



GDR IM 2015 - LABRI - Bordeaux
2-3 February

Probabilistic Coherent Spaces vs Probabilistic PCF an accurate Semantics.

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An example of Randomized Algorithm

- Formalize its syntax.
- Reason on its semantics.

A Full Abstraction Result in a Probabilistic Setting

- Semantics : PCoh, Probabilistic Coherent Spaces [Girard04,Danos-Ehrhard11]
- Syntax : PPCF, a Probabilistic extension of PCF [Plotkin77]

Full Abstraction :

A Bridge between Syntax and Semantics.

« *Decide what you want to say before you worry how you are going to say it.* » The Scott-Strachey Approach to Programming Language Theory, preface, Scott (77)

Denotational semantics :

a program as a function between mathematical spaces

Operational semantics :

a program as a sequence of computation steps

« *Full Abstraction studies connections between denotational and operational semantics.* » LCF Considered as a Programming Language, Plotkin (77)

Operational semantics :

$\text{Proba}(P \rightarrow^* v)$ is the probability that P computes v .

Denotational semantics :

Programs seen as **distribution transformers** which are **very regular**, so that

Probabilistic Full Abstraction :

$$\llbracket P \rrbracket = \llbracket Q \rrbracket$$

Adequacy
 $\begin{matrix} \Rightarrow \\ \Leftarrow \end{matrix}$
Full Completeness

$$P \simeq_o Q$$

$$\forall C[\], \\ \text{Proba}(C[P] \rightarrow^* v) = \text{Proba}(C[Q] \rightarrow^* v)$$

Randomized algorithm :

A Las Vegas example.

An example of Randomized algorithm

Input : A 0/1 array of length $n \geq 2$ in which half cells are 0.

0	1	2	3	4	5
<u>0</u>	<u>1</u>	<u>0</u>	<u>1</u>	<u>1</u>	<u>0</u>

$f : 0, 2, 5 \mapsto \underline{0}, \quad 1, 3, 4 \mapsto \underline{1}$

Output : Find the index of a cell containing 0.

```
let rec LasVegas (f: nat -> nat) (n:nat) =  
  let k = random n in  
    if (f k = 0) then k  
    else LasVegas f n
```

This algorithm succeeds with probability one.

- Success in 1 step is : $\frac{1}{2}$.
- Success in 2 steps is : $\frac{1}{2^2}$.
- Success in n steps is : $\frac{1}{2^n}$.

Success in any steps is :

$$\sum_{k=1}^{\infty} \frac{1}{2^k} = 1.$$

Modeling Probabilistic Data and Programs :

Type : set of positive vectors

set of probability distributions

Program : weighted relation, a positive matrix

distribution transformer

Interaction : composition seen as multiplication

Example : nat

`Coin`: `nat` returns the toss of a fair coin.

`Unif n`: `nat` returns uniformly any $\{0, \dots, n\}$.

Non Determinism, a first approximation : $|\text{nat}| = \mathbb{N}$.

$$|\text{Coin}| = \{0, 1\} \quad \text{and} \quad |\text{Unif } n| = \{0, \dots, n\}$$

Enriching with positive coefficients : $[[\text{nat}]] \subseteq (\mathbb{R}^+)^{\mathbb{N}}$.

$$[[\text{Coin}]] = \left(\frac{1}{2}, \frac{1}{2}, 0, \dots \right) \quad \text{and} \quad [[\text{Unif } n]] = \left(\frac{1}{n+1}, \dots, \frac{1}{n+1}, 0, \dots \right)$$

$$\begin{array}{ccccccc} \downarrow & \downarrow & \downarrow & \dots & & \downarrow & \dots & \downarrow & \dots \\ 0 & 1 & 2 & \dots & & 0 & \dots & n & \dots \end{array}$$

Subprobability Distributions over \mathbb{N} :

$$[[\text{nat}]] = \left\{ (\lambda_n)_{n \in \mathbb{N}} \mid \forall n, \lambda_n \in \mathbb{R}^+ \text{ and } \sum_n \lambda_n \leq 1 \right\}$$

Example : $\text{Unif} : \text{nat} \rightarrow \text{nat}$

Input : an integer n

Output : any integer $\{0, \dots, n\}$ uniformly chosen.

Non Determinism : $|\text{Unif}| \subseteq |\text{nat}| \times |\text{nat}|$ is a relation.

$$|\text{Unif}| = \{(n, k) \mid n \in \mathbb{N}, k \in \{0, \dots, n\}\}$$

Enriching with positive coefficients : $[[\text{Unif}]] \in (\mathbb{R}^+)^{(\mathbb{N} \times \mathbb{N})}$.

$$\begin{array}{cccccc} & 0 & 1 & \dots & n & \dots \\ & \downarrow & \downarrow & & \downarrow & \\ \left(\begin{array}{cccccc} 1 & \frac{1}{2} & \dots & \frac{1}{n+1} & \dots \\ 0 & \frac{1}{2} & \dots & \frac{1}{n+1} & \dots \\ \vdots & 0 & \ddots & \vdots & \\ & \vdots & 0 & \frac{1}{n+1} & \\ & & \vdots & \ddots & \ddots \end{array} \right) & \begin{array}{l} \rightarrow 0 \\ \rightarrow 1 \\ \vdots \\ \rightarrow n \\ \vdots \end{array} \end{array}$$

Modeling Probabilistic Programs

Once : $\text{nat} \rightarrow \text{nat}$

Input : an integer n

Output : if n=0 then 42
 else Coin

$$\begin{array}{cccc}
 0 & 1 & \dots & \dots \\
 \downarrow & \downarrow & & \\
 \left(\begin{array}{cccc}
 0 & \frac{1}{2} & \frac{1}{2} & \dots \\
 0 & \frac{1}{2} & \frac{1}{2} & \dots \\
 0 & 0 & 0 & \dots \\
 \dots & 0 & \dots & \ddots \\
 1 & 0 & \dots & \\
 \dots & 0 & \dots & \ddots
 \end{array} \right) & \begin{array}{l} \rightarrow 0 \\ \rightarrow 1 \\ \vdots \\ \vdots \\ \rightarrow 42 \\ \vdots \end{array}
 \end{array}$$

Twice : $\text{nat} \rightarrow \text{nat}$

Input : an integer n

Output : if n=0 then 42
 else Unif n

$$\begin{array}{ll}
 ([0], 42) & \mapsto 1 \\
 ([n_1, n_2], k) & \mapsto \frac{1}{n_1+1} + \frac{1}{n_2+1} \\
 & \text{if } 0 \leq k \leq n_1 \leq n_2 \\
 ([n_1, n_2], k) & \mapsto \frac{1}{n_2+1} \\
 & \text{if } n_1 < k \leq n_2 \\
 \text{Otherwise} & 0
 \end{array}$$

Twice : nat \rightarrow nat

Input : an integer n

Output : if n=0 then 42
 else Unif n

$([0], 42) \mapsto 1$

$([n_1, n_2], k) \mapsto \frac{1}{n_1+1} + \frac{1}{n_2+1}$
 if $0 \leq k \leq n_1 \leq n_2$

$([n_1, n_2], k) \mapsto \frac{1}{n_2+1}$
 if $n_1 < k \leq n_2$

Otherwise 0

Memo : nat \rightarrow nat

Input : an integer n

Output : Let z be n in
 if z=0 then 42
 else Unif z

$([0], 42) \mapsto 1$

$([n], k) \mapsto \frac{1}{n+1}$
 if $0 \leq n$

Otherwise 0

Probabilistic Data :

If $x : \text{nat}$, then $\llbracket x \rrbracket = (x_n)_{n \in \mathbb{N}}$

where x_n is the probability that x is n .

Probabilistic Program : $P : \text{nat} \rightarrow \text{nat}$

where $\llbracket P x \rrbracket_n$ is the probability that $P x$ computes n .

$$\llbracket \text{Once} \rrbracket \in (\mathbb{R}^+)^{\mathbb{N} \times \mathbb{N}}$$

$$\llbracket \text{Twice} \rrbracket \in (\mathbb{R}^+)^{\mathcal{M}_{\text{fin}}(\mathbb{N}) \times \mathbb{N}}$$

$$\begin{aligned} \llbracket \text{Once } x \rrbracket_n &= \llbracket \text{Once} \rrbracket \cdot \llbracket x \rrbracket \\ &= \sum_k \llbracket \text{Once} \rrbracket_{(k,n)} \llbracket x \rrbracket_k \end{aligned}$$

$$\begin{aligned} \llbracket \text{Twice } x \rrbracket_n &= \llbracket \text{Twice} \rrbracket \cdot \llbracket x \rrbracket^! \\ &= \sum_{\mu} \llbracket \text{Twice} \rrbracket_{(\mu,n)} \prod_{k \in \mu} \llbracket x \rrbracket_k \end{aligned}$$

Syntax :

probabilistic PCF

A Typed Probabilistic Functional Programming Language

Types :

$\sigma, \tau = \text{nat} \mid \sigma \Rightarrow \tau$

Programs :

$N, P, Q := \underline{n} \mid \text{succ}(N) \mid x \mid \lambda x^\sigma P \mid (P)Q \mid \text{fix}(N)$
| **if** ($N, P, [z] Q$), with z a variable
| **coin**(p), with $p \in \mathbb{Q} \cap [0, 1]$

Typing :

$\Gamma \vdash P : \sigma$

$$\frac{\Gamma \vdash N : \text{nat} \quad \Gamma \vdash P : \sigma \quad \Gamma, z : \text{nat} \vdash Q : \sigma}{\Gamma \vdash \text{if}(N, P, [z] Q) : \sigma}$$

$$\frac{p \in \mathbb{Q} \cap [0, 1]}{\text{coin}(p) : \text{nat}}$$

A Typed Probabilistic Functional Programming Language

Types :

$\sigma, \tau = \text{nat} \mid \sigma \Rightarrow \tau$

Programs :

$N, P, Q := \underline{n} \mid \text{succ}(N) \mid x \mid \lambda x^\sigma P \mid (P)Q \mid \text{fix}(N)$

$\mid \text{if}(N, P, [z] Q)$, with z a variable

$\mid \text{coin}(p)$, with $p \in \mathbb{Q} \cap [0, 1]$

Operational Semantics :

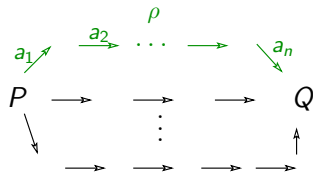
$P \xrightarrow{a} Q$

P reduces to Q in one step with probability a

$\text{Proba}(P \xrightarrow{*} Q) = \sum_{\rho} w(\rho)$

$\text{if}(\underline{0}, P, [z] Q) \xrightarrow{1} P$

$\text{if}(\underline{n+1}, P, [z] Q) \xrightarrow{1} Q[n/z]$



$w(\rho) = a_1 \cdots a_n$

$\text{coin}(p) \xrightarrow{p} \underline{0}$

$\text{coin}(p) \xrightarrow{(1-p)} \underline{1}$

Predecessor : $\text{nat} \Rightarrow \text{nat}$ takes integer n and returns 0 if $n = 0$ and $n - 1$ otherwise.

$$\text{pred} = \lambda x^{\text{nat}} \text{if } (x, \underline{0}, [z] z)$$

Addition : $\text{nat} \Rightarrow \text{nat} \Rightarrow \text{nat}$ takes integer n and m and returns $n + m$.

$$\text{add} = \lambda x^{\text{nat}} \text{fix}(\lambda a^{\text{nat} \Rightarrow \text{nat}} \lambda y^{\text{nat}} \text{if } (y, x, [z] \text{succ}((a)z)))$$

Exponential : $\text{nat} \Rightarrow \text{nat}$ takes integer n returns 2^n .

$$\text{exp} = \text{fix}(\lambda e^{\text{nat} \Rightarrow \text{nat}} \lambda x^{\text{nat}} \text{if } (x, \underline{1}, [z] (\text{add})(e)z(e)z))$$

Comparison : $\text{nat} \Rightarrow \text{nat} \Rightarrow \text{nat}$ takes integer n and m and returns 0 if $n \leq m$ and 1 otherwise.

$$\text{cmp} = \text{fix}(\lambda c^{\text{nat} \Rightarrow \text{nat} \Rightarrow \text{nat}} \lambda x^{\text{nat}} \lambda y^{\text{nat}} \text{if } (x, \underline{0}, [z] \text{if } (y, \underline{1}, [z'] (c)zz'))))$$

Memorization of the value of an expression of type `nat`.

let x **be** M **in** $N = \mathbf{if} (M, N[0/x], [z] N[\mathbf{succ}(z)/x])$

Uniform₂ : `nat` \Rightarrow `nat` takes integer n and returns a uniform probability distribution over $\{0, \dots, 2^n - 1\}$

Unif₂ =

$\mathbf{fix}(\lambda u^{\mathbf{nat} \Rightarrow \mathbf{nat}} \lambda x^{\mathbf{nat}} \mathbf{if} (x, \underline{0}, [z] \mathbf{if} (\mathbf{coin}(\frac{1}{2}), (u)z, [z'] (\mathbf{add})(\mathbf{exp})z(u)z)))$

Uniform : `nat` \Rightarrow `nat` takes integer n and returns a uniform probability distribution over $\{0, \dots, n\}$

Unif =

$\lambda x^{\mathbf{nat}} \mathbf{let} y \mathbf{be} x \mathbf{in} \mathbf{fix}(\lambda u^{\mathbf{nat}} \mathbf{let} z \mathbf{be} (\mathbf{Unif}_2)y \mathbf{in} \mathbf{if} ((\mathbf{cmp})zy, z, [w] (u)y))$

Caml encoding :

```

let rec LasVegas (f:nat->nat) (n:nat) =
  let k = random n in
    if (f k = 0) then k
    else LasVegas f n

```

PCF encoding :

```

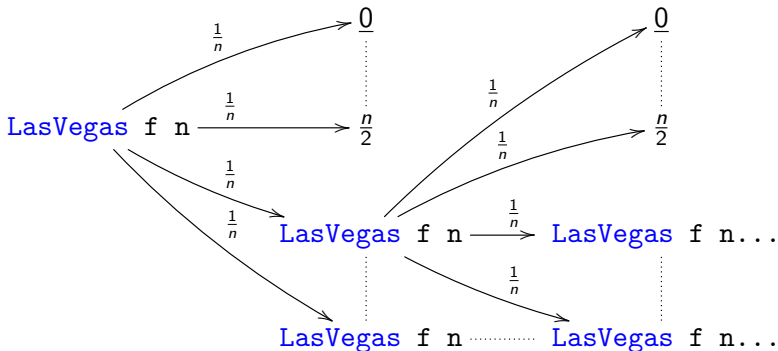
LasVegas = fix ( $\lambda LV^{(\text{nat} \Rightarrow \text{nat}) \Rightarrow \text{nat} \Rightarrow \text{nat}}$   $\lambda f^{\text{nat} \Rightarrow \text{nat}}$   $\lambda n^{\text{nat}}$ 
  let k be Unif n in
    if (f k, k, (LV) f n)

```

PCF encoding :

$\text{LasVegas} = \mathbf{fix} \left(\lambda LV^{(\text{nat} \Rightarrow \text{nat}) \Rightarrow \text{nat} \Rightarrow \text{nat}} \lambda f^{\text{nat} \Rightarrow \text{nat}} \lambda n^{\text{nat}} \right.$
 $\quad \mathbf{let} \ k \ \mathbf{be} \ \text{Unif} \ n \ \mathbf{in}$
 $\quad \mathbf{if} \ (f \ k, \ k, \ (LV) \ f \ n)$

Operational Semantics :



Quantitative Semantics :

Probabilistic Coherent Spaces (Pcoh)

Types (Object) : representing randomized data : nat, ...

Programs (Maps) : Input Type \rightarrow Output Type

Interaction (Composition):

Input Type \xrightarrow{P} Intermediate Type \xrightarrow{Q} Output Type

Orthogonality :

$$x, y \in (\mathbb{R}^+)^{|\sigma|}.$$

$$x \perp y \iff \sum_{a \in |\sigma|} x_a y_a \in [0, 1].$$

Given a set $P \subseteq (\mathbb{R}^+)^{|\sigma|}$ we define P^\perp , the *orthogonal* of P , as

$$P^\perp := \{y \in (\mathbb{R}^+)^{|\sigma|} \mid \forall x \in P \langle x, y \rangle \leq 1\}.$$

Probabilistic Coherent Space :

$$\mathcal{X} = (|\mathcal{X}|, P(\mathcal{X}))$$

where $|\mathcal{X}|$ is a countable set
and $P(\mathcal{X}) \subseteq (\mathbb{R}^+)^{|\mathcal{X}|}$

such that the following holds :

closedness : $P(\mathcal{X})^{\perp\perp} = P(\mathcal{X})$,

boundedness : $\forall a \in |\mathcal{X}|, \exists \mu > 0, \forall x \in P(\mathcal{X}), x_a \leq \mu$,

completeness : $\forall a \in |\mathcal{X}|, \exists \lambda > 0, \lambda e_a \in P(\mathcal{X})$.

Objects of PCoh :

$$\mathcal{X} = (|\mathcal{X}|, P(\mathcal{X}))$$

where $|\mathcal{X}|$ is a countable set
and $P(\mathcal{X}) \subseteq (\mathbb{R}^+)^{|\mathcal{X}|}$

Type Example :

$$\llbracket \text{nat} \rrbracket = (\mathbb{N}, P(\text{nat}) = \{(\lambda_n) \mid \sum_n \lambda_n \leq 1\})$$

Data Example :

if $M : \text{nat}$, then $\llbracket M \rrbracket \in P(\text{nat}) \subseteq (\mathbb{R}^+)^{\mathbb{N}}$
is a subprobability distributions.

Coin($\frac{1}{2}$) :

$$\llbracket \text{Coin}(\frac{1}{2}) \rrbracket = (\frac{1}{2}, \frac{1}{2}, 0, \dots)$$

Maps of PCoh :

$$f : (|\mathcal{X}|, P(\mathcal{X})) \rightarrow (|\mathcal{Y}|, P(\mathcal{Y}))$$

defined as a **matrix** $M(f) \in (\mathbb{R}^+)^{\mathcal{M}_{\text{fin}}(|\mathcal{X}|) \times |\mathcal{Y}|}$

thanks to **Taylor formula** :

$$f(x) = \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} M(f)_{\mu} \cdot x^{\mu}$$

$$\text{with } x^{\mu} = \prod_{a \in \text{Supp}(x)} x_a^{\mu(a)}$$

f can be seen as an **entire function** $f : (\mathbb{R}^+)^{|\mathcal{X}|} \rightarrow (\mathbb{R}^+)^{|\mathcal{Y}|}$
preserving probabilistic coherence, $f(P(\mathcal{X})) \subseteq P(\mathcal{Y})$

Example :

if $P : \text{nat} \rightarrow \text{nat}$, then $\llbracket P \rrbracket : (\mathbb{R}^+)^{\mathbb{N}} \rightarrow (\mathbb{R}^+)^{\mathbb{N}}$
is an entire function preserving subprobability distributions.

Typing judgement : $\Gamma \vdash Q : \tau$ with $\Gamma = (x_1 : \sigma_1, \dots, x_n : \sigma_n)$

$$\llbracket Q \rrbracket^\Gamma : P(\llbracket \sigma_1 \rrbracket) \times \dots \times P(\llbracket \sigma_n \rrbracket) \rightarrow P(\llbracket \tau \rrbracket)$$

Examples : $P(\llbracket \text{nat} \rrbracket) = (\mathbb{N}, P(\text{nat}) = \{(\lambda_n) \mid \sum_n \lambda_n \leq 1\})$

$$\llbracket \underline{n} \rrbracket = (0, \dots, 0, \underset{\uparrow n}{1}, 0, \dots) \quad \llbracket \text{coin}(p) \rrbracket = (p, 1 - p, 0, \dots)$$

$$\llbracket x_i \rrbracket^\Gamma(\vec{u}) = \llbracket x_1 : A_1, \dots, x_n : A_n \vdash x_i \rrbracket(\vec{u}) = u_i$$

$$\begin{aligned} \llbracket \text{if}(N, P, [z] Q) \rrbracket^\Gamma(\vec{u}) &= \llbracket N \rrbracket^\Gamma(\vec{u})_0 \llbracket P \rrbracket^\Gamma(\vec{u}) \\ &\quad + \sum_{n=0}^{\infty} \llbracket N \rrbracket^\Gamma(\vec{u})_{n+1} \llbracket Q[n/z] \rrbracket^\Gamma(\vec{u}) \end{aligned}$$

Once : nat \rightarrow nat

λn if $n=0$ then 42 else Coin

$$\llbracket \text{Once} \rrbracket(x)_0 = \frac{1}{2} \sum_{n \geq 1} x_n$$

$$\llbracket \text{Once} \rrbracket(x)_1 = \frac{1}{2} \sum_{n \geq 1} x_n$$

$$\llbracket \text{Once} \rrbracket(x)_{42} = x_0$$

Twice : nat \rightarrow nat

λn if $n=0$ then 42 else Random n

$$\llbracket \text{Twice} \rrbracket(x)_k = \sum_{p=1}^k \sum_{q \geq k+1} \frac{1}{q} x_p x_q + \sum_{p=1}^{k+1} \sum_{q \geq k+1} \left(\frac{1}{p} + \frac{1}{q}\right) x_p x_q, \text{ if } k \neq 42$$

$$\llbracket \text{Twice} \rrbracket(x)_{42} = x_0 + \sum_{p=1}^{42} \sum_{q \geq 43} \frac{1}{q} x_p x_q + \sum_{p=1}^{43} \sum_{q \geq 43} \left(\frac{1}{p} + \frac{1}{q}\right) x_p x_q$$

A model of PPCF :

in particular, **PCoh** is a model of linear logic.

Compositionality :

For $C : \sigma \Rightarrow \tau, P : \sigma,$

$$\llbracket (C)P \rrbracket_- = \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} \llbracket C \rrbracket_{\mu, -} \llbracket P \rrbracket^{\mu}$$

Adequacy Lemma :

Let $M : \text{nat}$ be a closed program. Then for all $n,$

$$\text{Proba}(M \xrightarrow{*} \underline{n}) = \llbracket M \rrbracket_n.$$

LasVegas f n : nat

$$\text{Proba}(\text{LasVegas f n} \rightarrow^* \bullet) = \sum_k \llbracket \text{LasVegas f n} \rrbracket_k = 1$$

Probabilistic Full Abstraction :

The completeness theorem

FA relates Semantical and Observational equivalences :

Let $P, Q : \sigma \quad \forall \alpha \in |\sigma|, \llbracket P \rrbracket_\alpha = \llbracket Q \rrbracket_\alpha$

Adequacy $\Downarrow \Uparrow$ Full Completeness

$\forall C : \sigma \Rightarrow \text{nat}, \forall n \in |\text{nat}|,$
 $\text{Proba}((C)P \xrightarrow{*} n) = \text{Proba}((C)Q \xrightarrow{*} n)$

Adequacy proof :

- ① Apply **Adequacy Lemma** : $\forall n, \text{Proba}((C)P \xrightarrow{*} n) = \llbracket (C)P \rrbracket_n$.
- ② Apply **Compositionality** :

$$\forall n, \llbracket (C)P \rrbracket_n = \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} \llbracket C \rrbracket_{\mu, n} \prod_{\alpha \in \mu} \llbracket P \rrbracket_\alpha^{\mu(\alpha)}$$

FA relates Semantical and Observational equivalences :

Let $P, Q : \sigma \quad \forall \alpha \in |\sigma|, \llbracket P \rrbracket_\alpha = \llbracket Q \rrbracket_\alpha$

Adequacy $\Downarrow \Uparrow$ Full Completeness

$\forall C : \sigma \Rightarrow \text{nat}, \forall n \in |\text{nat}|,$
 $\text{Proba}((C)P \xrightarrow{*} n) = \text{Proba}((C)Q \xrightarrow{*} n)$

Adequacy proof :

- ① Apply Adequacy Lemma : $\forall n, \text{Proba}((C)P \xrightarrow{*} n) = \llbracket (C)P \rrbracket_n$.
- ② Apply Compositionality :

$$\begin{aligned} \forall n, \llbracket (C)P \rrbracket_n &= \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} \llbracket C \rrbracket_{\mu, n} \prod_{\alpha \in \mu} \llbracket P \rrbracket_\alpha^{\mu(\alpha)} \\ &= \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} \llbracket C \rrbracket_{\mu, n} \prod_{\alpha \in \mu} \llbracket Q \rrbracket_\alpha^{\mu(\alpha)} = \llbracket (C)Q \rrbracket_n \end{aligned}$$

FA relates Semantical and Observational equivalences :

Let $P, Q : \sigma$ $\quad \forall \alpha \in |\sigma|, \llbracket P \rrbracket_\alpha = \llbracket Q \rrbracket_\alpha$

Adequacy $\Downarrow \Uparrow$ Full Completeness

$\forall C : \sigma \Rightarrow \text{nat}, \forall n \in |\text{nat}|,$
 $\text{Proba}((C)P \xrightarrow{*} n) = \text{Proba}((C)Q \xrightarrow{*} n)$

Full Completeness proof :

- ① By **contradiction** : $\exists \alpha \in |\sigma|, \llbracket P \rrbracket_\alpha \neq \llbracket Q \rrbracket_\alpha$
- ② Find **testing context** : T_α such that $\llbracket (T_\alpha)P \rrbracket_0 \neq \llbracket (T_\alpha)Q \rrbracket_0$
- ③ Prove **definability** : $T_\alpha \in \text{PPCF}$
- ④ Apply **Adequacy** :
 $\text{Proba}((T_\alpha)P \xrightarrow{*} 0) \neq \text{Proba}((T_\alpha)Q \xrightarrow{*} 0).$

Sum up

- A **Probabilistic** extension of **PCF** encoding **LasVegas**
- A Quantitative semantics **Pcoh** enjoying
 - Regularity : $f(x) = \sum_{\mu \in \mathcal{M}_{\text{fin}}(|\sigma|)} M(f)_{\mu} \cdot x^{\mu}$
 - Adequacy : $\forall n, \text{Proba}(P \xrightarrow{*} \underline{n}) = \llbracket P \rrbracket_n$.
- **Full Abstraction** exploiting mathematical properties.

Connected works

- **Probabilistic power domains** [Saheb-Djahromi, Lawson, Plotkin and Jones, Goubault-Larrecq and Varracca]
- **Probabilistic Games** [Harmer-Danos]