

From Hard Work to Trickery: A Systematic Approach to Probabilistic Rewriting.

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Probabilistic Rewriting

Motivations

Randomized Algorithms

Efficiency analysis

Cryptography

Security

Machine Learning

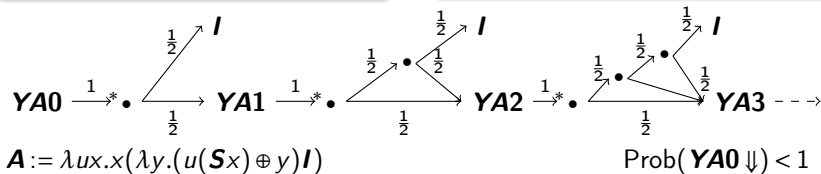
Modeling

Probabilistic λ -calculus (Weak head-reduction)

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

$(\lambda x.M) N \xrightarrow{1} M[N/x]$

$M \oplus N \begin{cases} \xrightarrow{\frac{1}{2}} M \\ \xrightarrow{\frac{1}{2}} N \end{cases}$



The Quest for a Semantics

Denotational Semantics: a 20 years old challenge

Finding subclass of domains that:

- is a Cartesian close
- have probabilistic powerdomains

Real issue:

higher order probabilities

Operational Semantics: a hidden challenge

Rewriting theory with:

- probabilistic behaviors
- systematic proof-schemes

Real issue:

proba forces topological arguments

Unconventional solution:
Probabilistic coherent spaces

[EhrhardPaganiTasson2014]



Unconventional solution
???

Our objective

What is a Systematic Proof Schema?

The Example of Infinite Rewriting

Question:

How to relate small-step and multi-step?

At the beginning: Topology

Limits for Cantor topology of sequential small-step reductions.

Now-day: Coinduction

$$\frac{M \rightarrow^* f(L_1, \dots, L_k) \quad \forall i \leq k, L_i \rightarrow^\omega N_i}{M \rightarrow^\omega f(N_1, \dots, N_k)}$$

Coinduction Schema

For any relation \rightsquigarrow over terms, if for all $M \rightsquigarrow f(N_1, \dots, N_k)$, there is L_1, \dots, L_k such that $M \rightarrow^* f(L_1, \dots, L_k)$ and $L_i \rightsquigarrow N_i$, then $\rightsquigarrow \subseteq \rightarrow^\omega$.

What About Probabilistic Rewriting

Probabilities and Non-determinism does not mix well

For now, let's forget about non determinism. This means:

Fixing a strategy

Big step rather than multistep

Probabilities are
inherently topological

$[0, 1]$ is, before all,
a topological space...

Most rewriting theory's tools
are continuous

Bisimulations, encoding,
typing, modeling...

Can we treat those tools without referring to topology?

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Yes! but there is a price to pay: a dynamic target

Probabilistic Rewriting System

Randomized function

$f : U \rightarrow V$ denotes a function $f : U \rightarrow \mathcal{D}(V)$

$$\mathcal{D}(V) = \{d \in V \rightarrow [0,1] \mid \sum_{v \in V} d(v) \leq 1\}$$

Definition of (Abstract) Probabilistic Rewriting System

Terms

Λ

Normal forms

$$\Lambda = \Lambda_V \uplus \Lambda_R$$

Small step

$$\text{reduction} : \Lambda_R \rightarrow \Lambda$$

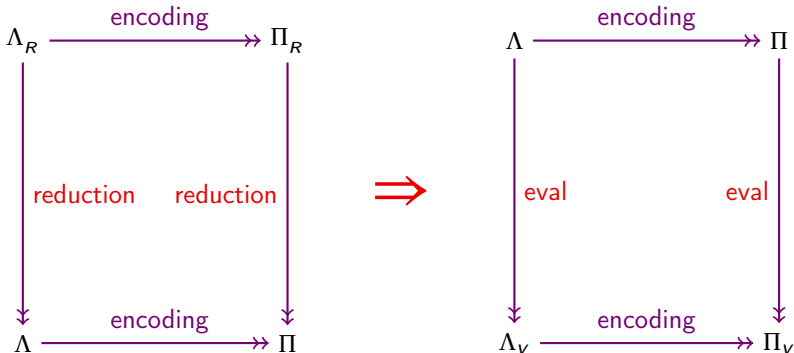
Remark: We only consider **deterministic** systems (with a strategy)

Coalgebraic approach of the big step reduction [Hasuo]

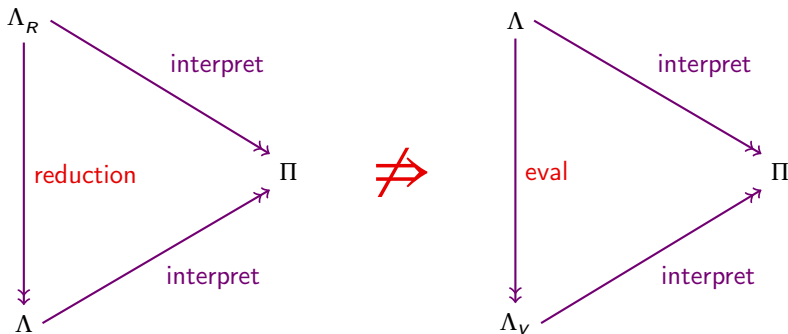
The evaluation $\text{eval} : \Lambda \rightarrow \Lambda_V$ corresponds to the final arrow in the $(_ \times \mathbb{N})$ -coalgebra category over set and randomized functions.

Theorem: Randomized Encoding

Λ, Π probabilistic rewriting systems.
encoding: $\Lambda \rightarrow \Pi$ a randomized function preserving NF and reducibles.



Dynamic is Essential



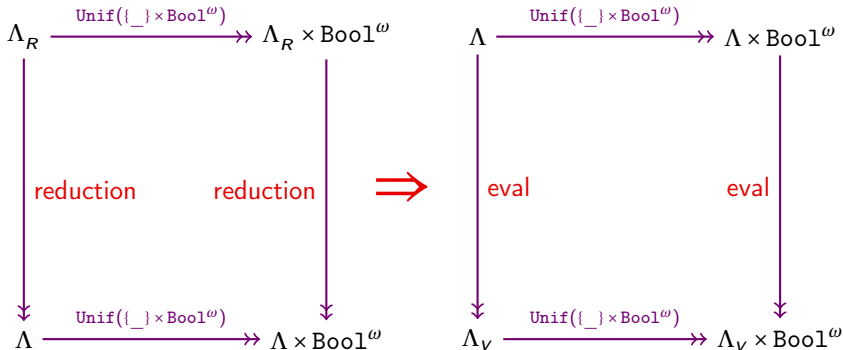
Only true if $\|\text{interpret}(M)\| \leq \|\text{eval}(M)\|$

Require a nontrivial realisability proof.

Example: Performing Choices First

Example

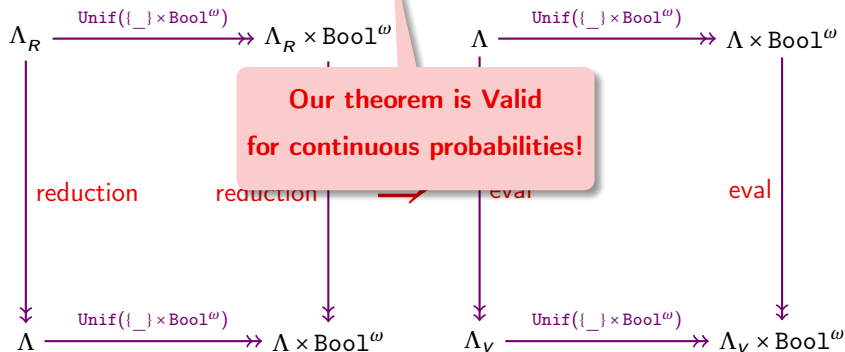
In any (binary) probabilistic rewriting system, the probabilistic choices can be chosen uniformly over Bool^ω at the beginning.



Example: Performing Choices First

Example

In any (binary) probabilistic rewriting system, the probabilistic choices can be chosen uniformly over Bool^ω at the beginning.



Probabilistic Intersection Types

From Probabilistic Coherence Spaces To Probabilistic Intersection Types

Standard translation:

- compact points \rightsquigarrow intersection types
- prime algebraic points \rightsquigarrow linear types

$$\frac{\pi}{\vdash M : p \cdot \alpha}$$

probabilistic bound
or weight

The adequation (reformulation)

$$\text{Prob}(M \Downarrow) = \sum W(M)$$

$$\text{where } W(M) = \left[p \mid \frac{\pi}{\vdash M : p \cdot * } \right]$$

Underlying function

$$\| \text{eval}(M) \| = \| \text{deriv}(M) \| \quad \text{deriv} : \left(\begin{array}{l} \Lambda \rightarrow \\ M \mapsto \end{array} \left\{ \frac{\mathcal{D}(\Pi)}{\pi} \right. \right. \left. \left. \frac{\pi}{\vdash M : p \cdot * } \mapsto p \right\} \right)$$

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$$\frac{\pi}{\vdash M : p \cdot \alpha}$$

**p IS NOT the probability
for M to be of type α**

**Rather, p IS the probability
for π to be a proof of $\vdash M : \alpha$**

The ade

Pro

where $W(M) = \left[p \mid \frac{\pi}{\vdash M : p \cdot * } \right]$

Underlying function

$$\|eval(M)\| = \|\text{deriv}(M)\| \quad \text{deriv: } \left(\begin{array}{l} \Lambda \rightarrow \\ M \mapsto \end{array} \left\{ \frac{\mathcal{D}(\Pi)}{\pi} \mapsto p \right\} \right)$$

Sketching the Proof of intersection types

Cut Elimination

$$\frac{\pi}{\vdash M : p \cdot *} \rightsquigarrow \frac{\pi'}{\vdash M' : q \cdot *}$$

such that: \rightsquigarrow is

- normalizing
- deterministic

“Poliadic λ -calculus”

Small-step distrib.

$$\begin{array}{ccc} \Lambda_R & \xrightarrow{\text{deriv}} & \Pi_R \\ \downarrow \text{reduction} & & \downarrow \text{red.} \\ \Lambda & \xrightarrow{\text{deriv}} & \Pi \end{array}$$

Value determinism

\forall normal form V ,

Unicity of derivation

$$\vdash V : 1 \cdot *$$

$$\frac{}{\vdash \lambda x.M : 1 \cdot *}$$

Big-step distribution

$$\begin{array}{ccc} \Lambda & \xrightarrow{\text{deriv}} & \Pi \\ \downarrow \text{eval} & & \downarrow \text{eval} \\ \Lambda_V & \xrightarrow{\text{deriv}} & \Pi_V \end{array}$$

Conclusion

$$\begin{aligned} \|\text{eval}(M)\| &= \|\text{deriv}(\text{eval}(M))\| \\ &= \|\text{eval}(\text{deriv}(M))\| \\ &= \|\text{deriv}(M)\| \end{aligned}$$

And Then?

Introduce non-determinism

Convex set of distributions

A **randomized simulation** is a function

$$f : U \rightarrow \mathcal{C}(\mathcal{D}(V))$$

targeting convex sets of distributions.

- Our Theorem holds for randomized simulation,
- A randomized encoding is a functional randomized simulation,
- A probabilistic bisimulation a derandomized randomized simulation.

Maybe a direction to treat real rewriting issues

Probabilistic confluence, Powerful bisimulations...