement:*

$$E_0 \vdash h_0, k, td_0, e_f \Downarrow h'_0, k, v_0, (\delta_0, m_0, s_0)$$

*corresponding to the function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f$ . Let us define, for each  $i \in \{1..n\}$ ,  $s_i = \text{size}(h, E_0(x_i))$ , and  $\llbracket e \rrbracket \Sigma \Gamma td = (\Delta, \mu, \sigma)$ , where  $\Sigma$  is an environment  $\Sigma$  containing correct signatures for the functions being called from  $f$ , and  $\Gamma$  is an environment typing  $e_f$ . If  $\text{closed}(h, E_0(x_i))$  for each  $i \in \{1..n\}$ , then  $\Delta \bar{s}_i^n \neq \perp$ ,  $\mu \bar{s}_i^n \neq \perp$ , and  $\sigma \bar{s}_i^n \neq \perp$ .*

*Proof.* (Sketch) By induction on the size of the  $\Downarrow$  derivation. All cases are straightforward. In the case of function applications, the guards of the resulting triple  $(\Delta, \mu, \sigma)$  hold because  $\Sigma$  is a correct signature. In **case** expressions, the guards are satisfied because the initial heap is closed, and closedness is propagated through the execution of a well-typed function. □

**Theorem B.6.** *Let  $\Delta = \text{computeDelta}(f \bar{x}_i @ \bar{r}_j = e_f) \Sigma \Gamma nb nr$ . If the following conditions hold:*

1.  $\Sigma$  is a correct signature for all the functions being called from  $f$ .
2.  $\Gamma \vdash e : s$  for some  $s \in \mathbf{SafeType}$ .
3.  $nb$  and  $nr$  are correct approximations of the number of base and recursive calls of  $f$ , respectively.
4. The abstract heaps  $\Delta_b$  and  $\Delta_r$  occurring in the definition of *computeDelta* are parameter-decreasing.

*Then  $\Delta$  is a correct abstract heap for  $f$ .*

*Proof.* (Sketch) It is a consequence of Theorem B.4, and the  $\Delta_b$  and  $\Delta_r$  being upper bounds of the  $\delta_{base}$  and  $\delta_{rec}$ , respectively. The abstract heap  $\Delta_b$  is a correct bound of  $\delta_{base}$  because of Theorem 7.26, whereas  $\Delta_r$  is a correct bound for  $\delta_{rec}$ , because  $[ ]_f$  is a correct bound to the heap charges of the function definition  $f \bar{x}_i^n @ \bar{r}_j^m = \mathbf{dmask} e_f$ . Assume a judgement:

$$\varphi \equiv E \vdash h, k, td, e_f \Downarrow h', k, v, (\delta, m, s)$$

Let us define, for each  $i \in \{1..n\}$ ,  $s_i = \text{size}(h, E(x_i))$ , and let us denote by  $\eta$  the region instantiation consistent with  $E$  and  $\Gamma$ . We get:

$$\Delta \sqsupseteq [\Delta_b] * nb + [\Delta_r] * nr \succeq_{\bar{s}_i^n, k, \eta} \delta_{base}|_k * N_b^f(\varphi) + \delta_{rec}|_k * N_r^f(\varphi) \geq \delta|_k$$

□

### B.3 Correctness of the initial $\mu_0$

For proving the correctness of the *computeMu* algorithm we follow a similar approach as in the previous section. In this case we need an additional construct **mmask**, which resets the  $m$  component of the resource vector:

$$\frac{E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)}{E \vdash h, k, td, \mathbf{mmask} e \Downarrow h', k, v, (\delta, 0, s)}$$

Similarly as above, the following theorem gives the memory needs as an expression which resembles the cost expression given by *computeDelta*. The difference is that it uses elements from the concrete domain, rather than the abstract one.

**Theorem B.7.** *Assume a function definition  $f \bar{x}_i @ \bar{r}_j = e_f \in \Sigma$  and the following judgement:*

$$\varphi \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma} h', k + 1, v, (\delta, m, s)$$

*Let us define  $\Sigma' = (\Sigma \setminus f) \uplus [f \mapsto f \bar{x}_i @ \bar{r}_j = \mathbf{mmask} e_f]$ , and assume the following execution under  $\Sigma'$ ,*

$$\varphi' \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma'} h, k + 1, v, (\delta, m', s)$$

*which is derivable by using the  $\Downarrow$ -rules. Given the following definitions:*

$$\begin{aligned} \delta_{self} &= \max \{ \delta_b^f(\psi)(k + 1) \mid \psi \in \Phi(\varphi'), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\ m_{bef} &= \max \{ m_b^f(\psi) \mid \psi \in \Phi(\varphi'), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\ m_{aft} &= \max \{ m_a^f(\psi) \mid \psi \in \Phi(\varphi'), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\ m_{base} &= \max \{ m(\psi) \mid \psi \in \Phi(\varphi), \text{Exp}(\psi) = e_f, L^f(\psi) = 1 \} \end{aligned}$$

*We get:*

$$m \leq |\delta_b^f(\varphi')|_k + \delta_{self} * (L^f(\varphi) - 1) + \max\{m_{bef}, m_{aft}, m_{base}\} \quad (\text{B.10})$$

*Proof.* By induction on the maximum number of nested calls,  $L^f(\varphi)$ . We distinguish cases:

- **Case**  $L^f(\varphi) = 1$ .

Then  $m \leq m_{base}$ . Moreover,  $|\delta_b^f(\varphi')|_k \geq 0$ , since we are assuming absence of explicit destruction. Hence we get:

$$m \leq m_{base} \leq |\delta_b^f(\varphi')|_k + \delta_{self} * 0 + \max\{m_{bef}, m_{aft}, m_{base}\}$$

- **Case**  $L^f(\varphi) > 1$ .

By Lemma B.2, there exists a sequence  $[G \rightarrow be_1, \dots, be_n]$  such that  $\varphi_i$  belongs to  $\Phi(\varphi)$ , where  $\varphi_i$  is defined as follows:

$$\varphi_i \equiv E_i \vdash h_i, k, td_i, be_i \Downarrow_{\Sigma} h'_i, k, v_i, (\delta_i, m_i, s_i)$$

for every  $i \in \{1..n\}$ , and for some  $E_i, h_i, td_i, h'_i, v_i, \delta_i, m_i$  and  $s_i$  of their respective types. Since the only difference between  $\Sigma$  and  $\Sigma'$  is that the latter resets the  $m$  component of each recursive call to  $f$ , there exists, for each of these  $\varphi_i$  judgements, a counterpart  $\varphi'_i$  belonging to  $\varphi'$ ,

$$\varphi'_i \equiv E_i \vdash h_i, k, td_i, be_i \Downarrow_{\Sigma'} h'_i, k, v_i, (\delta_i, m'_i, s_i)$$

for some  $m'_i$ . Let us denote by  $I$  the set of indices  $i$  such that  $be_i$  has the form  $f \bar{a}_i @ \bar{r}_j$  for some  $\bar{a}_i$  and  $\bar{r}_j$ . The difference between  $\Sigma$  and  $\Sigma'$  is not relevant in the judgements  $\varphi'_i$  when  $i \notin I$ , since there are no applications of  $f$  in their derivations. Then we can substitute  $\Sigma$  for  $\Sigma'$  in those  $\varphi'_i$ , and we would get the same resource vector  $(\delta_i, m'_i, s_i)$  as a result. Since the computation of this vector is deterministic, and because  $\varphi_i$  holds, we get  $m_i = m'_i$  whenever  $i \notin I$ .

Let us define  $p = \max I$ . By applying Lemma B.1 we know that

$$m = \max\{m_b^f(\varphi), |\delta_b^f(\varphi)| + m_a^f(\varphi)\}$$

We have to prove that each one of the expressions in the  $\max\{\dots\}$  is lower than the right-hand side of (B.10). Let us start with the second one. By Lemma (B.2) it holds that

$$\delta_b^f(\varphi) = \sum_{i=1}^p \delta_i = \delta_b^f(\varphi') \tag{B.11}$$

and, similarly,

$$m_a^f(\varphi) = \max_{i \in \{p+1..n\}} \left\{ \sum_{j=p+1}^{i-1} |\delta_j| + m_i \right\} = \max_{i \in \{p+1..n\}} \left\{ \sum_{j=p+1}^{i-1} |\delta_j| + m'_i \right\} = m_a^f(\varphi') \tag{B.12}$$

The second step can be done because  $i \notin I$  for every  $i \in \{p+1..n\}$ , by definition of  $I$ . Besides this, we know that  $\text{dom } \delta_b^f(\varphi') = \{0..k+1\}$ , so we can split  $|\delta_b^f(\varphi')|$  as follows:

$$|\delta_b^f(\varphi')| = |\delta_b^f(\varphi')|_k + \delta_b^f(\varphi')(k+1) \tag{B.13}$$

Hence we get:

$$\begin{aligned}
& |\delta_b^f(\varphi)| + m_a^f(\varphi) \\
= & |\delta_b^f(\varphi')| + m_a^f(\varphi') && \{ \text{by (B.11) and (B.12)} \} \\
= & |\delta_b^f(\varphi')|_k + \delta_b(\varphi')(k+1) + m_a^f(\varphi') && \{ \text{by (B.13)} \} \\
\leq & |\delta_b^f(\varphi')|_k + \delta_{self} + m_{aft} && \{ \text{by definition of } \delta_{self} \text{ and } m_{aft} \} \\
\leq & |\delta_b^f(\varphi')|_k + \delta_{self} * (L^f(\varphi) - 1) + m_{aft} && \{ \text{since } L^f(\varphi) > 1 \} \\
\leq & |\delta_b^f(\varphi')|_k + \delta_{self} * (L^f(\varphi) - 1) + \max\{m_{bef}, m_{aft}, m_{base}\}
\end{aligned}$$

and we are done. Now let us prove that  $m_b^f(\varphi)$  is lower than the right-hand side of (B.10). We can unfold, by Lemma B.2 the value of  $m_b^f(\varphi)$  as follows:

$$m_b^f(\varphi) = \max_{i \in \{1..p\}} \{M_i\}, \text{ where } \forall i \in \{1..p\}. M_i \stackrel{\text{def}}{=} \sum_{j=1}^{i-1} |\delta_j| + m_i$$

Thus, the required result amounts to proving that  $M_i$  is lower or equal than the right-hand side of (B.10), for each  $M_i \in \{1..p\}$ . So, let us assume  $i \in \{1..p\}$  and distinguish cases:

–  $i \notin I$

In this case it holds that:

$$M_i = \sum_{j=1}^{i-1} |\delta_j| + m_i = \sum_{j=1}^{i-1} |\delta_j| + m'_i \leq m_b^f(\varphi') \leq m_{bef}$$

and, hence,

$$M_i \leq m_{bef} \leq |\delta_b^f(\varphi')|_k + \delta_{self} * (L^f(\varphi) - 1) + \max\{m_{bef}, m_{aft}, m_{base}\}$$

since  $\delta_b^f(\varphi')(j) \geq 0$  for each  $j \in \{1..k\}$ ,  $\delta_{self} \geq 0$ , and  $L^f(\varphi) > 1$ .

–  $i \in I$

In this case,  $be_i$  is a function application. We get that the following judgement:

$$\chi_i \equiv E_{i,f} \vdash h_{i,f}, k + 2, td_{i,f}, e_f \Downarrow_{\Sigma} h'_{i,f}, k + 2, v_{i,f}, (\delta_{i,f}, m_{i,f}, s_{i,f})$$

belongs to the derivation of  $\varphi_i$ , and a analogous one  $\chi'_i$  belongs to  $\varphi'_i$ :

$$\chi'_i \equiv E_{i,f} \vdash h_{i,f}, k + 2, td_{i,f}, e_f \Downarrow_{\Sigma'} h'_{i,f}, k + 2, v_{i,f}, (\delta_{i,f}, m'_{i,f}, s_{i,f})$$

By induction hypothesis:

$$m_{i,f} \leq |\delta_b^f(\chi'_i)|_{k+1} + \delta_{self,i} * (L^f(\chi_i) - 1) + \max\{m_{bef,i}, m_{aft,i}, m_{base,i}\} \quad (\text{B.14})$$

where

$$\begin{aligned}
\delta_{self,i} &= \max \{ \delta_b^f(\psi)(k+1) \mid \psi \in \Phi(\chi'_i), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\
m_{bef,i} &= \max \{ m_b^f(\psi) \mid \psi \in \Phi(\chi'_i), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\
m_{aft,i} &= \max \{ m_a^f(\psi) \mid \psi \in \Phi(\chi'_i), \text{Exp}(\psi) = e_f, L^f(\psi) > 1 \} \\
m_{base,i} &= \max \{ m(\psi) \mid \psi \in \Phi(\chi_i), \text{Exp}(\psi) = e_f, L^f(\psi) = 1 \}
\end{aligned}$$

By Lemma B.1 (3) we know that  $\delta_b^f(\chi'_i) \leq \delta_{i,f}$ . Moreover, by the [App] rule,  $\delta_{i,f}|_{k+1} = \delta_i$ . Therefore:

$$|\delta_b^f(\chi'_i)|_{k+1} \leq |\delta_{i,f}|_{k+1} \leq |\delta_i| \quad (\text{B.15})$$

Besides this, the judgement  $\chi_i$  belongs to the derivation of  $\varphi$ , and  $\chi'_i$  belongs to the derivation of  $\varphi'$ . Thus  $\Phi(\chi_i) \subseteq \Phi(\varphi)$  and  $\Phi(\chi'_i) \subseteq \Phi(\varphi')$ , which implies:

$$\delta_{self,i} \leq \delta_{self} \quad m_{bef,i} \leq m_{bef} \quad m_{aft,i} \leq m_{aft} \quad m_{base,i} \leq m_{base} \quad (\text{B.16})$$

Finally, by the [App<sub>NC</sub>] rule, it holds that  $L^f(\varphi_i) = L^f(\chi_i) + 1$ . By combining (B.14), (B.15), (B.16) with this equality and the fact that  $m_i = m_{i,f}$ , we obtain:

$$m_i = m_{i,f} \leq |\delta_i| + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\}$$

Therefore:

$$\begin{aligned}
M_i &= \sum_{j=1}^{i-1} |\delta_j| + m_i \\
&\leq \sum_{j=1}^{i-1} |\delta_j| + |\delta_i| + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\} \\
&= \sum_{j=1}^i |\delta_j| + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\} \\
&\leq \sum_{j=1}^p |\delta_j| + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\} \quad \{ \text{since } i \leq p \}
\end{aligned}$$

The last step is justified by the fact that  $|\delta_i| \geq 0$  for each  $i \in \{1..p\}$ . By Lemma B.2 we know that  $\delta_b^f(\varphi') = \sum_{i=1}^p \delta_i$ , and, since  $\text{dom } \delta_b^f(\varphi') = \{0..k+1\}$  we get:

$$\begin{aligned}
M_i &\leq |\delta_b^f(\varphi')|_k + \delta_b^f(\varphi')(k+1) + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\} \\
&\leq |\delta_b^f(\varphi')|_k + \delta_{self} + \delta_{self} * (L^f(\varphi_i) - 2) + \max\{m_{bef}, m_{aft}, m_{base}\} \\
&= |\delta_b^f(\varphi')|_k + \delta_{self} * (1 + (L^f(\varphi_i) - 1) - 1) + \max\{m_{bef}, m_{aft}, m_{base}\} \\
&\leq |\delta_b^f(\varphi')|_k + \delta_{self} * (L^f(\varphi) - 1) + \max\{m_{bef}, m_{aft}, m_{base}\}
\end{aligned}$$

□

The correctness of *computeMu* is established by connecting the elements of the abstract domain appearing in its result with the elements of the concrete domain occurring in the previous theorem.

**Theorem B.8.** *Let  $\mu = \text{computeMu}(f \ \bar{x}_i @ \bar{r}_j = e_f) \ \Sigma \ \Gamma \ \Delta$  len. If the following conditions hold:*

1.  $\Sigma$  is a correct signature for all the functions being called from  $f$ .
2.  $\Gamma \vdash e : s$  for some  $s \in \mathbf{SafeType}$ .
3. *len* is a correct approximation of the maximum length of the call tree of  $f$ .
4.  $\Delta$  is a correct abstract heap for  $f$ .

5. The space cost functions  $\Delta_{self}$ ,  $\mu_{aft}$ , and  $\mu_{bef}$  occurring in the definition of `computeMu` are parameter-decreasing.

Then  $\mu$  is correct with respect to  $f$ .

*Proof.* (Sketch) The proof proceeds in a similar way as Corollary B.6. In this case, it follows from Theorem B.7. If we have a judgement:

$$\varphi \equiv E \vdash h, k, td, e_f \Downarrow h', k, v, (\delta, m, s)$$

and denote, for each  $i \in \{1..n\}$ ,  $s_i = \text{size}(h, E(x_i))$ , we obtain:

$$\mu \sqsupseteq \Delta_{self} * (len - 1) + |\Delta_{bef}| + \sqcup \{ \mu_{bef}, \mu_{aft}, \mu_b \} \succeq_{s_i^n} \delta_{self} * (L^f(\varphi) - 1) + |\delta_b^f(\varphi)|_k + \max\{m_{bef}, m_{aft}, m_{base}\} \geq m$$

□

## B.4 Correctness of the initial $\sigma_0$

The computation of the initial approximation to the stack costs relies on the concept of stack level, defined in Section 7.5.5. This level stands for the number of words existing in the stack at a given time. Neither the semantic rules of Figure 2.26 nor the rules we have defined in Figure B.1 is able to capture this information. Recall that the  $td$  component only counts the number of words from the top of the stack to the topmost continuation, whereas we are interested in the total number of words in the stack including those lying below that continuation. For this reason, we have to extend again the semantic rules of Figure 2.26 in order to include this information. Our judgements will have the following form:

$$E \vdash h, k, (td, s_0), e \Downarrow h', k, v, (\delta, m, s)$$

where  $s_0$  is the new component that counts the number of words in the stack. In Figure B.3 we show the rules defining this extension. Most of them are self-explaining. The  $s_0$  components mimics the  $td$  component in almost every expression, except when evaluating the bound expression of a **let**, where the  $td$  component becomes zero (as a new continuation has been pushed), and the  $s_0$  component is incremented by two (the number of stack words taken by a continuation).

We are particularly interested in the difference between the stack levels between two points: when a function starts, and when a recursive call is going to be done. That is what the  $SD_f$  function defined in Figure 7.15 computes. The following lemma shows how we can compute the stack costs of the execution of a given expression from these differences between stack levels, and the stack costs of each recursive call. Before this, we need to introduce the semantics of the **smask** expressions, which play a similar role as **dmask** and **mmask**.

$$\frac{E \vdash h, k, td, e \Downarrow h', k, v, (\delta, m, s)}{E \vdash h, k, td, \mathbf{smask} e \Downarrow h', k, v, (\delta, m, 0)}$$

**Lemma B.9.** *Let us assume the execution of an expression under an environment  $\Sigma$ , and that there are  $p$  direct calls to  $f$  in the corresponding execution (and hence  $p$  judgements evaluating  $e_f$ ). For each  $i \in \{1 \dots p\}$ , we*

$$\begin{array}{c}
\frac{}{E \vdash h, k, (td, s_0), c \Downarrow h, k, c, ([ ]_k, 0, 1)} \text{ [Lit}_{ST}\text{]} \\
\\
\frac{}{E[x \mapsto v] \vdash h, k, (td, s_0), x \Downarrow h, k, v, ([ ]_k, 0, 1)} \text{ [Var}_{ST}\text{]} \\
\\
\frac{}{E \vdash h, k, (td, s_0), a_1 \oplus a_2 \Downarrow h, k, E(a_1) \oplus E(a_2), ([ ]_k, 0, 2)} \text{ [PrimOp}_{ST}\text{]} \\
\\
\frac{j \leq k \quad (h', p') = \text{copy}(h, p, j) \quad m = \text{size}(h, p)}{E[x \mapsto p, r \mapsto j] \vdash h, k, (td, s_0), x @ r \Downarrow h', k, p', ([j \mapsto m]_k, m, 2)} \text{ [Copy}_{ST}\text{]} \\
\\
\frac{(g \bar{y}_i^n @ \bar{r}_j^l = e_g) \in \Sigma}{\frac{[y_i \mapsto E(a_i)^n, \bar{r}_j^l \mapsto E(r_j)^l, \text{self} \mapsto k+1] \vdash h, k+1, (n+l, s_0+n+l-td), e \Downarrow h', k+1, v, (\delta, m, s)}{E \vdash h, k, (td, s_0), g \bar{a}_i^n @ \bar{r}_j^l \Downarrow h' \mid_k, k, v, (\delta \mid_k, m, \max\{n+l, s+n+l-td\})} \text{ [App}_{ST}\text{]} \\
\\
\frac{E(r) = j \quad j \leq k \quad \text{fresh}_h(p)}{E \vdash h, k, (td, s_0), C \bar{a}_i^n @ r \Downarrow h \uplus [p \mapsto (j, C E(a_i)^n)], k, v, ([j \mapsto 1]_k, 1, 1)} \text{ [Cons}_{ST}\text{]} \\
\\
\frac{E \vdash h, k, (0, s_0+2), e_1 \Downarrow h', k, v_1, (\delta_1, m_1, s_1) \quad E \cup [x_1 \mapsto v_1] \vdash h', k, (td+1, s_0+1), e_2 \Downarrow h'', k, v, (\delta_2, m_2, s_2)}{E \vdash h, k, (td, s_0), \mathbf{let} \ x_1 = e_1 \ \mathbf{in} \ e_2 \Downarrow h'', k, v, (\delta_1 + \delta_2, \max\{m_1, |\delta_1| + m_2\}, \max\{2 + s_1, 1 + s_2\})} \text{ [Let}_{ST}\text{]} \\
\\
\frac{C = C_r \quad E(x) = p \quad E \uplus [\bar{x}_{rj} \mapsto \bar{v}_j^{nr}] \vdash h, k, (td + n_r, s_0 + n_r), e_r \Downarrow h', k, v, (\delta, m, s)}{E \vdash h[p \mapsto (j, C \bar{v}_i^{nr}), k, (td, s_0), \mathbf{case} \ x \ \mathbf{of} \ C_i \bar{x}_{ij}^{mi} \rightarrow e_i^n \Downarrow h', k, v, (\delta, m, s + n_r)]} \text{ [Case}_{ST}\text{]}
\end{array}$$

Figure B.3: Operational semantics of Core-Safe expressions with stack level information.

denote the initial stack size (resp. stack cost) of the  $i$ -th derivation by  $s_0^{(i)}$  (resp.  $s^{(i)}$ ):

$$\frac{\dots (td^{(1)}, s_0^{(1)}), e_f \Downarrow \dots (\delta^{(1)}, m^{(1)}, s^{(1)}) \quad \dots (td^{(p)}, s_0^{(p)}), e_f \Downarrow \dots (\delta^{(p)}, m^{(p)}, s^{(p)})}{\vdots \quad \dots \quad \vdots} \dots (td, s_0), e \Downarrow \dots (\delta, m, s)$$

In addition, we assume the execution of the same expression  $e$  under the environment  $\Sigma \setminus f \uplus [f \mapsto f \bar{x}_i @ \bar{r}_j = \mathbf{smask} e_f]$ , obtaining  $s'$  as the stack cost. Then the following relation between  $s$  and  $s'$  holds:

$$s = \max \left( \{s'\} \cup \{s_0^{(i)} - s_0 + s^{(i)} \mid i = 1 \dots p\} \right)$$

*Proof.* By induction on the structure of  $e$ .

- **Cases**  $e \equiv c, e \equiv x, e \equiv a_1 \oplus a_2, e \equiv C \bar{a}_i @ r, \mathbf{and} e \equiv g \bar{a}_i @ \bar{r}_j$  with  $g \neq f$ .

This is trivial, since  $p = 0$  and both executions with  $\Sigma$  and with  $\Sigma'$  produce the same stack costs. Hence  $s = s' = \max\{s'\}$

- **Case**  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$

There is a single recursive call ( $p = 1$ ):

$$\frac{\dots (n + m, s_0 + n + m - td), [\mathbf{smask}] e_f \Downarrow \dots (\dots, \dots, s^{(1)})}{\dots (td, s_0), f \bar{a}_i^n @ \bar{r}_j^m \Downarrow \dots (\dots, \dots, \max\{n + m, s^{(1)} + n + m - td\})}$$

Under  $\Sigma'$  we get  $s^{(1)} = 0$  and hence  $s' = \max\{n + m, 0 + n + m - td\} = n + m$ . Since  $s_0^{(1)} - s_0 = (s_0 + n + m - td) - s_0 = n + m - td$ , we get:

$$s = \max\{n + m, s^{(1)} + n + m - td\} = \max\{s', s_0^{(1)} - s_0 + s^{(1)}\}$$

- **Case**  $e \equiv \mathbf{let} x_1 = e_1 \mathbf{in} e_2$

We assume that there are  $q$  calls to  $f$  in the execution of  $e_1$  and  $p - q$  calls in the execution of  $e_2$ , with  $0 \leq q \leq p$ . We have got the following situation:

$$\frac{\frac{(q \text{ calls to } f)}{\dots, (0, s_0 + 2), e_1 \Downarrow \dots (\delta_1, m_1, s_1)} \quad \frac{(p - q \text{ calls to } f)}{\dots, (td + 1, s_0 + 1), e_2 \Downarrow \dots (\delta_2, m_2, s_2)}}{\dots, (td, s_0), \mathbf{let} x_1 = e_1 \mathbf{in} e_2 \Downarrow \dots (\delta, m, s)}$$

If we denote by  $s'$  (resp.  $s'_1, s'_2$ ) the costs of  $e$  (resp.  $e_1, e_2$ ) when evaluated under  $\Sigma'$ , we get:

$$\begin{aligned} s &= \max\{2 + s_1, 1 + s_2\} \\ &= \max \left\{ 2 + \max \left( \{s'_1\} \cup \{s_0^{(i)} - (s_0 + 2) + s^{(i)} \mid i \in \{1 \dots q\}\} \right), \right. \\ &\quad \left. 1 + \max \left( \{s'_2\} \cup \{s_0^{(j)} - (s_0 + 1) + s^{(j)} \mid j \in \{q + 1 \dots p\}\} \right) \right\} \quad \{ \text{by i.h.} \} \\ &= \max \left\{ \max\{2 + s'_1, 1 + s'_2\}, \max\{s_0^{(i)} - s_0 + s^{(i)} \mid i \in \{1 \dots q\}\}, \right. \\ &\quad \left. \max\{s_0^{(j)} - s_0 + s^{(j)} \mid j \in \{q + 1 \dots p\}\} \right\} \\ &= \max \left( \{s'\} \cup \{s_0^{(i)} - s_0 + s^{(i)} \mid i \in \{1 \dots p\}\} \right) \end{aligned}$$

- **Case**  $e \equiv \text{case } x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i}$

We assume that the  $r$ -th branch is executed and we denote by  $s_r$  and  $s'_r$  the costs associated to the derivations with  $\Sigma$  and  $\Sigma'$ .

$$\begin{aligned}
s &= n_r + s_r \\
&= n_r + \max \left( \{s'_r\} \cup \{s_0^{(i)} - (s_0 + n_r) + s^{(i)} \mid i \in \{1 \dots p\}\} \right) \quad \{ \text{by i.h.} \} \\
&= \max \left( \{n_r + s'_r\} \cup \{s_0^{(i)} - s_0 + s^{(i)} \mid i \in \{1 \dots p\}\} \right) \\
&= \max \left( \{s'_r\} \cup \{s_0^{(i)} - s_0 + s^{(i)} \mid i \in \{1 \dots p\}\} \right)
\end{aligned}$$

□

The following lemma shows that the  $SD_f$  function returns an upper bound to the above mentioned stack difference.

**Lemma B.10** (Correctness lemma for  $SD_f$ ). *We assume the execution  $E \vdash h, k, (td, s_0), e \Downarrow h', k, v, (\delta, m, s)$ , in which there are  $p \geq 1$  direct calls to a function  $f$ . For a given  $i \in \{1 \dots p\}$ , we denote by  $s_0^{(i)}$  the stack size before executing the  $e_f$  body corresponding to the  $i$ -th call of  $f$ :*

$$\begin{array}{c}
\frac{E' \vdash h_i, k + 1, (td_i, s_0^{(i)}), e_f \Downarrow h'_i, k + 1, v_i, (\delta_i, m_i, s_i) \quad \dots}{\dots} \\
\vdots \\
\frac{\dots}{E \vdash h, k, (td, s_0), e \Downarrow h', k, v, (\delta, m, s)}
\end{array}$$

Then:  $SD_f e \, td \geq s_0^{(i)} - s_0$

*Proof.* By induction on the size of the  $\Downarrow$ -derivation. Since the derivation contains a call to  $f$ , we assume that  $e$  contains a sub-expression  $f \bar{a}_i @ \bar{r}_j$ . This rules out the cases  $e \equiv c, x, a_1 \oplus a_2, C \bar{a}_i^n @ r$ , and  $g \bar{b}_i^q @ \bar{s}_j^q$  with  $g \neq f$ . We distinguish the remaining cases:

- **Case**  $e \equiv f \bar{a}_i^n @ \bar{r}_j^m$

We get  $s_0^{(i)} = s_0 + n + m - td$ , and hence,  $SD_f e \, td = n + m - td = s_0^{(i)} - s_0$

- **Case**  $e \equiv \text{let } x_1 = e_1 \text{ in } e_2$

If the  $i$ -th call to  $f$  is in  $e_2$  we get:

$$\begin{aligned}
SD_f e \, td &\geq 1 + (SD_f e_2 (td + 1)) \quad \{ \text{by definition of } SD_f \} \\
&\geq 1 + s_0^{(i)} - (s_0 + 1) \quad \{ \text{by i.h.} \} \\
&= s_0^{(i)} - s_0
\end{aligned}$$

If the  $i$ -th call to  $f$  is in  $e_1$  we get:

$$\begin{aligned}
SD_f e \, td &\geq 2 + (SD_f e_1 0) \quad \{ \text{by definition of } SD_f \} \\
&\geq 2 + s_0^{(i)} - (s_0 + 2) \quad \{ \text{by i.h.} \} \\
&= s_0^{(i)} - s_0
\end{aligned}$$

- **Case**  $e \equiv \text{case } x \text{ of } \overline{C_i \bar{x}_{ij}^{n_i} \rightarrow e_i^n}$

We assume the  $r$ -th branch being executed ( $1 \leq r \leq n$ ). Therefore:

$$\begin{aligned} SD_f e \text{ td} &\geq n_r + (SD_f e_r (td + n_r)) \quad \{ \text{definition of } SD_f \} \\ &\geq n_r + s_0^{(i)} - (s_0 + n_r) \quad \{ \text{by i.h.} \} \\ &= s_0^{(i)} - s_0 \end{aligned}$$

□

Given these results, we are ready to proceed as in previous sections: we find an expression for  $s$  that resembles its abstract counterpart given by *computeSigma*.

**Theorem B.11.** *Assume a function definition  $f \bar{x}_i^n @ \bar{r}_j^m = e_f \in \Sigma$  such that the following execution takes place:*

$$\varphi \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma} h', k + 1, v, (\delta, m, s)$$

Let us define  $\Sigma' = (\Sigma \setminus f) \uplus [f \mapsto f \bar{x}_i @ \bar{r}_j = \mathbf{smask} e_f]$ , and assume the following execution under  $\Sigma'$ :

$$\varphi' \equiv E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma'} h', k + 1, v, (\delta, m, s')$$

Given the following definition,

$$s_{max} = \max \{s(\psi) \mid \psi \in \Phi(\varphi')\}$$

we get,

$$s \leq \max\{0, SD e_f (n + m)\} * (L^f(\varphi) - 1) + s_{max} \quad (\text{B.17})$$

*Proof.* By induction on  $L^f(\varphi)$ . We distinguish cases:

- **Case**  $L^f(\varphi) = 1$

There are no function applications to  $f$  in the derivation of  $\varphi'$ . Hence the value of  $\Sigma'(f)$  is not relevant in that judgement, and we can substitute  $\Sigma$  for  $\Sigma'$  in  $\varphi'$ , so as to get:

$$E \vdash h, k + 1, td, e_f \Downarrow_{\Sigma} h', k + 1, v, (\delta, m, s')$$

But, because of Lemma 2.19, we get  $s = s'$ , which implies  $s \leq s_{max}$ . Therefore:

$$s \leq s_{max} \leq \max\{0, SD_f e_f (n + m)\} * 0 + s_{max}$$

- **Case**  $L^f(\varphi) > 1$

Let us assume there are  $p$  calls to  $f$  in the derivation of  $\varphi$ . We know that  $p \geq 1$  (otherwise,  $L^f(\varphi)$  would not be greater than 1). For each  $i \in \{1..p\}$  there exists a judgement

$$\chi_i \equiv E_i \vdash h_i, k + 2, (td_i, s_{0,i}), e_f \Downarrow h'_i, k + 2, v_i, (\delta_i, m_i, s_i)$$

in the derivation  $\varphi$ . Similarly, there exists another judgement

$$\chi'_i \equiv E_i \vdash h_i, k + 2, (td_i, s_{0,i}), e_f \Downarrow h'_i, k + 2, v_i, (\delta_i, m_i, s'_i)$$

in the derivation of  $\varphi'$ . Let us denote by  $\varphi_i$  (resp.  $\varphi'_i$ ) the judgements situated “below”  $\chi_i$  (resp.  $\chi'_i$ ) in the corresponding derivation tree. These judgements must represent the execution of a function

application of  $f$ . By Lemma B.9 we get:

$$s = \max\{\{s_{0,i} - s_0 + s_i \mid i \in \{1..p\}\} \cup \{s'\}\} \quad (\text{B.18})$$

But we know that  $s' \leq s_{max}$  by definition of  $s_{max}$ . By Lemma B.10 we get  $SD_f e_f (n+m) \geq s_{0,i} - s_0$  for each  $i \in \{1..p\}$ . Moreover, by induction hypothesis, we get, for each  $i \in \{1..p\}$ :

$$s_i \leq \max\{0, SD_f e_f (n+m)\} * (L^f(\chi_i) - 1) + s_{max,i}$$

where:

$$s_{max,i} = \max\{s(\psi) \mid \psi \in \Phi(\chi'_i), \text{Exp}(\psi) = e_f\}$$

Since  $\Phi(\chi'_i) \subseteq \Phi(\varphi')$ , we get  $s_{max,i} \leq s_{max}$ . Besides this,  $L^f(\varphi_i) = 1 + L^f(\chi_i)$ . Therefore:

$$s_i \leq \max\{0, SD_f e_f (n+m)\} * (L^f(\varphi_i) - 2) + s_{max}$$

Finally, we get  $1 + (L^f(\varphi_i) - 1) \leq L^f(\varphi)$ . Thus:

$$s_i \leq \max\{0, SD_f e_f (n+m)\} * (L^f(\varphi) - 2) + s_{max}$$

We can rewrite (B.18) as follows:

$$\begin{aligned} s &\leq \max\{\{SD_f e_f (n+m) + \max\{0, SD_f e_f (n+m)\} * (L^f(\varphi) - 2) + s_{max}\} \cup \{s_{max}\}\} \\ &\leq \max\{0, SD_f e_f (n+m)\} * (L^f(\varphi) - 1) + s_{max} \end{aligned}$$

which proves the Lemma. □

Finally, the required result follows from this theorem. We have to prove that  $\sigma$  actually approximates  $s_{max}$ .

**Theorem B.12.** *Let  $\sigma = \text{computeSigma}(f \bar{x}_i^n @ \bar{r}_j^m = e_f) \Sigma(n+m)$  len. If the following conditions hold:*

1.  $\Sigma$  is a correct signature for all the functions being called from  $f$ .
2. len is a correct approximation of the maximum length of the call tree of  $f$ .
3. The space cost function  $\sigma$  occurring in the definition of  $\text{computeSigma}$  is parameter-decreasing.

Then  $\sigma$  is correct with respect to  $f$ .

*Proof.* (Sketch) Similarly to Corollaries B.6 and B.8, it follows from Theorem B.11. Now  $\sigma$  is a correct bound to  $s_{max}$ , by Theorem (7.26). By assuming a judgement:

$$\varphi \equiv E \vdash h, k, td, e_f \Downarrow h', k, v, (\delta, m, s)$$

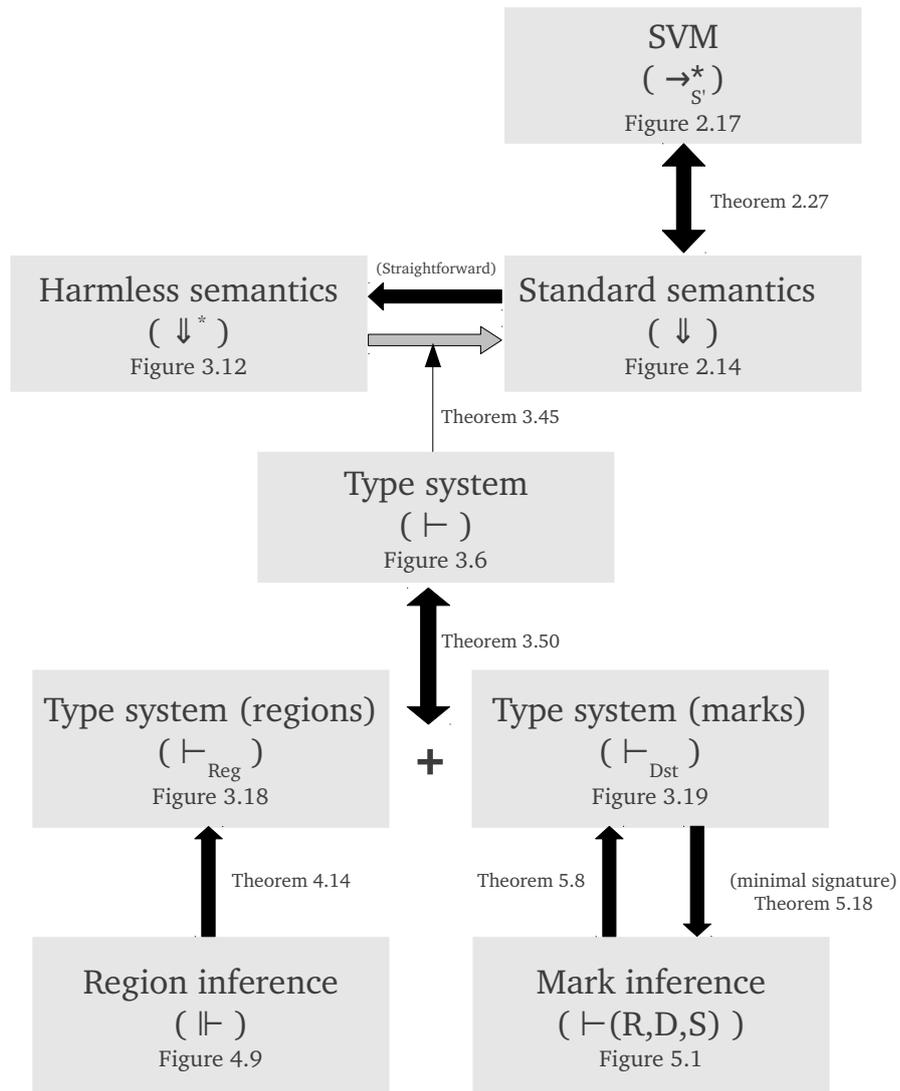
and defining, for each  $i \in \{1..n\}$ ,  $s_i = size(h, E(x_i))$ , we obtain:

$$\begin{aligned}
 \sigma &\sqsupseteq \sqcup\{0, SD_f e_f (n + m)\} * (len - 1) + \sigma \\
 &\succeq_{\vec{s}_i^n} \max\{0, SD_f e_f (n + m)\} * L^f(\varphi) + s_{max} \\
 &\geq s
 \end{aligned}$$

□

# Appendix C

## Map of semantic definitions and type systems





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