Certification of Safe Polynomial Memory Bounds*

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Abstract. In previous works, we have developed several algorithms for inferring upper bounds to heap and stack consumption for a simple functional language called *Safe*. The bounds inferred for a particular recursive function with n arguments takes the form of symbolic n-ary functions from $(\mathbb{R}^+)^n$ to $\mathbb{R}^+$ relating the input argument sizes to the number of cells or words respectively consumed in the heap and in the stack. Most frequently, these functions are multivariate polynomials of any degree, although exponential and other functions can be inferred in some cases.

Certifying memory bounds is important because the analyses could be unsound, or have been wrongly implemented. But the certifying process should not be necessarily tied to the method used to infer those bounds. Although the motivation for the work presented here is certifying the bounds inferred by our compiler, we have developed a certifying method which could equally be applied to bounds computed by hand.

The certification process is divided into two parts: (a) an off-line part consisting of proving the soundness of a set of proof rules. This part is independent of the program being certified, and its correctness is established once forever by using the proof assistant Isabelle/HOL; and (b) a compile-time program-specific part in which the proof rules are applied to a particular program and their premises proved correct.

The key idea for the first part is proving an Isabelle/HOL theorem for each syntactic construction of the language, relating the symbolic information asserted by the proof-rule to the dynamic properties about the heap and stack consumption satisfied at runtime. For the second part, we use a mathematical tool for proving instances of Tarski's decision problem on quantified formulas in real closed fields.

Keywords: Memory bounds, formal certificates, proof assistants, Tarski's decision problem.

1 Introduction

Certifying program properties consists of providing mathematical evidence about them. In a Proof Carrying Code (PCC) environment [17], these proofs should

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be checked by an appropriate tool. The certified properties may be obtained either manually, interactively, or automatically, but whatever is the effort needed for generating them, the PCC paradigm insists on their checking to be fully automatic.

In our setting, the certified property (safe memory bounds) is automatically inferred as the product of several static analyses, so that the certificate can be generated by the compiler without any human intervention. Certifying the inferred property is needed to convince a potential consumer that the static analyses are sound and that they have been correctly implemented in the compiler.

Inferring safe memory bounds in an automatic way is a complex task, involving in our case several static analyses:

- A region inference analysis [15] decides in which regions different data structures should be allocated, so that they could be safely destroyed when the region is deallocated. At the same time, the live memory is kept to a minimum (in other words, the analysis detects the maximum possible garbage).
- A size analysis infers upper bounds to the size of certain variables.
- A termination analysis [14] is used to infer upper bounds to the number of internal calls of recursive functions.
- A space inference analysis [16], uses the results of the above analyses to infer upper bounds to the heap and stack consumption.

Memory bounds could also be manually obtained, but in this case the computation must determine all the additive and multiplicative constants. This is usually a tedious and error-prone task.

But, once the memory bounds have been obtained, certifying them should be a simpler task. It is common folklore in the PCC framework that to find a proof is always more complex than to check it. A good example of this is ranking function synthesis in termination proofs of recursive and iterative programs. A ranking function is a kind of certificate or *witness* of termination. To find them is a rather complex task. Sometimes, linear methods [20] or sophisticated polyhedra libraries are used [10,1]. Others, more powerful methods such as SAT solvers [3] or non-linear constraint solvers [11] are needed. But, once the ranking function has been obtained, certifying termination consists of 'simply'[1] proving that it strictly decreases at each program transition in some well-founded order. This shows that the certifying and the inference processes are not necessarily tied.

In this paper we propose a simple way of certifying upper memory bounds whatever complex the method to obtain them has been. In the first part, we develop a set of syntax-driven proof-rules allowing to infer safe upper memory bounds to the execution of any expression, provided we have already upper bounds for its sub-expressions. Then we prove their soundness by relating the symbolic information inferred by a rule to the dynamic properties about the heap and stack consumption satisfied at runtime. In order to get complete confidence on the rules, we have used the Isabelle/HOL proof assistant [19] for this task.

[1] If the ranking function is not linear, proving that it decreases may be not so simple, and even it might be undecidable.

In the second part we explain how, given a candidate upper bound for a recursive function, the compiler can apply the proof-rules and infer a new upper bound, which will be correct provided the candidate upper bound is correct. Our main proof-rule states that if the derived bound *is smaller than or equal to the candidate one*, then both are correct. In order to certify this latter inequality, we propose to use a computer algebra tool for proving instances of Tarski's decision problem on quantified formulas involving polynomials over the real numbers [21]. To our knowledge, this is the first time that the described method is used to certify memory upper bounds.

The plan of the paper is as follows: after this introduction, in Sec. 2 we briefly summarize the characteristics and semantics of our functional language *Safe*; sections 3, 4, and 5 are devoted to presenting the proof-rules and to proving their soundness; Sec. 6 explains the certification process and how a symbolic algebra tool is used as a certificate checker; Sec. 7 presents a small case study illustrating the certificate generation and checking; finally, Sec. 8 presents some related work and draws the paper conclusions.

2 The Language

Safe is a first-order eager language with a syntax similar to Haskell's. Fig. 1 shows a mergesort algorithm written in *Full-Safe*. Its runtime system uses *regions*, i.e. disjoint parts of the heap where the program allocates data structures. They are automatically inferred [15] and made explicit in the intermediate language, called *Core-Safe*, and in the internal types. For instance, the types inferred for the functions of Fig. 1 are (the ρ's are region types):

$$\begin{aligned} \mathit{unshuffle} &:: [a]@\rho \to \rho_1 \to \rho_2 \to ([a]@\rho_1, [a]@\rho_1)@\rho_2 \\ \mathit{merge} &:: [a]@\rho \to [a]@\rho \to \rho \to [a]@\rho \\ \mathit{msort} &:: [a]@\rho' \to \rho \to [a]@\rho \end{aligned}$$

The meaning for e.g. *unshuffle* is that it receives a list in region ρ and two region arguments of types ρ_1, ρ_2. The first one is used to return the two result lists, and the second one for storing the tuple containing them. It is important to note that the number of regions a function may deal with can be statically determined. The *Core-Safe* versions of *merge* and *msort* can be seen in Sec. 7.

```
unshuffle []     = ([],[])
unshuffle (x:xs) = (x:ys2, ys1)
                   where (ys1,ys2) = unshuffle xs

merge []     ys  = ys
merge (x:xs) []  = x:xs
merge (x:xs) (y:ys) | x <= y    = x : merge xs     (y:ys)
                    | otherwise = y : merge (x:xs) ys

msort []  = []
msort [x] = [x]
msort xs  = merge (msort xs1) (msort xs2)
            where (xs1, xs2) = unshuffle xs
```

Fig. 1. *mergesort* algorithm in *Full-Safe*

$$E \vdash (h,k), td, c \Downarrow (h,k), c, ([\,]_k, 0, 1)\ [Lit]$$

$$E[x \mapsto v] \vdash (h,k), td, x \Downarrow (h,k), v, ([\,]_k, 0, 1)\ [Var]$$

$$\frac{j \le k \quad fresh(p)}{E[\overline{a_i \mapsto v_i}^n, r \mapsto j] \vdash (h,k), td, C\ \overline{a_i}^n\ @\ r \Downarrow (h \uplus [p \mapsto (j, C\ \overline{v_i}^n)], k), p, ([j \mapsto 1]_k, 1, 1)}\ [Cons]$$

$$\frac{(f\ \overline{x_i}^n\ @\ \overline{r_j}^l = e) \in \Sigma_D \quad [\overline{x_i \mapsto E(a_i)}^n, \overline{r_j \mapsto E(r'_j)}^l, self \mapsto k+1] \vdash (h,k+1), n+l, e \Downarrow (h',k+1), v, (\delta, m, s)}{E \vdash (h,k), td, f\ \overline{a_i}^n\ @\ \overline{r'_j}^l \Downarrow (h'|_k, k), v, (\delta|_k, m, \max\{n+l, s+n+l-td\})}\ [App]$$

$$\frac{E \vdash (h,k), td, a_1 \Downarrow (h,k), v_1, ([\,]_k, 0, 1) \quad E \vdash (h,k), td, a_2 \Downarrow (h,k), v_2, ([\,]_k, 0, 1)}{E \vdash (h,k), td, a_1 \oplus a_2 \Downarrow (h,k), v_1 \oplus v_2, ([\,]_k, 0, 2)}\ [Primop]$$

$$\frac{\begin{array}{c} E \vdash (h,k), 0, e_1 \Downarrow (h',k), v_1, (\delta_1, m_1, s_1) \\ E \cup [x_1 \mapsto v_1] \vdash (h',k), td+1, e_2 \Downarrow (h'',k), v, (\delta_2, m_2, s_2) \end{array}}{E \vdash (h,k), td, \mathbf{let}\ x_1 = e_1\ \mathbf{in}\ e_2 \Downarrow (h'',k), v, (\delta_1 + \delta_2, \max\{m_1, |\delta_1| + m_2\}, \max\{2+s_1, 1+s_2\})}\ [Let]$$

$$\frac{C = C_r \quad E \cup [\overline{x_{r_i} \mapsto v_i}^{n_r}] \vdash (h,k), td + n_r, e_r \Downarrow (h',k), v, (\delta, m, s)}{E[x \mapsto p] \vdash (h[p \mapsto (j, C\ \overline{v_i}^n)], k), td, \mathbf{case}\ x\ \mathbf{of}\ \overline{C_i\ \overline{x_{ij}}^{n_i} \to e_i}^n \Downarrow (h',k), v, (\delta, m, s + n_r)}\ [Case]$$

Fig. 2. Resource-Aware Operational semantics of *Core-Safe* expressions

The smallest memory unit is the *cell*, a contiguous memory space big enough to hold a data construction. A cell contains the constructor identity and a representation of the free variables to which the constructor is applied. These may consist either of basic values, or of pointers to other constructions. Each cell is allocated at constructor application time. A *region* is a collection of cells. It is created empty and it may grow up while it is active. Region deallocation frees all its cells. The allocation and deallocation of regions are associated with function calls: a *working region*, denoted by the reserved identifier *self*, is allocated when entering the call, and deallocated upon exiting it. Inside the function, data structures not belonging to the output may be built there.

Fig. 2 shows the *Core-Safe* big-step semantic rules, with extra annotations added in order to obtain the resources used by evaluating an expression. The expressions are self-explained and typical of most first-order functional languages. A judgement of the form $E \vdash (h,k), td, e \Downarrow (h',k), v, (\delta, m, s)$ means that expression e is evaluated in an environment E mapping variables to values, and in a heap (h,k) with $0 \ldots k$ active regions. As a result, a heap (h',k) and a value v are obtained, and a resource vector (δ, m, s), explained below, is consumed. Argument td refers to the number of positions used by E in the abstract machine stack, and it plays a role in rule *App*.

We denote data constructors by C, constants by c, variables by x, and atoms —an atom is either a constant or a variable— by a. Σ_D is a global environment containing all the function definitions. By $h|_k$ we denote the heap h with all regions above k deleted. A heap (h,k) contains a mapping h between pointers p and constructor cells $(j, C\ \overline{v_i}^n)$, where j, $0 \le j \le k$, is the cell region. The allocation and deallocation of regions is apparent in rule *App*.

The first component of the resource vector is a partial function $\delta : \mathbb{N} \to \mathbb{N}$ giving for each active region i the difference between the cells in the final and initial heaps. By $dom\ \delta$ we denote the subset $\{0 \ldots k\}$ in which δ is defined. By $[\,]_k$ we denote the function $\lambda i \in \{0 \ldots k\}.0$. By $|\delta|$ we mean the sum $\sum_{i \in dom\ \delta} \delta\ i$ giving the total balance of cells. The remaining components m and s respectively give the *minimum* number of fresh cells in the heap and of words in the stack

needed to successfully evaluate e, i.e. the peak memory used during e's evaluation. When e is the main expression, these figures give us the total memory needs of a particular run of the *Safe* program.

3 Function Signatures

A *Core-Safe* function f is defined as an $n+m$ argument expression:

$$f :: t_1 \to \ldots \to t_n \to \rho_1 \to \ldots \to \rho_m \to t$$
$$f\ x_1 \cdots x_n\ @\ r_1\ \cdots r_m = e_f$$

where $r_1\ \cdots r_m$ are the region arguments. A function may charge space costs to heap regions and to the stack. In general, these costs depend on the *sizes* of the function arguments. We define the size of an algebraic type term to be the number of cells of its recursive spine. This is always at least 1. We define the size of a Boolean value to be zero. However, for an integer argument we choose its size to be its value because frequently space costs depend on the value of a numeric argument. As a consequence, all the costs and sizes of a function f can be expressed as functions on f's argument sizes:

$$\mathbb{F}_f = \{\xi : (\mathbb{R}^+ \cup \{+\infty\})^n \to \mathbb{R}^+ \cup \{+\infty, -\infty\} \ \mid\ \xi \text{ is monotonic}\}$$

Cost or size $+\infty$ are used to represent that the analysis is not able to infer a bound, while $-\infty$ is used to express that the cost or size is not defined. For instance, the following function, where xs is assumed to be a list size,

$$\lambda xs. \begin{cases} xs - 3 & \text{if } xs \geq 4 \\ -\infty & \text{otherwise} \end{cases}$$

is undefined for sizes xs smaller than 4 (i.e. for lists with less than 3 elements).

They are ordered as expected, $-\infty \leq 0$, and $\forall x \in \mathbb{R}^+ . x \leq +\infty$, so $-\infty \sqcup x = x$ and $+\infty \sqcup x = +\infty$. Arithmetic monotonic operations with $\pm\infty$ are defined as follows, where $x \in \mathbb{R}^+$ while $y \in \mathbb{R}^+ \cup \{+\infty, -\infty\}$:

$$-\infty + y = -\infty \quad -\infty * y = -\infty \quad +\infty + x = +\infty \quad +\infty * x = +\infty$$

The domain of cost functions $(\mathbb{F}_f, \sqsubseteq, \bot, \top, \sqcup, \sqcap)$ is a complete lattice with the usual order $\sqsubseteq$ between functions. The rest of the components are standard. Notice that it is closed under the operations $\{+, \sqcup, *\}$.

Function f above may charge space costs to a maximum of $m+1$ regions: it may create cells in any output region $r_1 \ldots r_m$, and additionally in its *self* region. Each region r has a region type. We denote by R_f the set $\{\rho_1 \ldots \rho_m\}$ of argument region types, and by ρ^f_{self} the type of region *self* of function f.

Looked at from outside, the charges to the *self* region are not visible, so we define $\mathbb{D}_f = \{\Delta \mid \Delta : R_f \to \mathbb{F}_f\}$ as the complete lattice of functions describing the space costs charged by f to visible regions. We will call *abstract heaps* to these functions.

Definition 1. *A function* signature *for f is a triple $(\Delta_f, \mu_f, \sigma_f)$, where Δ_f belongs to $\mathbb{D}_f$, and μ_f, σ_f belong to $\mathbb{F}_f$.*

The aim is that Δ_f is an upper bound to the cost charged by f to visible regions, (i.e. to the increment in *live memory* due to a call to f), and μ_f, σ_f respectively are upper bounds to the heap and stack *peaks* contributed by f's evaluation.

4 Proof-Rules

When dealing with an expression e, we assume it belongs to the body e_f of a function definition $f\ \overline{x_i}^n\ @\ \overline{r_j}^m = e_f$, that we will call the *context function*, assumed to be well-typed.

We consider available a local type environment θ giving the types of all (free and bound) variables in e_f. It allows to type e_f and all its sub-expressions. Also a local environment ϕ giving for every (free and bound) variable its size as a symbolic function of the sizes of f's formal arguments $\overline{x_i}^n$. Finally, a *global type environment* Σ_T giving for every function and data constructor of the program their most general types.

Let Σ be a *global signature environment* giving, for each *Safe* function g in scope, its signature $(\Delta_g, \mu_g, \sigma_g)$, and let td (abbreviation of *top-depth*) be a natural number. This is a quantity used by the compiler to control the size of the runtime environment stored in the stack (it is the same argument used in the operational semantics, see Sec. 2). It has an impact on the stack consumption and so it will be needed in our judgements.

We inductively define a *derivation* relation as a set of proof-rules. The intended meaning of a judgement of the form $\theta, \phi, td \rhd_f e, \Sigma \vdash (\Delta, \mu, \sigma)$ is that Δ, μ, σ are safe upper bounds for respectively the live heap contributed by evaluating the expression e, the additional peak heap needed by e, and its additional peak stack. The context information needed is: a valid global signature environment Σ, two valid local environments θ (for types) and ϕ (for sizes), a runtime environment top depth td, and the name f of the context function.

In Figure 3 we show the proof-rules for the most relevant *Core-Safe* expressions. Predicate $def(\xi)$ expresses that the size ξ is defined according to its type: if ξ has an algebraic type, $def(\xi) \equiv \xi \geq 1$; if it is an integer, $def(\xi) \equiv \xi \geq 0$; otherwise $def(\xi) \equiv True$. We use the guarded notation $[G \to \xi]$, as equivalent to ξ if G holds, and to $-\infty$ otherwise. By $[\,]_f$ we denote the constant function $\lambda\rho \in R_f \cup \{\rho^f_{self}\}\,.\,\lambda\overline{x_i}^n\,.\,[\wedge_{i=1}^n def(x_i) \to 0]$, and by $[\rho' \to \xi]_f$ we denote:

$$\lambda\rho \in R_f \cup \{\rho^f_{self}\}\,.\,\lambda\overline{x_i}^n\,.\,\begin{cases}[\wedge_{i=1}^n def(x_i) \to 0] & \text{if } \rho \neq \rho' \\ [\wedge_{i=1}^n def(x_i) \to \xi] & \text{if } \rho = \rho'\end{cases}$$

We abbreviate $\lambda\overline{x_i}^n\,.\,[\wedge_{i=1}^n def(x_i) \to c]$ by c, when $c \in \mathbb{R}^+$. By $|\Delta|$ we mean $\sum_{\rho \in dom\ \Delta} \Delta\ \rho$.

Rules $[Lit]$, $[Var]$, $[Primop]$ and $[Cons]$ exactly reflect the corresponding resource-aware semantic rules shown in Fig. 2.

When a function application $g\ \overline{a_i}^l\ @\ \overline{r_j}^q$ is found, its signature $\Sigma\ g$ is applied to the sizes of the actual arguments, $\overline{\phi\ a_i\ \overline{x_j}^n}^l$. Some different region types of g may instantiate to the same actual region type of f. This type instantiation mapping $\psi : R_g \to R_f \cup \{\rho^f_{self}\}$ is provided by the compiler, and we will require it to be consistent with the typing environment θ and with the actual region arguments of the application. We call this property *argument preserving* and denote it as $argP(\psi, \overline{\rho_j}^q, \theta, \overline{r_j}^q)$. In essence it says that the types of the actual region arguments $\overline{r_j}^q$ given by θ coincide with the formal region types $\overline{\rho_j}^q$ of g after being instantiated by ψ.

The memory consumed by g in the formal regions mapped by ψ to the same f's actual region must be accumulated in order to get the charge to this region of f. In the $[App]$ rule of Figure 3, function $instance_f$ converts an abstract heap for g into an abstract heap for f. We define $instance_f(\Delta_g, \psi, \overline{a_i}^l)$ as the abstract heap Δ with domain $R_f \cup \{\rho^f_{self}\}$ such that:

$$\forall \rho \in dom\ \Delta\ .\ \Delta\ \rho = \lambda\ \overline{x_i}^n\ .\ [G\ (\overline{a_i}^n) \to \sum_{\rho' \in R_g \wedge \psi\ \rho' = \rho} \Delta_g\ \rho'\ (\overline{\phi\ a_i\ \overline{x_i}^n}^l)]$$

where $G\ (\overline{a_i}^n) \equiv \bigwedge_{i=1}^{l} def(\phi\ a_i\ \overline{x}^n)$. Notice that if any of the sizes $\phi\ a_i\ \overline{x_i}^n$ is not defined, then Δ_g applied to it is neither defined. It is easy to see that Δ, μ and σ defined in rule $[App]$ are monotonic. If $\exists i \in \{1 \ldots l\}\ .\ \neg def(\phi\ a_i\ \overline{x_i}^n)$, then Δ, μ and σ return $-\infty$, which guarantees monotonicity since $-\infty$ is the smallest value in the domain. For the rest of the arguments, monotonicity is guaranteed by the monotonicity of Δ_g, μ_g and σ_g.

Rule $[Let]$ reflects the corresponding resource-aware semantic rule, while rule $[Case]$ uses the least upper bound operator $\bigsqcup$ in order to obtain an upper bound to the cost of any of the branches.

In Fig. 4 we show the proof rule for recursive functions. In fact, it could also be applied to non-recursive ones. By $\lfloor \Delta \rfloor$ we denote the projection of Δ over R_f, obtained by removing the region ρ^f_{self} from Δ. This rule is the most relevant

$$\theta, \phi, td \rhd_f c, \Sigma \vdash ([\,]_f, 0, 1) \quad [Lit]$$

$$\theta, \phi, td \rhd_f x, \Sigma \vdash ([\,]_f, 0, 1) \quad [Var]$$

$$\theta, \phi, td \rhd_f a_1 \oplus a_2, \Sigma \vdash ([\,]_f, 0, 2) \quad [Primop]$$

$$\theta, \phi, td \rhd_f C\ \overline{a_i}^n\ @\ r, \Sigma \vdash ([\theta\ r \mapsto 1]_f, 1, 1) \quad [Cons]$$

$$\frac{\begin{array}{ccc}\Sigma\ g = (\Delta_g, \mu_g, \sigma_g) & G\ (\overline{a_i}^n) \equiv \bigwedge_{i=1}^{l} def(\phi\ a_i\ \overline{x}^n) & argP(\psi, \overline{\rho_j}^q, \theta, \overline{r_j}^q) \\ \mu = \lambda \overline{x}^n.[G\ (\overline{a_i}^n) \to \mu_g\ (\overline{\phi\ a_i\ \overline{x}^n}^l)] & \sigma = \lambda \overline{x}^n.[G\ (\overline{a_i}^n) \to \sigma_g\ (\overline{\phi\ a_i\ \overline{x}^n}^l)] & \Delta = instance_f(\Delta_g, \psi, \overline{a_i}^l)\end{array}}{\theta, \phi, td \rhd_f g\ \overline{a_i}^l\ @\ \overline{r_j}^q, \Sigma \vdash (\Delta, \mu, \sqcup\{l + q, \sigma + l + q - td\})} \ [App]$$

$$\frac{\theta, \phi, 0 \rhd_f e_1, \Sigma \vdash (\Delta_1, \mu_1, \sigma_1) \quad \theta, \phi, td + 1 \rhd_f e_2, \Sigma \vdash (\Delta_2, \mu_2, \sigma_2)}{\theta, \phi, td \rhd_f \mathbf{let}\ x_1 = e_1\ \mathbf{in}\ e_2, \Sigma \vdash (\Delta_1 + \Delta_2, \sqcup\{\mu_1, |\Delta_1| + \mu_2\}, \sqcup\{2 + \sigma_1, 1 + \sigma_2\})} \ [Let]$$

$$\frac{(\forall i)\ \theta, \phi, td + n_i \rhd_f e_i, \Sigma \vdash (\Delta_i, \mu_i, \sigma_i)}{\theta, \phi, td \rhd_f \mathbf{case}\ x\ \mathbf{of}\ \overline{C_i\ \overline{x_{ij}}^{n_i} \to e_i}^n, \Sigma \vdash (\bigsqcup_{i=1}^{n} \Delta_i, \bigsqcup_{i=1}^{n} \mu_i, \bigsqcup_{i=1}^{n} (\sigma_i + n_i))} \ [Case]$$

Fig. 3. Proof-rules for *Core-Safe* expressions

$$\frac{(f\ \overline{x_i}^l @\ \overline{r_j}^q = e_f) \in \Sigma_D \quad \theta, \phi, l + q \rhd_f e_f, \Sigma \uplus \{f \mapsto (\Delta, \mu, \sigma)\} \vdash (\Delta', \mu', \sigma') \quad (\lfloor \Delta' \rfloor, \mu', \sigma') \sqsubseteq (\Delta, \mu, \sigma)}{\theta, \phi, l + q \rhd_f e_f, \Sigma \vdash (\Delta', \mu', \sigma')}\ [Rec]$$

Fig. 4. Proof-rule for a (possibly) recursive *Core-Safe* function definition

contribution of the paper, since it reduces proving upper memory bounds to checking inequalities between functions over the real numbers. In words, its says that if a triple (Δ, μ, σ) (obtained by whatever means) is to be proved a safe upper bound for the recursive function f, a sufficient condition is:

1. Introduce (Δ, μ, σ) in the environment Σ as a candidate signature for f.
2. By using the remaining proof-rules, derive a triple (Δ', μ', σ') as a new upper bound for f's body.
3. Prove $(\lfloor \Delta' \rfloor, \mu', \sigma') \sqsubseteq (\Delta, \mu, \sigma)$.

The rule asserts that (Δ', μ', σ') is a correct bound for e_f without any assumption for f in Σ. By deleting the *self* region, then $(\lfloor \Delta' \rfloor, \mu', \sigma')$ is a correct signature for f, and so will it be (Δ, μ, σ), which is greater than or equal to it.

The first two steps are routine tasks. The only difficulty remaining is proving the third. As we will see in Sec. 6, for polynomial functions this can be done by converting it into a decision problem of Tarski's theory of closed real fields.

5 Soundness Theorems

Let $f\ \overline{x_i}^n\ @\ \overline{r_j}^m = e_f$, be the context function and θ, ϕ the local type and size environments for f. The steps we shall follow in this section are: (1) we shall introduce a notion of semantic satisfaction of a memory bound by an expression; (2) we shall define what a valid signature and a valid signature environment are; (3) we shall refine the semantic satisfaction to a conditional one subject to the validity of a global signature environment; and (4) we shall prove that the proof-rules of figures 3 and 4 are sound with respect to the given semantic notions.

Definition 2. *Given a pointer p belonging to a heap h, the following function returns the number of cells in h of the data structure starting at p:*

$$size(h[p \mapsto (j, C\ \overline{v_i}^l)], p) = 1 + \sum_{i \in RecPos\ C} size(h, v_i)$$

where $RecPos\ C$ denotes the recursive positions of constructor C, given by the global type environment Σ_T. We agree that $size(h, c) = 0$ for any constant c, except if c is an integer argument of f. In that case, $size(h, c) = c$.

For example, if p points to the first cons cell of the list $[1, 2, 3]$ in the heap h then $size(h, p) = 4$.

At runtime, a region type ρ becomes mapped to an active region $i \in \{0 \ldots k\}$. Let us call $\eta :: R_f \cup \{\rho^f_{self}\} \to \{0 \ldots k\}$ to this mapping. Our type system guarantees that the following property holds across any evaluation:

Definition 3. *Assuming that k denotes the topmost region of a given heap, we say that the mapping η is* admissible*, denoted admissible η k, if:*

$$\rho^f_{self} \in dom\ \eta \wedge \eta\ \rho^f_{self} = k \wedge \forall \rho \in (dom\ \eta) - \{\rho^f_{self}\}\ .\ \eta\ \rho\ < k$$

The type system also guarantees a consistency property between types and values that we will not define formally here[2]. By *consistent* $\theta\ \eta\ E\ h$ we mean that the types given by θ to the free variables, the values of these variables in the runtime environment E and heap h, and the mapping η, do not contradict each other.

Finally, by $valid_f\ \theta\ \phi$ we intuitively mean that each variable of the body e_f has a type in θ, and a symbolic size in ϕ as a function of the domain $\mathbb{F}_f$. Moreover, in any evaluation of e_f, the types are consistent with the values and the region mapping η, this one is admissible, and the symbolic sizes are upper bounds of the corresponding runtime sizes.

The semantic satisfaction of a memory bound by an expression is defined as follows: whenever θ, ϕ are valid environments, and some minor static and dynamic properties hold, then (Δ, μ, σ) is a correct bound for the memory consumption of expression e in any of its possible evaluations.

Definition 4. *Let $f\ \overline{x_i}^n\ @\ \overline{r_j}^m = e_f$ be the context function, and e a subexpression of e_f. We say that e satisfies the bound (Δ, μ, σ) in the context of θ, ϕ, and td, denoted $\theta, \phi, td \rhd_f e \models [\![(\Delta, \mu, \sigma)]\!]$, if:*

$$valid_f\ \theta\ \phi \rightarrow P_{dom} \wedge (\forall E\ h\ k\ h'\ v\ \eta\ \delta\ m\ s\ \overline{s_i}^n\ .\ P_{\Downarrow} \wedge P_{dyn} \wedge P_{size} \wedge P_\eta \rightarrow P_\Delta \wedge P_\mu \wedge P_\sigma)$$

$$\begin{array}{ll}
P_{dom} & \stackrel{def}{=}\ dom\ \Delta = R_f \cup \{\rho^f_{self}\} \\
P_{\Downarrow} & \stackrel{def}{=}\ E \vdash (h,k), td, e \Downarrow (h',k), v, (\delta, m, s) \\
P_{dyn} & \stackrel{def}{=}\ (\overline{x_i}^n \cup fv\ e \cup \overline{r_j}^m \cup self) \subseteq dom\ E \wedge dom\ \eta = dom\ \Delta \\
P_{size} & \stackrel{def}{=}\ \forall i \in \{1..n\}\ .\ s_i = size(h, E\ x_i) \\
P_\eta & \stackrel{def}{=}\ admissible(\eta, k) \\
P_\Delta & \stackrel{def}{=}\ \forall j \in \{0 \ldots k\}\ .\ \sum_{\eta\ \rho = j}\ \Delta\ \rho\ \overline{s_i}^n \geq \delta\ j \\
P_\mu & \stackrel{def}{=}\ \mu\ \overline{s_i}^n \geq m \\
P_\sigma & \stackrel{def}{=}\ \sigma\ \overline{s_i}^n \geq s
\end{array}$$

Definition 5. *A global bound environment Σ is* valid*, denoted $\models \Sigma$, if it belongs to the following inductively defined set:*

1. $\models \emptyset$*, i.e. the empty environment is always valid.*
2. *If $\models \Sigma$, and $(f\ \overline{x_i}^l\ @\ \overline{r_j}^m = e_f) \in \Sigma_D$, and there exist Δ, μ, σ such that for any valid local environments θ and ϕ, property $\theta, \phi, (l + m) \rhd_f e_f \models [\![(\Delta, \mu, \sigma)]\!]$ holds, then $\models \Sigma \uplus \{f \mapsto (\lfloor \Delta \rfloor, \mu, \sigma)\}$.*

[2] `http://dalila.sip.ucm.es/safe` provides an extended version of this paper with all the formal definitions and hand-written proofs; `http://dalila.sip.ucm.es/safe/bounds` provides the Isabelle/HOL proof-scripts of the lemmas of this section.

When proving a memory bound for an expression, we will usually need a valid global environment in order to get from it correct signatures for the functions called by the expression. We will then say that the satisfaction of the bound is *conditional* to the validity of the environment.

Definition 6. *An expression* e conditionally satisfies *a bound* (Δ, μ, σ) *with respect to a signature environment* Σ*, in the context of* θ, ϕ*, and* td*, denoted* $\theta, \phi, td \rhd_f e, \Sigma \models [\![(\Delta, \mu, \sigma)]\!]$*, if* $\models \Sigma \to \theta, \phi, td \rhd_f e \models [\![(\Delta, \mu, \sigma)]\!]$.

We are now in a position to state and prove the main theorem establishing that the proof-rules of figures 3 and 4 are sound.

Theorem 1 (Soundness)

$$\textit{If}\ \ \theta, \phi, td \rhd_f e, \Sigma \vdash (\Delta, \mu, \sigma),\ \ \textit{then}\ \ \theta, \phi, td \rhd_f e, \Sigma \models [\![(\Delta, \mu, \sigma)]\!]$$

The proof of the theorem is rather involved (around 4500 Isabelle/HOL lines). We sketch here the main steps:

1. We define a restricted big-step semantics with an upper bound n to the longest chain of f's recursive calls: $E \vdash (h, k), td, e \Downarrow_{f,n} (h', k), v, (\delta, m, s)$.
2. We prove that $E \vdash (h, k), td, e \Downarrow (h', k), v, (\delta, m, s)$ if and only if $\exists n\,.\, E \vdash (h, k), td, e \Downarrow_{f,n} (h', k), v, (\delta, m, s)$.
3. We define appropriate notions of satisfaction $\theta, \phi, td \rhd_f e \models_{f,n} [\![(\Delta, \mu, \sigma)]\!]$, validity $\models_{f,n} \Sigma$, and conditional validity $\theta, \phi, td \rhd_f e, \Sigma \models_{f,n} [\![(\Delta, \mu, \sigma)]\!]$ in which the longest chain of f's recursive calls is bounded by n.
4. We prove:

$$\begin{array}{ll} \forall n\,.\, \theta, \phi, td \rhd_f e \models_{f,n} [\![(\Delta, \mu, \sigma)]\!] & \equiv \theta, \phi, td \rhd_f e \models [\![(\Delta, \mu, \sigma)]\!] \\ \forall n\,.\, \models_{f,n} \Sigma & \equiv \models \Sigma \\ \forall n\,.\, \theta, \phi, td \rhd_f e, \Sigma \models_{f,n} [\![(\Delta, \mu, \sigma)]\!] & \Rightarrow \theta, \phi, td \rhd_f e, \Sigma \models [\![(\Delta, \mu, \sigma)]\!] \end{array}$$

5. By induction on the $\vdash$ derivation, and by cases on the last rule applied, we prove: $\theta, \phi, td \rhd_f e, \Sigma \vdash (\Delta, \mu, \sigma) \Rightarrow \forall n\,.\, \theta, \phi, td \rhd_f e, \Sigma \models_{f,n} [\![(\Delta, \mu, \sigma)]\!]$.

6 Certification

The proof-rules presented in Sec. 4 are valid whatever are the monotonic functions considered for describing sizes and costs. However, for certification purposes we restrict ourselves to the smaller class of monotonic **Max-Poly** functions:

Definition 7. *The class* ***Max-Poly*** *over* $\overline{x}^n$ *is the smallest set of expressions containing constants in* $\mathbb{R}^+$*, variables* $y \in \overline{x}^n$*, and closed under the operations* $\{+, *, \sqcup\}$*. We will call any element of* ***Max-Poly*** *a max-poly.*

We will call a max-poly function to a function of the form $\lambda \overline{x}^n.p$ in $(\mathbb{R}^+)^n \to \mathbb{R}^+$, where p is a max-poly over $\overline{x}^n$.

Notice that all the three operations are commutative and associative, and that $+$ and $*$ distribute over $\sqcup$ in $\mathbb{R}^+$. The latter makes that any max-poly can be normalized to a form $p_1 \sqcup \ldots \sqcup p_n$, where all the p_i are ordinary polynomials. This property extends also to max-poly functions.

In our case and disregarding $+\infty$ (which in fact means absence of a bound), the size and cost functions return a value in $\mathbb{R}^+ \cup \{-\infty\}$. As they are monotonic, in each dimension i they return $-\infty$ in some (possibly empty) interval $[0..k_i)$, and when $(\forall i \,.\, x_i \geq k_i)$ they return a value greater than or equal to 0. This property can be expressed by a Boolean guard on the x_i. Inspired by this, we restrict our elementary functions to have the form $[G\rightarrow f]$, where G is a guard of the form $\bigwedge_{i=1}^{n}(p_i \geq k_i)$, $k_i \in \mathbb{R}^+$, and all the $p_i(\overline{x}^n)$ and $f(\overline{x}^n)$ are multivariate max-polys over the set $\overline{x}^n$ of variables. The meaning of this notation, which we will call *atomic guarded function* (AGF in what follows), is:

$$[G\rightarrow f] \stackrel{\text{def}}{=} \lambda\overline{x}^n \,.\, \begin{cases} -\infty & \text{if } \neg G(\overline{x}^n) \\ f(\overline{x}^n) & \text{if } G(\overline{x}^n) \end{cases}$$

Operating with AGFs satisfies the following properties (a, b, c denote AGFs):

1. $[G_1\rightarrow f_1] + [G_2\rightarrow f_2] = [G_1 \wedge G_2\rightarrow f_1 + f_2]$
2. $[G_1\rightarrow f_1] * [G_2\rightarrow f_2] = [G_1 \wedge G_2\rightarrow f_1 * f_2]$
3. $[G_1\rightarrow [G_2\rightarrow f]] = [G_1 \wedge G_2\rightarrow f]$
4. $(a \sqcup b) + c = (a + c) \sqcup (b + c)$
5. $(a \sqcup b) * c = (a * c) \sqcup (b * c)$

As a consequence, any function obtained by combining AGFs with $\{+, *, \sqcup\}$ can be normalized to:

$$[G_1\rightarrow f_1] \sqcup \ldots \sqcup [G_l\rightarrow f_l]$$

We will call it a *normalized AGF set*. Now, coming back to the proof-rules of figures 3 and 4, if we introduce in the environment Σ of the *Rec* rule a triple (Δ, μ, σ) consisting of normalized AGF sets, and then derive a triple (Δ', μ', σ'), the latter can be expressed also as normalized AGF sets. This is because the operations involved in the remaining proof-rules are $\{+, *, \sqcup\}$, and the instantiations of the *App* rule. The latter consists of substituting max-polys for variables inside a max-poly. The result will also be a max-poly.

So, the check $(\lfloor\Delta'\rfloor, \mu', \sigma') \sqsubseteq (\Delta, \mu, \sigma)$ of the *Rec* rule reduces to checking inequalities of the form:

$$[G_1\rightarrow f_1] \sqcup \ldots \sqcup [G_l\rightarrow f_l] \sqsubseteq [G'_1\rightarrow f'_1] \sqcup \ldots \sqcup [G'_m\rightarrow f'_m]$$

Assuming that all the AGFs are functions over $\overline{x}^n$, this is in turn equivalent to:

$$\forall\overline{x}^n \,.\, \bigwedge_{i=1}^{l} \bigvee_{j=1}^{m} [G_i\rightarrow f_i] \sqsubseteq [G'_j\rightarrow f'_j]$$

Then, the elementary operation is comparing two AGFs. This can be expressed as follows:

$$[G\rightarrow f] \sqsubseteq [G'\rightarrow f'] = G \rightarrow (G' \wedge f \leq f')$$

The comparison $f \leq f'$ consists of comparing two max-polys of the form $p_1 \sqcup \ldots \sqcup p_r$ and $q_1 \sqcup \ldots \sqcup q_s$, which we can decide by applying again the same idea:

$$f \leq f' = \bigwedge_{i=1}^{r} \bigvee_{j=1}^{s} p_i \leq q_j$$

Summarizing, to decide $(\lfloor \Delta' \rfloor, \mu', \sigma') \sqsubseteq (\Delta, \mu, \sigma)$ we generate first-order formulas in Tarski's theory of real closed fields [21]. It is well known that this theory is decidable, although the existent algorithms are not efficient at all. For instance, Collins' quantifier elimination algorithm [9], which is recognized to be a great improvement over the original Tarski's procedure, has still a worst case complexity polynomial in the maximum degree of the involved polynomials and doubly exponential in the number of quantified variables. It is implemented in several symbolic algebra tools such as Mathematica. We have used the QEPCAD system built by Collins' group [7] which contains an improved version of original Collins' algorithm.

Fortunately, the number of quantified variables in our case is the number of arguments of the *Safe* function being certified, and this is usually very small, typically from one to three. So for practical purposes the QEPCAD system, or a similar tool, can be used as a certificate checker. The *Safe* compiler is used, not only to generate the initial triple (Δ, μ, σ) for every *Safe* function, but also to derive the triple (Δ', μ', σ'), to normalize both, and eventually to generate the proof obligations in the form of Tarski's formulas. For the moment, the compiler and the QEPCAD system have not been directly connected and some manual intervention is required.

7 Case Study

In Fig. 5 we show the *Core-Safe* versions of the algorithms `merge` and `msort`, in which regions are explicit. We will explain in detail how the proof-rules are

```
merge x y @ r = case x of
                 []    -> y
                 ex:x' -> case y of
                           []    -> x
                           ey:y' -> let c = ex <= ey in
                                    case c of
                                     True  -> let z1 = merge x' y @ r in
                                              ex:z1 @ r
                                     False -> let z2 = merge x y' @ r in
                                              ey:z2 @ r

msort x @ r = case x of
               []    -> x
               ex:x' -> case x' of
                         []  -> x
                         _:_ -> let (x1,x2) = unshuffle x @ self self in
                                let z1      = msort x1 @ r            in
                                let z2      = msort x2 @ r            in
                                merge z1 z2 @ r
```

Fig. 5. functions *merge* and *msort* in *Core-Safe*

applied to `merge` (a simple example which produces an uninteresting linear Tarski problem), and then we will show in less detail the process for `msort` (which produces a more interesting quadratic one). Let us assume that the candidate memory bound obtained by the *Safe* compiler for *merge* live heap, assuming $\theta\ r = \rho$, is:

$$\begin{aligned} \Delta_{merge}\ \rho = {} & [x \geq 2 \wedge y \geq 1 \rightarrow x + y - 2] && \text{-- } A \\ & \sqcup\ [x \geq 1 \wedge y \geq 2 \rightarrow x + y - 2] && \text{-- } B \\ & \sqcup\ [x \geq 1 \wedge y \geq 1 \rightarrow 0] && \text{-- } C \end{aligned}$$

This signature gives 0 cells when both lists are empty, i.e. $x = 1 \wedge y = 1$, and $x + y - 2$ cells otherwise.

Now, we introduce this signature in the environment Σ and apply the proof rules of Fig. 3. Remember that the *Cons* proof-rule gets $[x \geq 1 \wedge y \geq 1 \rightarrow 1]$ charged to region ρ. This is because, for a list l, $def(l) \equiv l \geq 1$. The *Let* rule asks for the addition of the involved Δ's, and the *Case* one for $\sqcup$ of the branches. Also, the sizes of the internal call arguments are $x' = x - 1$ and $y' = y - 1$, because of the pattern matching. All in all, we obtain as derived bound the following function:

$$\begin{aligned} \Delta'_{merge}\ \rho = {} & [x \geq 1 \wedge y \geq 1 \rightarrow 0] \\ & \sqcup\ [x \geq 1 \wedge y \geq 1 \rightarrow 0] \\ & \sqcup\ (([x - 1 \geq 2 \wedge y \geq 1 \rightarrow x - 1 + y - 2] \sqcup [x - 1 \geq 1 \wedge y \geq 2 \rightarrow x - 1 + y - 2] \\ & \quad \sqcup\ [x - 1 \geq 1 \wedge y \geq 1 \rightarrow 0]) + [x \geq 1 \wedge y \geq 1 \rightarrow 1]) \\ & \sqcup\ (([x \geq 2 \wedge y - 1 \geq 1 \rightarrow x + y - 1 - 2] \sqcup [x \geq 1 \wedge y - 1 \geq 2 \rightarrow x + y - 1 - 2] \\ & \quad \sqcup\ [x \geq 1 \wedge y - 1 \geq 1 \rightarrow 0]) + [x \geq 1 \wedge y \geq 1 \rightarrow 1]) \end{aligned}$$

The first two terms correspond to the branches ending in a variable, so the *Var* rule applies. The other two correspond to the branches having internal calls, by applying the rules *App*, *Cons*, and *Let*. The primitive operator $\leq$ adds a trivial term not shown. After normalization and simplification, we get:

$$\begin{aligned} \Delta'_{merge}\ \rho = {} & [x \geq 1 \wedge y \geq 1 \rightarrow 0] && \text{-- } C' \\ & \sqcup\ [x \geq 3 \wedge y \geq 1 \rightarrow x + y - 2] \sqcup [x \geq 2 \wedge y \geq 1 \rightarrow 1] && \text{-- } A' \sqcup A'' \\ & \sqcup\ [x \geq 2 \wedge y \geq 2 \rightarrow x + y - 2] && \text{-- } D' \\ & \sqcup\ [x \geq 1 \wedge y \geq 3 \rightarrow x + y - 2] \sqcup [x \geq 1 \wedge y \geq 2 \rightarrow 1] && \text{-- } B' \sqcup B'' \end{aligned}$$

Obviously, for all x, y we get $C' \sqsubseteq C$, $A' \sqsubseteq A$, $B' \sqsubseteq B$, and both $D' \sqsubseteq A$ and $D' \sqsubseteq B$. It is also easy to convince ourselves that A'' is dominated by A and B'' is dominated by B. Then, the inequality $\lfloor \Delta'_{merge} \rfloor \sqsubseteq \Delta_{merge}$ holds.

The candidate `msort` live memory bound inferred by our compiler, assuming Δ_{merge} as above, and the following bound obtained for *unshuffle*:

$$\Delta_{unshuffle} = \begin{bmatrix} \rho_1 \mapsto [x \geq 2 \rightarrow x + 1] \sqcup [x \geq 1 \rightarrow 2] \\ \rho_2 \mapsto [x \geq 2 \rightarrow x] \sqcup [x \geq 1 \rightarrow 1] \end{bmatrix}$$

is

$$\Delta_{msort}\ \rho = [x \geq 2 \rightarrow \frac{4}{3}x^2 - 3x] \sqcup [x \geq 1 \rightarrow 0]$$

Introducing this candidate bound in the environment, applying the proof-rules, normalizing, and simplifying lead to:

$$\Delta'_{msort} = \begin{bmatrix} \rho \mapsto [x \geq 3 \rightarrow \frac{2}{3}x^2 - \frac{3}{2}x - \frac{17}{6}] \sqcup [x \geq 1 \rightarrow 0] \\ \rho_{self} \mapsto [x \geq 2 \rightarrow 2x + 1] \sqcup [x \geq 1 \rightarrow 3] \end{bmatrix}$$

Notice that the charges to the *self* region are not needed in the comparison $\lfloor \Delta'_{msort} \rfloor \sqsubseteq \Delta_{msort}$. The relevant inequality is then:

$$\forall x\,. \quad \ldots \left(x \geq 3 \rightarrow x \geq 2 \wedge (\frac{2}{3}x^2 - \frac{3}{2}x - \frac{17}{6} \leq \frac{4}{3}x^2 - 3x) \right) \ldots$$

When this formula is given to QEPCAD, it answers *True* in about 100 msec. Then, $\lfloor \Delta'_{msort} \rfloor \sqsubseteq \Delta_{msort}$ holds.

8 Related Work and Conclusion

A seminal paper on static inference of memory bounds is [13]. A special type inference algorithm generates a set of linear constraints which, if satisfiable, they build a safe linear bound on the heap consumption. Afterwards, the authors extended this work to certificate generation [4], the certificate being an Isabelle/HOL proof-script which in essence was a proof of correctness of the type system, specialized for the types of the program being certified.

One of the authors extended in [12] the type system of [13] in order to infer polynomial bounds. Although not every polynomial could be inferred by this system, the work was a remarkable step forward in the area. They do not pay attention to certificates in this paper but there is an occasional comment on that the same ideas of [4] could be applied here.

In [8] an abstract interpretation based algorithm for controlling that memory is not allocated inside loops in Java programs is verified by using the Coq proof-assistant [5]. Here there is no program-specific certificate, but a general proof of correctness of the analysis algorithm.

With respect to our proof-rules, they clearly have an abstract interpretation flavour, and that is the reason why the lattice points above or equal to the fix-point of the interpretation are correct solutions. For recursive functions, we have adapted to our framework the technique first explained in [18]. This technique has also been used in other works (see e.g. [2]) where procedure global environments occur, and recursive procedures must be verified. The main idea is to explicitly introduce the depth of recursive call chains in the environment, and then doing some form of induction on this depth.

We have found inspiration on some work on quasi-interpretations for characterizing the complexity classes of rewriting systems [6], where **Max-Poly** plays a role. The existence of a quasi-interpretation belonging to **Max-Poly** is used to decide that some systems are in the classes PTIME or PSPACE. They show that the problem is decidable by generating formulas in first-order Tarski's theory. The formulas are existentially quantified and they assert the existence of a quasi-interpretation, although no attempt to synthesize one is done.

Our work finds for the first time a way of separating the bound inference problem from the certification one. We have shown that certification need not be a kind of proof of correctness of the inference algorithm. The *Rec* proof-rule and the idea of certifying bounds by checking an inequality $P \sqsubseteq Q$ between polynomial-like functions should work as well for other languages admitting syntax-driven proof-rules monotonic in a complete lattice. It could then be applied to languages, such as the functional one used in [13,12], where other algorithms are used to compute the candidate bounds. Additionally, our language deals with the memory deallocation due to the region mechanism. Most of other approaches infer and/or certify bounds to the *total* allocated memory, as opposed to the *live* and *peak* memory, respectively reached after and during program evaluation.

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