

Verification of (probabilistic) functional programs

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MABioS team seminar
June 3rd, 2019

Introduction

- **Type theory**: allows to label parts of a program to prove properties about it.
- **Model-checking**: abstract a program as a model, and (try to) prove automatically properties about it.
- Both will meet in this talk, to allow the verification of **functional programs**, in which functions can take functions as inputs.

Advantages of functional programs

- **Very mathematical**: calculus of functions.
- ... and thus very much studied from a mathematical point of view. This notably leads to **strong typing**, a marvellous feature.
- Much **less error-prone**: no manipulation of global state.

More and more used, from Haskell and Caml to Scala, Javascript and even Java 8 nowadays.

Also emerging for **probabilistic programming**.

Price to pay: **analysis of higher-order constructs**.

Advantages of functional programs

Price to pay: **analysis of higher-order constructs**.

Example of higher-order function: `map`.

`map φ [0, 1, 2]` returns `$[\varphi(0), \varphi(1), \varphi(2)]$` .

Higher-order: `map` is a function taking a function φ as input.

Roadmap

- ① A few words on the λ -calculus and an introduction to type systems
- ② Intersection type systems for higher-order model-checking
- ③ Towards the verification of probabilistic programs

A few words on the λ -calculus

Definition, simply-typed fragment, towards intersection types

λ -terms

Grammar:

$$M, N ::= x \mid \lambda x.M \mid M N$$

Calculus of functions:

- x is a variable,
- $\lambda x.M$ is intuitively a function $x \mapsto M$,
- $M N$ is the application of functions.

λ -terms

Grammar:

$$M, N ::= x \mid \lambda x.M \mid M N$$

Examples:

- $\lambda x.x$: identity $x \mapsto x$,
- $\lambda x.y$: constant function $x \mapsto y$,
- $(\lambda x.x) y$: application of the identity to y ,
- $\Delta = \lambda x.x x$: **duplication**.

β -reduction

$$(\lambda x.x) y$$

is an application of functions which should compute y :

$$(\lambda x.x) y \rightarrow_{\beta} y$$

Beta-reduction gives the dynamics of the calculus.
(= the evaluation of the functions/programs).

This calculus is equivalent in expressive power, for functions $\mathbb{N} \rightarrow \mathbb{N}$, to Turing machines.

β -reduction

Formally:

$$(\lambda x.M) N \rightarrow_{\beta} M[x/N]$$

Examples:

$$(\lambda x.y) z \rightarrow_{\beta} y$$

β -reduction

Formally:

$$(\lambda x.M) N \rightarrow_{\beta} M[x/N]$$

Examples:

$$\begin{aligned} & (\lambda f.\lambda x.f (f x)) (g g) y \\ \rightarrow_{\beta} & (\lambda x.g (g (g x))) y \\ \rightarrow_{\beta} & g (g (g y)) \end{aligned}$$

The looping term Ω

Just like with Turing machines, there are computations that never stop.

Set $\Omega = \Delta \Delta = (\lambda x.x x)(\lambda x.x x)$.

Then:

$$\begin{aligned}\Omega &= (\lambda x.x x)(\lambda x.x x) \\ &\rightarrow_{\beta} (x x) [x/\lambda x.x x] = \Omega \\ &\rightarrow_{\beta} \Omega \\ &\rightarrow_{\beta} \dots\end{aligned}$$

The looping term Ω

Just like with Turing machines, there are computations that never stop.
But that may depend on how we compute.

$$(\lambda x.y) \Omega \rightarrow_{\beta} y$$

if we reduce the first redex, or

$$(\lambda x.y) \Omega \rightarrow_{\beta} (\lambda x.y) \Omega$$

if we try to reduce the second (inside Ω)...

- **Weak normalization**: at least one way of computing terminates
- **Strong normalization (SN)**: all ways of computing terminate.

Simple types and strong normalization

Problem with Ω : it contains $x x$.

So x is **at the same time** a function and an argument of this function.

Simple types forbid this: you have to be a function $A \rightarrow A$ or an argument of type A , but not both.

It is **enough** to guarantee strong normalization:

M has a simple type $\Rightarrow M$ is SN.

It's an **incomplete** characterization: $\Delta = \lambda x.x x$ is SN (no way to reduce it!) but not typable.

(simple typing is decidable, so it couldn't be complete).

Simple types

Simple types: $\sigma, \tau ::= o \mid \sigma \rightarrow \tau$.

$$\frac{}{\Gamma, x : \sigma \vdash x : \sigma} \qquad \frac{\Gamma, x : \sigma \vdash M : \tau}{\Gamma \vdash \lambda x. M : \sigma \rightarrow \tau}$$

$$\frac{\Gamma \vdash M : \sigma \rightarrow \tau \quad \Gamma \vdash N : \sigma}{\Gamma \vdash M N : \tau}$$

Intersection types and strong normalization

A complete (and undecidable) characterization of SN: **intersection types**.

Now, $\lambda x. x x$ has type $((\tau \rightarrow \tau) \wedge \tau) \rightarrow \tau$ for all (intersection) types $\tau \dots$

A term is SN iff it is typable in an appropriate intersection type system. Δ is typable, Ω isn't.

Crucial feature: intersection type systems enjoy both **subject reduction** and **subject expansion**.

In other words: typing is invariant by reduction. We'll use that to do **static analysis!**

Modeling functional programs using higher-order recursion schemes

Model-checking

Approximate the program \longrightarrow build a **model** \mathcal{M} .

Then, formulate a **logical specification** φ over the model.

Aim: design a **program** which checks whether

$$\mathcal{M} \models \varphi.$$

That is, whether the model \mathcal{M} meets the specification φ .

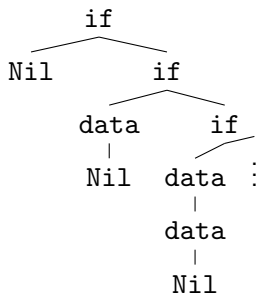
An example

```
    Main    =    Listen Nil
Listen x   =    if end_signal() then x
              else Listen received_data() :: x
```

An example

```
Main      = Listen Nil
Listen x  = if end_signal() then x
           else Listen received_data():x
```

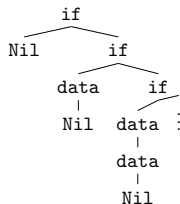
A **tree** model:



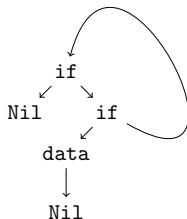
We abstracted **conditionals** and **datatypes**.

The approximation contains a non-terminating branch.

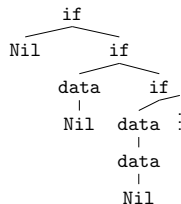
Finite representations of infinite trees



is not **regular**: it is not the unfolding of a **finite** graph as



Finite representations of infinite trees



but it is represented by a **higher-order recursion scheme** (HORS).

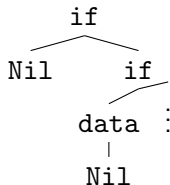
Higher-order recursion schemes

```
Main      = Listen Nil
Listen x   = if end_signal() then x
            else Listen received_data() :: x
```

is abstracted as

$$\mathcal{G} = \begin{cases} S & = L \text{ Nil} \\ L x & = \text{if } x (L (\text{data } x)) \end{cases}$$

which represents the higher-order tree of actions



Higher-order recursion schemes

$$\mathcal{G} = \begin{cases} S & = & L \text{ Nil} \\ L \ x & = & \text{if } x \ (L \ (\text{data } x)) \end{cases}$$

Rewriting starts from the **start symbol** S:

$$S \quad \rightarrow_{\mathcal{G}} \quad \begin{array}{c} L \\ | \\ \text{Nil} \end{array}$$

Higher-order recursion schemes

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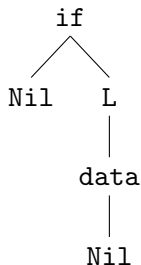
L
|
Nil

$\rightarrow_{\mathcal{G}}$

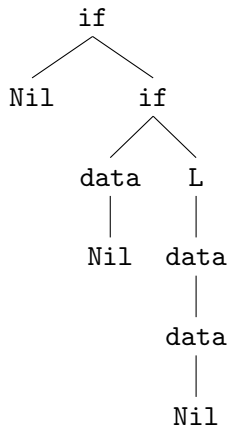
if
/ \
Nil L
|
data
|
Nil

Higher-order recursion schemes

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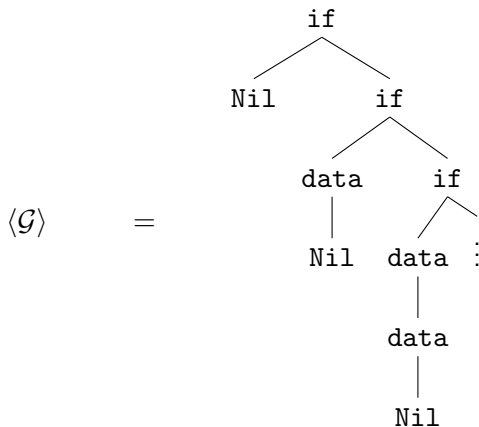


$\rightarrow_{\mathcal{G}}$



Higher-order recursion schemes

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Higher-order recursion schemes

$$\mathcal{G} = \begin{cases} S & = L \text{ Nil} \\ L x & = \text{if } x (L (\text{data } x)) \end{cases}$$

can be rewritten in λ -calculus style as

$$\mathcal{G} = \begin{cases} S & = L \text{ Nil} \\ L & = \lambda x. \text{if } x (L (\text{data } x)) \end{cases}$$

HORS can alternatively be seen as **simply-typed** λ -terms with **simply-typed recursion operators** $Y_\sigma : (\sigma \rightarrow \sigma) \rightarrow \sigma$.

Note that, in general, arguments may be functions of functions of functions. . .

Alternating parity tree automata

Checking specifications over trees

Monadic second order logic

MSO is a common logic in verification, allowing to express properties as:

“ all executions halt ”

“ a given operation is executed infinitely often in some execution ”

“ every time data is added to a buffer, it is eventually processed ”

Alternating parity tree automata

Checking whether a formula holds can be performed using an **automaton**.

For an MSO formula φ , there exists an equivalent APT \mathcal{A}_φ s.t.

$$\langle \mathcal{G} \rangle \models \varphi \quad \text{iff} \quad \mathcal{A}_\varphi \text{ has a run over } \langle \mathcal{G} \rangle.$$

APT = **alternating** tree automata (ATA) + **parity** condition.

Alternating tree automata

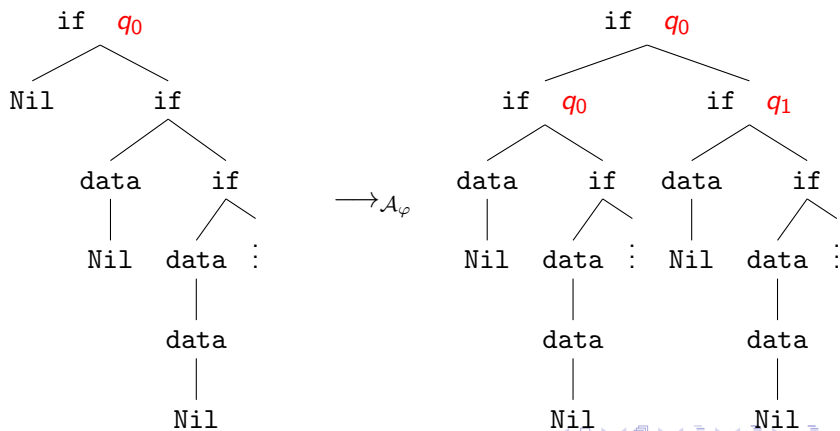
ATA: **non-deterministic** tree automata whose transitions may **duplicate** or **drop** a subtree.

Typically: $\delta(q_0, \text{if}) = (2, q_0) \wedge (2, q_1)$.

Alternating tree automata

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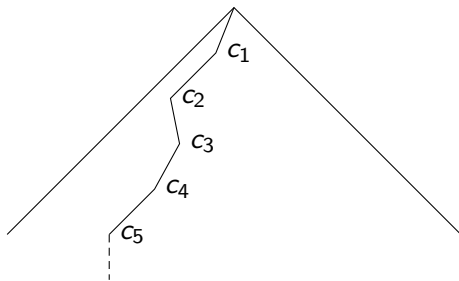


Alternating parity tree automata

Each state of an APT is attributed a **color**

$$\Omega(q) \in Col \subseteq \mathbb{N}$$

An infinite branch of a run-tree is **winning** iff the **maximal color among the ones occurring infinitely often along it is even**.



Alternating parity tree automata

Each state of an APT is attributed a **color**

$$\Omega(q) \in Col \subseteq \mathbb{N}$$

An infinite branch of a run-tree is **winning** iff the **maximal color among the ones occurring infinitely often along it is even**.

A run-tree is **winning** iff all its infinite branches are.

For a MSO formula φ :

$$\mathcal{A}_\varphi \text{ has a } \mathbf{winning} \text{ run-tree over } \langle \mathcal{G} \rangle \quad \text{iff} \quad \langle \mathcal{G} \rangle \models \varphi.$$

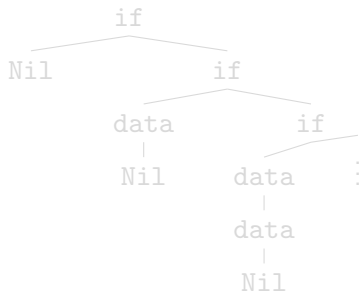
The higher-order model-checking problem

The (local) HOMC problem

Input: HORS \mathcal{G} , formula φ .

Output: true if and only if $\langle \mathcal{G} \rangle \models \varphi$.

Example: $\varphi =$ “ there is an infinite execution ”



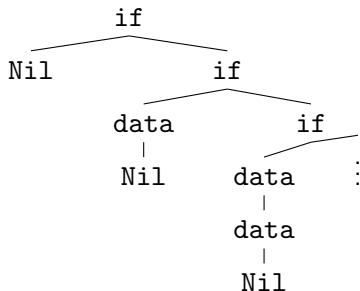
Output: true.

The (local) HOMC problem

Input: HORS \mathcal{G} , formula φ .

Output: true if and only if $\langle \mathcal{G} \rangle \models \varphi$.

Example: $\varphi =$ “ there is an infinite execution ”



Output: true.

Our line of work

This problem is **decidable** (Ong 2006), and its complexity is n -EXPTIME where n is the **order** of the HORS of interest.

But there are practical algorithms that work quite well!

Our contributions (with Melliès, Clairambault and Murawski):

- A connection with linear logic and its models, based on a refinement of an intersection type system and on a connection between intersection types and linear logic
- **Explain why it works**: in fact, complexity depends on the **linear order** of the HORS
- For this, we introduce a linear-nonlinear version of HORS and of APT. This framework allows us to give simpler proofs of existing results of HOMC, and allows to **unify** these existing approaches.

Intersection types and alternation

A first connection with linear logic

Alternating tree automata and intersection types

A key remark (Kobayashi 2009):

$$\delta(q_0, \text{if}) = (2, q_0) \wedge (2, q_1)$$

can be seen as the intersection typing

$$\text{if} : \emptyset \rightarrow (q_0 \wedge q_1) \rightarrow q_0$$

refining the simple typing

$$\text{if} : o \rightarrow o \rightarrow o$$

Alternating tree automata and intersection types

In a derivation typing the tree $\text{if } T_1 \ T_2$:

$$\text{App} \frac{\delta \frac{\emptyset \vdash \text{if} : \emptyset \rightarrow (q_0 \wedge q_1) \rightarrow q_0}{\emptyset \vdash \text{if } T_1 : (q_0 \wedge q_1) \rightarrow q_0} \quad \emptyset \quad \frac{\vdots}{\emptyset \vdash T_2 : q_0} \quad \frac{\vdots}{\emptyset \vdash T_2 : q_1}}{\emptyset \vdash \text{if } T_1 \ T_2 : q_0}$$

Intersection types naturally lift to higher-order – and thus to \mathcal{G} , which **finitely** represents $\langle \mathcal{G} \rangle$.

Theorem (Kobayashi 2009)

$\vdash \mathcal{G} : q_0$ *iff* *the ATA \mathcal{A}_φ has a run-tree over $\langle \mathcal{G} \rangle$.*

A form of **static analysis!**

A type-system for verification: without parity conditions

$$\text{Axiom} \quad \frac{}{x : \bigwedge_{\{i\}} \theta_i :: \kappa \vdash x : \theta_i :: \kappa}$$

$$\delta \quad \frac{\{(i, q_{ij}) \mid 1 \leq i \leq n, 1 \leq j \leq k_i\} \text{ satisfies } \delta_A(q, a)}{\emptyset \vdash a : \bigwedge_{j=1}^{k_1} q_{1j} \rightarrow \dots \rightarrow \bigwedge_{j=1}^{k_n} q_{nj} \rightarrow q :: o \rightarrow \dots \rightarrow o}$$

$$\text{App} \quad \frac{\Delta \vdash t : (\theta_1 \wedge \dots \wedge \theta_k) \rightarrow \theta :: \kappa \rightarrow \kappa' \quad \Delta_i \vdash u : \theta_i :: \kappa}{\Delta, \Delta_1, \dots, \Delta_k \vdash tu : \theta :: \kappa'}$$

$$\lambda \quad \frac{\Delta, x : \bigwedge_{i \in I} \theta_i :: \kappa \vdash t : \theta :: \kappa'}{\Delta \vdash \lambda x. t : (\bigwedge_{i \in I} \theta_i) \rightarrow \theta :: \kappa \rightarrow \kappa'}$$

$$\text{fix} \quad \frac{\Gamma \vdash \mathcal{R}(F) : \theta :: \kappa}{F : \theta :: \kappa \vdash F : \theta :: \kappa}$$

A closer look at the Application rule

In the intersection type system:

$$\text{App} \quad \frac{\Delta \vdash t : (\theta_1 \wedge \dots \wedge \theta_n) \rightarrow \theta \quad \Delta_i \vdash u : \theta_i}{\Delta, \Delta_1, \dots, \Delta_n \vdash t u : \theta}$$

This rule could be decomposed as:

$$\frac{\Delta \vdash t : (\bigwedge_{i=1}^n \theta_i) \rightarrow \theta' \quad \frac{\Delta_i \vdash u : \theta_i \quad \forall i \in \{1, \dots, n\}}{\Delta_1, \dots, \Delta_n \vdash u : \bigwedge_{i=1}^n \theta_i}}{\Delta, \Delta_1, \dots, \Delta_n \vdash t u : \theta'} \quad \text{Right } \wedge$$

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A closer look at the Application rule

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Right \bigwedge

Linear decomposition of the intuitionistic arrow:

$$A \Rightarrow B = !A \multimap B$$

Two steps: **duplication / erasure**, then **linear use**.

Right \bigwedge corresponds to the **Promotion** rule of indexed linear logic.
(see G.-Melliès, ITRS 2014)

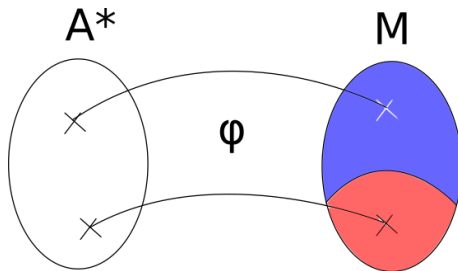
Overview of our results

Automata and recognition

For the usual **finite** automata on **words**: given a **regular** language $L \subseteq A^*$,

there exists a finite **automaton** \mathcal{A} recognizing L

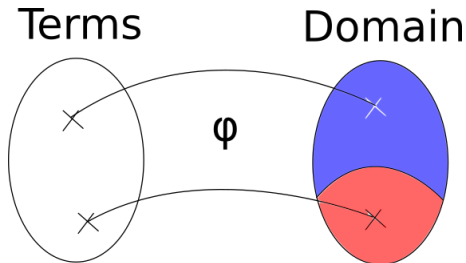
if and only if...



there exists a finite **monoid** M , a subset $K \subseteq M$
and a **homomorphism** $\varphi : A^* \rightarrow M$ such that $L = \varphi^{-1}(K)$.

Automata and recognition

The picture we want:



(after Aehlig 2006, Salvati 2009)

but with **recursion** and w.r.t. an APT.

Finitary semantics of linear logic

In ScottL (a finitary model of linear logic), we define \square , λ and \mathbf{Y} in an appropriate way.

$ScottL_{\frac{1}{2}}$ is a model of the λY -calculus.

Theorem

An APT \mathcal{A} has a winning run from q_0 over $\langle \mathcal{G} \rangle$ if and only if

$$q_0 \in \llbracket \lambda(\mathcal{G}) \rrbracket.$$

Corollary

The local higher-order model-checking problem is decidable (and is n -EXPTIME complete).

See Grellois-Melliès: CSL 2015, Fossacs 2015, MFCS 2015, and my thesis.

Linear order and the true complexity of HOMC

Clairambault, G., Murawski, POPL 2018: order isn't the good measure for complexity. We can use **linear** order.

Idea: when the automaton doesn't use alternation, complexity doesn't increase that much. . .

A big advantage: allows to reprove several works on HOMC in a much simpler way!

Probabilistic Termination

Motivations

- **Probabilistic** programming languages are more and more pervasive in computer science: modeling uncertainty, robotics, cryptography, machine learning, AI. . .
- **Quantitative** notion of termination: **almost-sure termination** (AST)
- AST has been studied for imperative programs in the last years. . .
- . . . but what about the **functional** probabilistic languages?

We introduce a **monadic, affine sized type system** sound for AST.

Sized types: the deterministic case

Simply-typed λ -calculus is strongly normalizing (SN).

$$\frac{}{\Gamma, x : \sigma \vdash x : \sigma} \quad \frac{\Gamma, x : \sigma \vdash M : \tau}{\Gamma \vdash \lambda x.M : \sigma \rightarrow \tau}$$

$$\frac{\Gamma \vdash M : \sigma \rightarrow \tau \quad \Gamma \vdash N : \sigma}{\Gamma \vdash M N : \tau}$$

where $\sigma, \tau ::= o \mid \sigma \rightarrow \tau$.

Forbids the looping term $\Omega = (\lambda x.x x)(\lambda x.x x)$.

Strong normalization: all computations terminate.

Sized types: the deterministic case

Simply-typed λ -calculus is strongly normalizing (SN).

No longer true with the **letrec** construction. . .

Sized types: a **decidable** extension of the simple type system ensuring SN for λ -terms with letrec.

See notably:

- Hughes-Pareto-Sabry 1996, *Proving the correctness of reactive systems using sized types*,
- Barthe-Frade-Giménez-Pinto-Uustalu 2004, *Type-based termination of recursive definitions*.

Sized types: the deterministic case

Sizes: $s, t ::= i \mid \infty \mid \widehat{s}$

+ size comparison underlying **subtyping**. Notably $\widehat{\infty} \equiv \infty$.

Idea: k successors = at most k constructors.

- $\text{Nat}^{\widehat{i}}$ is 0,
- $\text{Nat}^{\widehat{i}}$ is 0 or $S\ 0$,
- ...
- Nat^{∞} is any natural number. Often denoted simply Nat .

The same for lists, ...

Sized types: the deterministic case

Sizes: $\mathfrak{s}, \mathfrak{t} ::= \mathfrak{i} \mid \infty \mid \widehat{\mathfrak{s}}$

+ size comparison underlying **subtyping**. Notably $\widehat{\infty} \equiv \infty$.

Fixpoint rule:

$$\frac{\Gamma, f : \text{Nat}^{\mathfrak{i}} \rightarrow \sigma \vdash M : \text{Nat}^{\widehat{\mathfrak{i}}} \rightarrow \sigma[\mathfrak{i}/\widehat{\mathfrak{i}}] \quad \mathfrak{i} \text{ pos } \sigma}{\Gamma \vdash \text{letrec } f = M : \text{Nat}^{\mathfrak{s}} \rightarrow \sigma[\mathfrak{i}/\mathfrak{s}]}$$

“To define the action of f on size $n + 1$,
we only call recursively f on size at most n ”

Sized types: the deterministic case

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Sound for SN: typable \Rightarrow SN.

Decidable type inference (implies incompleteness).

Sized types: example in the deterministic case

From Barthe et al. (op. cit.):

$$\begin{aligned} \text{plus} \equiv & (\text{letrec } \text{plus} : \text{Nat}' \rightarrow \text{Nat} \rightarrow \text{Nat} = \\ & \lambda x : \text{Nat}' . \lambda y : \text{Nat} . \text{case } x \text{ of } \{ \text{o} \Rightarrow y \\ & \quad | \text{s} \Rightarrow \lambda x' : \text{Nat}' . \text{s } \underbrace{(\text{plus } x' y)}_{:\text{Nat}} \\ & \quad \} \\ &) : \quad \text{Nat}^s \rightarrow \text{Nat} \rightarrow \text{Nat} \end{aligned}$$

The case rule ensures that the size of x' is lesser than the one of x .
Size decreases during recursive calls \Rightarrow SN.

A probabilistic λ -calculus

With Dal Lago, we studied a call-by-value λ -calculus extended with a probabilistic choice operator.

We designed a type system, inspired from sized types, in which

typability \Rightarrow AST

Random walks as probabilistic terms

- **Biased** random walk:

$$M_{bias} = \left(\text{letrec } f = \lambda x. \text{case } x \text{ of } \left\{ S \rightarrow \lambda y. f(y) \oplus_{\frac{2}{3}} (f(SSy)) \mid 0 \rightarrow 0 \right\} \right) \eta$$

- **Unbiased** random walk:

$$M_{unb} = \left(\text{letrec } f = \lambda x. \text{case } x \text{ of } \left\{ S \rightarrow \lambda y. f(y) \oplus_{\frac{1}{2}} (f(SSy)) \mid 0 \rightarrow 0 \right\} \right) \eta$$

$$\sum \llbracket M_{bias} \rrbracket = \sum \llbracket M_{unb} \rrbracket = 1$$

This is checked by our type system.

Another term

We also capture terms as:

$$M_{nat} = \left(\text{letrec } f = \lambda x.x \oplus_{\frac{1}{2}} S (f x) \right) 0$$

of semantics

$$\llbracket M_{nat} \rrbracket = \left\{ (0)^{\frac{1}{2}}, (S 0)^{\frac{1}{4}}, (S S 0)^{\frac{1}{8}}, \dots \right\}$$

summing to 1.

Remark that this recursive function generates the **geometric** distribution.

Probabilistic termination of probabilistic HORS

Termination analysis for functional programs, one more step:

Kobayashi - Dal Lago - Grellois, LICS 2019: a probabilistic extension of HORS and approximate techniques for termination verification.

Conclusion

- Type theory is perfectly fit to verify functional programs, by considering appropriate type systems.
- Type theory can be extended to accommodate the probabilistic setting.
- Linear logic is very connected to intersection types, and allows to consider more carefully the behavior of programs
- **Contributions:** models of LL for HOMC, refinement of the complexity measures, a new framework to carry proofs in HOMC, termination analysis for probabilistic programs.

Thank you for your attention!

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