Implicit Computational Complexity

Simone Martini

Dipartimento di Scienze dell'Informazione Università di Bologna Italy

Bertinoro International Spring School for Graduate Studies in Computer Science, 6–17 March, 2006



Outline: third part

Logic and Programming Languages

Challenges



From Logic to Programming Languages

- How can a host machine assure the amount of resource needed to run a mobile program? A resource-aware type system or program-logic would provide implicit and verifiable certificates.
- In the realm of (first-order) term-rewriting systems, techniques like quasi interpretations have been shown to be useful for inferring complexity properties of programs (Bonfante et al.).
- Type-systems derived from non-size increasing computations have been exploited in the context of mobile resource guarantees (Hofmann et al., Beringer et al.).
- Enforcing resource-awareness in programming languages is not an easy task. The additional control provided cannot come at the price of unacceptable restrictions to programs.



Inferring Linear Bounds on Heap Size – Hofmann & Jost

- Language: first-order functional programming language with recursion; explicit memory management with freelist.
- Type-system: simple types, including lists, with resource annotations.
- Goal of the resource annotation is to derive a linear relation between the memory used to represent the input and the memory needed to complete the task.
- Example: Consider a program P : string list -> unit
 - We want a linear relation s(n) = an + b with the following meaning:
 - ▶ If we evaluate (the compiled) P on a input list of lenght n
 - Then, the program will not get stuck from insufficient memory availability
 - Provided that we have a freelist containing initially at least s(n) cells.



Inferring Linear Bounds on Heap Size, II

- The example of the previous slide would get a type
 - P : L(string,a),b -> unit
- "If the input list has lenght n, then P needs an + b cells in the freelist"
- ▶ In general, we need memory assertions also in the result type
- Example: $x : L(B,2), 3 \vdash e : L(B,4), 5$

means

- if we evaluate e starting with x bound to a list $[u_1, \ldots, u_m]$,
- and we have a free-list of at least 2m + 3 cells,
- then the computation will not get stuck from insufficient memory availability;
- ▶ moreover, if the result is a list [v₁,..., v_n], then at the end the free-list will have at least 4n + 5 cells.



Inferring Linear Bounds on Heap Size, II

► **Type-system**: Contraction can only be done splitting the corresponding resource annotations: for example, from

$$x: L(B,3), y: L(B,6) \vdash e: C, 7$$

we can derive

$$z: L(B,9) \vdash e\{z/x, z/y\}: C, 7$$

Decorations: given a skeleton of a type derivation (types, but not resource annotations) for e, a set of linear inequalities L(e) is derived.
 Solutions to L(e) are in one-to-one correspondence with valid type derivations for e.



An example: Mobile Resource Guarantees

- In 2002-2005 a EU funded project tried to embed some of the techniques we discussed in a software architecture.
- MRG, a joint Edinburgh / LMU Munich project funded under the Global Computing pro-active initiative
- ▶ Based on the notion of proof carrying code (Necula, 1997):
 - A high-level functional language with a type system ensuring certain bounds on resources
 - A certifying compiler maps programs and their type annotation to a target language, packaging together the code and a (compact version of the) proof that it satifies the required bounds
 - Such packages are unforgeable and tamper evident
 - Clients of the code (e.g., over an untrusted network) receive the package and check the proof before executing the code
 - Checking proof is simple (vs building the proof, which may be hard)



The architecture of MRG





Camelot

- Camelot is a high-level functional language, based on OCaml
- Polymorphic types à la ML
- Compiled (through Grail) into standard Java bytecode
- Memory model: freelist, managed directly by the compiled code (as opposed to just rely on garbage collection)
- Programs in Camelot are subjected to space analysis, to express heap usage and linear relations between input/output memory usage



Example

```
type iList = !Nil | Cons of int * iList
let ins a | - match | with
               Nil -> Cons(a,Nil)
             | Cons(x,t)Q_{-} -> if a < x then Cons(a,Cons(x,t))
                                else Cons(x, ins a t)
let sort | = match | with
              Nil -> Nil
            | Cons(a,t) -> ins a (sort t)
let show_list0 | = match | with
                    Nil -> ""
                  | Cons(h,t) -> begin
                      match t with
                        Nil -> string_of_int h
                      | Cons(h0,t0) -> (string_of_int h) ^ ", " ^ (show_list0 t)
                    end
let show_list I = "[" ^ (show_list0 I) ^ "]"
let stringList_to_intList ss =
       match ss with
         \Pi -> Nil
       |(h::t) -> Cons((int_of_string h),(stringList_to_intList t))
let start args =
       let |1 = (stringList_to_intList args)
     in let _ = print_string ("\nlnput list:\n l1 = " ^ (show_list l1))
     in let |2 = sort |1
     in let _ = print_string ("\nResult list:\n |2 = " (show_list |2\rangle)
     in ()
```

Fig. 1. A standalone Camelot program

Fig. 2. Output of space analysis on the program in I



The architecture of MRG





Grail

- It is the target of the Camelot compiler, which performs a resource exact compilation That is, compilation preserves non only meaning, but also resource behaviour.
- It is the vehicle for proof-carrying code:
 - It is the basis to which to attach the resource assertions
 - It is amenable to formal proofs about resource usage
 - It is the format for sending and receiving guaranteed code
- It can be assembled to (and dissambled from) standard JVM classfiles



Bytecode logic of resources

- The logic allowing to state and prove that the Grail bytecode satify the resource usage
- The construction of proofs uses the type annotations
- Verification is much easier
- But we are not concerned here with this issues...



At the end of this series of lectures...

Many challenges remain...



Challenges

- The area of implicit computational complexity appears very fragmented, with many different proposals.
- It is very difficult to compare relative intensional expressive power.
- It is not usually the case a system can be extended with new features preserving its quantitative properties
- Defining just another characterization of polynomial time is not enough.
- Importing these results into the design of (even academic) programming languages is extremely difficult (especially for time bounds).
- ▶ Deep, foundational results are extremely needed.





