

# Semi-decidability of may, must and probabilistic testing in a higher-type setting

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# Main result

## Theorem

*May, must and probabilistic testing are semi-decidable, in a fairly general setting including higher-types.*

### Observations:

- 1 Must testing is perhaps surprising:  
It involves universal quantification over an infinite set.
- 2 The other two involve existential quantification and integration.

# Ingredient 1

Can reduce to quantification and integration over the Cantor space.

This is the space of infinite sequences of binary digits.

## Ingredient 2

Can algorithmically quantify and integrate over the Cantor space.

Quantification amounts exhaustive search in finite time.

# Organization

- ① A programming language for non-determinism and probability.
- ② Logical types. [For results of semi-decisions.](#)
- ③ An executable program logic.
- ④ Operational semantics of the executable logic. [Algorithms.](#)
- ⑤ Denotational semantics of the executable logic. [Correctness.](#)

# Brief discussion of effects

## ML way.

- ① All effects are possible at all types.
- ② Come up with a monad that combines all effects.
- ③ The semantics is in the Kleisli category of that big monad.

## Haskell way.

- ① Explicitly define various monads as type constructors.
- ② For each effect, or maybe for each combination of a set of effects.
- ③ Several monads are used in the same program.
- ④ The programmer decides which monads he wants for each sub-program.

We develop our results in the Haskell way.

# A programming language for non-determinism and probability

Ground types:

$$\gamma ::= \text{Bool} \mid \text{Nat}$$

Powertype constructors:

$$F ::= H \mid S \mid P \mid V$$

- ① Hoare, Smyth, Plotkin, Probabilistic.
- ② May, must, may/must, on average.
- ③ Angelic, demonic, human.

Types:

$$\sigma, \tau ::= \gamma \mid \sigma \times \tau \mid \sigma \rightarrow \tau \mid F\sigma$$

Cartesian closed language.

# Example

The type

$$\sigma \times \tau \rightarrow \mathbf{V} \tau$$

can be used to code labeled Markov processes with:

- 1 label space  $A = \sigma$ ,
- 2 state space  $S = \tau$ , and
- 3 transition function  $t : A \times S \rightarrow \mathbf{V} S$ .



# Terms

For the sub-language over the PCF types

$$\sigma, \tau ::= \gamma \mid \sigma \times \tau \mid \sigma \rightarrow \tau$$

we take the PCF terms.

(Conditional, arithmetic,  $\lambda$ -calculus, fixed-point recursion.)

So no non-determinism or probability.

# Non-deterministic choice constants

For each type  $\sigma$  and each type constructor  $F \in \{\mathbf{H}, \mathbf{S}, \mathbf{P}\}$ , we have a constant

$$(\mathbb{V}^\sigma): F\sigma \times F\sigma \rightarrow F\sigma,$$

**Idea.** The term

this  $\mathbb{V}$  that

non-deterministically evaluates to this or that, angelically or demonically.

# Probabilistic choice constants

For each type  $\sigma$ , we have an infix constant

$$(\oplus^\sigma): V\sigma \times V\sigma \rightarrow V\sigma.$$

**Idea.** The term

this  $\oplus$  that

non-deterministically evaluates to this or that, with equal probability.

# Monad syntax

**Functor.** If  $f: \sigma \rightarrow \tau$  is a term, then so is

$$Ff: F\sigma \rightarrow F\tau.$$

**Unit.** For each type  $\sigma$ , we have a term

$$\eta_F^\sigma: \sigma \rightarrow F\sigma.$$

**Multiplication.** For each type  $\sigma$ , we have a constant

$$\mu_F^\sigma: FF\sigma \rightarrow F\sigma.$$

**Strength.** Left to the audience.

# Remark

We could have worked with monads as Kleisli triples  
(as in Haskell).

This makes no difference, but our choice is presentationally more  
convenient.

# Example

$$\eta(\lambda x.0) \circledast \eta(\lambda x.1) : F(\sigma \rightarrow \text{Nat})$$

$$\lambda x.\eta(0) \circledast \eta(1) : \sigma \rightarrow F\text{Nat}$$

**Remark.** If we apply the ML way to a call-by-name language, the terms

$$(\lambda x.0) \circledast (\lambda x.1)$$

and

$$\lambda x.(0 \circledast 1)$$

behave in the same way!

# Example: randomly choose an infinite sequence of booleans with uniform distribution

$\text{Cantor} = (\text{Nat} \rightarrow \text{Bool}).$

$\text{cons}: \text{Bool} \rightarrow \text{Cantor} \rightarrow \text{Cantor}.$

$\text{prefix}: \text{Bool} \rightarrow V \text{Cantor} \rightarrow V \text{Cantor}.$

$\text{prefix } p = V(\text{cons } p).$

$\text{random}: V \text{Cantor}.$

$\text{random} = (\text{prefix False random}) \oplus (\text{prefix True random}).$

# Possible-results operational semantics

$$\frac{M \Downarrow v}{M \otimes N \Downarrow v}$$

$$\frac{N \Downarrow v}{M \otimes N \Downarrow v}$$

$$\frac{M \Downarrow v}{M \oplus N \Downarrow v}$$

$$\frac{N \Downarrow v}{M \oplus N \Downarrow v}$$

$$\frac{M \Downarrow \eta(v) \quad f(v) \Downarrow w}{Ff(M) \Downarrow \eta(w)}$$

$$\frac{M \Downarrow v}{\eta(M) \Downarrow \eta(v)}$$

$$\frac{M \Downarrow \eta(V) \quad V \Downarrow \eta(W)}{\mu(M) \Downarrow \eta(W)}$$



# Schedulers

Think of elements of the Cantor space as “schedulers”.

Can decorate the operational semantics with schedulers,

$$M \Downarrow^s v,$$

so that

$$M \Downarrow v \text{ iff there is some } s \text{ with } M \Downarrow^s v.$$

# May and must convergence

$M$  *must converge*  $\iff$  for every  $s$  there is  $v$  with  $M \Downarrow^s v$ .

$M$  *may converge*  $\iff$  there are  $s$  and  $v$  with  $M \Downarrow^s v$ .

Our approach is based on this idea.  
But we implement it in a different way.

# The Sierpinski type

Term formation rules for a Sierpinski type  $S$ :

- ①  $\top : S$  is a term.
- ② If  $M : S$  and  $N : \sigma$  are terms then  $(\text{if } M \text{ then } N) : \sigma$  is a term.
- ③ If  $M, N : S$  are terms then so is  $M \vee N : S$ .

The only value (or canonical form) of type  $S$  is  $\top$ .

$$\frac{M \Downarrow \top \quad N \Downarrow V}{\text{if } M \text{ then } N \Downarrow V} \qquad \frac{M \Downarrow \top}{M \vee N \Downarrow \top} \qquad \frac{N \Downarrow \top}{M \vee N \Downarrow \top}.$$

# Computational adequacy of Scott model

If  $M$  is a closed term of ground type and  $v$  is a value then

$$\llbracket M \rrbracket = v \text{ iff } M \Downarrow v.$$

# The vertical unit-interval type $\mathbb{I}$

- 1 Interpreted as the dcpo  $([0, 1], \leq)$ .
- 2 Computations of terms  $M : \mathbb{I}$  allow to semi-decide the condition  $p < M$  with  $p$  rational.
- 3 But **not** the conditions  $M = p$  or  $M < p$  in general.

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- 6 We take the primitive operations those for **Real PCF**, restricted to such intervals.
- 7 Arithmetic functions,  $p < (-): \mathbb{I} \rightarrow \mathbb{S}$  and  $\text{pif}$ .
- 8 Same operational rules.

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# Computational adequacy

$\llbracket M \rrbracket = x$  iff for every rational number  $p$ , we have that

$$p < x \iff (p < M) \Downarrow \top.$$

# Definability results

There are programs:

①  $x \oplus y = (x + y)/2$ , **min**, **max**, ....

②  $\exists, \forall: (\text{Cantor} \rightarrow S) \rightarrow S$ .

③  $\int: (\text{Cantor} \rightarrow I) \rightarrow I$ .

# Definability results

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Based on papers:

- ① PCF extended with real numbers, 1996.
- ② Integration in Real PCF (with Edalat), 2000.
- ③ Synthetic topology of data types and classical spaces, 2004.
- ④ Exhaustible sets in higher-computation, 2008.

# Some code

$$\exists(p) = p(\perp) \vee (\exists(\lambda s. p(\text{cons False } s)) \vee \exists(\lambda s. p(\text{cons True } s)))$$

$$\forall(p) = p(\text{if } \forall(\lambda s. p(\text{cons False } s)) \wedge \forall(\lambda s. p(\text{cons True } s)) \text{ then } c)$$

$$\int f = \max \left( f(\perp), \int \lambda s. f(\text{cons False } s) \oplus \int \lambda s. f(\text{cons True } s) \right)$$

# Executable program logic

We extend the programming language  $\text{PCF} + \text{S} + \text{I}$  with modal operators.

We get an executable program logic,  $\text{MMP}$ .

# May and must testing

The  $\mathcal{S}$ -valued terms are characteristic functions of open sets:

$$\mathcal{O} \sigma = (\sigma \rightarrow \mathcal{S}).$$

$$\diamond_F^\sigma: \mathcal{O} \sigma \rightarrow \mathcal{O} F \sigma, \quad \text{for } F \in \{\mathcal{H}, \mathcal{P}\},$$

$$\square_F^\sigma: \mathcal{O} \sigma \rightarrow \mathcal{O} F \sigma, \quad \text{for } F \in \{\mathcal{S}, \mathcal{P}\}.$$

**Idea.** If  $u: \mathcal{O} \sigma$  and  $N: \mathcal{P} \sigma$ ,

$$\diamond(u)(N) = \top \iff u(x) = \top \text{ for some outcome } x \text{ of a run of } N$$

and

$$\square(u)(N) = \top \iff u(x) = \top \text{ for all outcomes } x \text{ of runs of } N.$$

# Example

- 1 Want to semi-decide whether  $n: \text{FNat}$  must be prime.
- 2 Write a semi-decision term  $\text{prime}: \text{Nat} \rightarrow \text{S}$ .
- 3 Run, in the executable logic, the ground term  $\Box \text{prime } n$ .

Of course, one can also semi-decide whether  $n$  must be non-prime.

However:

- 1 It doesn't follow that primeness of all outcomes of  $n$  is decidable.
- 2 If  $n$  has at least one non-divergent run, then both must tests diverge.

# Example

Recursively define a term  $f: \text{Nat} \rightarrow \text{P Nat}$  by

$$f(n) = \eta(n) \odot f(n+1),$$

and let  $\text{converge}: \text{Nat} \rightarrow \text{S}$  be a term such that

$$\text{converge}(n) = \top \iff n \neq \perp.$$

Then we intend that

$$\diamond \text{converge}(f(0)) = \top$$

and that

$$\square \text{converge}(f(0)) = \perp$$

but

$$\square \text{converge}(\eta(0) \odot \eta(1)) = \top.$$



# Parallel-convergence is definable from may testing

Taking  $\text{converge}: S \rightarrow S$  as the identity, the function

$$(\vee): S \times S \rightarrow S$$

is characterized by the equation

$$(p \vee q) = \Diamond \text{converge}(\eta(p) \otimes \eta(q)).$$

However, it cannot be defined from must testing.

Notice that  $(p \wedge q) = \Box \text{converge}(\eta(p) \otimes \eta(q))$ .

# Probabilistic testing

Define a type of expectations:

$$\mathcal{E} \sigma = (\sigma \rightarrow \mathbb{I}).$$

We add a constant to the logic:

$$\bigcirc^\sigma : \mathcal{E} \sigma \rightarrow \mathcal{E} \mathbb{V} \sigma.$$

For a  $\{0, 1\}$ -valued term  $u : \mathcal{E} \sigma$  and a term  $N : \mathbb{V} \sigma$ ,

$\bigcirc(u)(N) : \mathbb{I}$  is the **probability** that  $u$  holds for outcomes of runs of  $N$ .

# Example

Recursively define a term  $g: \text{Nat} \rightarrow \text{VNat}$  by

$$g(n) = \eta(n) \oplus g(n+1),$$

Then we intend that

$$\bigcirc \text{converge}(g(0)) = 1$$

and

$$\bigcirc \text{converge}_n(g(0)) = 2^{-n-1}$$

where  $\text{converge}_n: \text{Nat} \rightarrow \text{S}$  is a term such that

$$\text{converge}_n(x) = \top \iff x = n.$$

# Parallel-convergence is definable from probabilistic testing

$$(p \vee q) = 0 < \bigcirc \text{converge}(\eta(p) \oplus \eta(q)).$$

## Example: uniform distribution on $I$

Define a term  $\text{prefix}: I \rightarrow V\ I \rightarrow V\ I$  by

$$\text{prefix } x = V(\lambda y. x \oplus y),$$

Define  $\text{random}: V\ I$  by

$$\text{random} = (\text{prefix } 0\ \text{random}) \oplus (\text{prefix } 1\ \text{random}).$$

For example  $\bigcirc(\lambda x. p < x)\ \text{random} = 1 - p$  for any  $p \in I$ .

# Existential quantification in MMP

Recall that  $\mathcal{O}\sigma = (\sigma \rightarrow \mathbf{S})$

$$\Diamond: \mathcal{O}\sigma \rightarrow \mathcal{O}\mathbf{H}\sigma$$

Define

$$\exists: \mathbf{H}\sigma \rightarrow ((\sigma \rightarrow \mathbf{S}) \rightarrow \mathbf{S})$$

as

$$\exists(C)(u) = \Diamond(u)(C).$$

The idea is that this stands for

$$\exists x \in C. u(x).$$

# Universal quantification in MMP

Similarly, from the must testing operator

$$\Box: \mathcal{O}\sigma \rightarrow \mathcal{O}\mathbf{S}\sigma,$$

we get a term

$$\forall: \mathbf{S}\sigma \rightarrow ((\sigma \rightarrow \mathbf{S}) \rightarrow \mathbf{S}),$$

The Ploktin powertype has both quantifiers.

# Integration in MMP

Recalling that  $\mathcal{E} \sigma = (\sigma \rightarrow \mathbb{I})$ , from the probabilistic testing operator

$$\bigcirc: \mathcal{E} \sigma \rightarrow \mathcal{E} \mathbb{V} \sigma$$

we get a term

$$\int: \mathbb{V} \sigma \rightarrow ((\sigma \rightarrow \mathbb{I}) \rightarrow \mathbb{I})$$

defined by

$$\int_{\nu} u = \bigcirc(u)(\nu).$$

where  $\nu: \mathbb{V} \sigma$  and  $u: \sigma \rightarrow \mathbb{I}$ .



# Operational semantics of the executable logic MMP

- 1 By compositional compilation into its deterministic sub-language  $\text{PCF} + \text{S} + \text{I}$ .
- 2 The translation is the identity on  $\text{PCF} + \text{S} + \text{I}$  terms.
- 3 Reduce may, must and probabilistic testing in  $\text{MMP}$  to quantification and integration in  $\text{PCF} + \text{S} + \text{I}$ .

# Translation of types

This is defined by induction:

$$\begin{aligned}\phi(\gamma) &= \gamma, \\ \phi(\sigma \times \tau) &= \phi(\sigma) \times \phi(\tau), \\ \phi(\sigma \rightarrow \tau) &= \phi(\sigma) \rightarrow \phi(\tau), \\ \phi(F\sigma) &= \mathbf{Cantor} \rightarrow \phi(\sigma).\end{aligned}$$

Recall that  $\mathbf{Cantor} = (\mathbf{Nat} \rightarrow \mathbf{Bool})$ .

(Hence the translation is the identity on  $\mathbf{PCF} + \mathbf{S} + \mathbf{Ic}$  types.)

# Translation of terms

$$\phi(x) = x$$

$$\phi(\lambda x.M) = \lambda x.\phi(M)$$

$$\phi(MN) = \phi(M)\phi(N)$$

$$\phi(\text{PCF} + \text{S} + \text{I constant}) = \text{itself}$$

$$\phi(\text{any fixed-point combinator}) = \text{itself}$$

(Hence the translation is the identity on **PCF + S + I** terms.)

# Translation of choice operators

For  $\star \in \{\otimes, \oplus\}$ , we define

$$\phi(\star) = \lambda(k_0, k_1). \lambda s. \text{if head}(s) \text{ then } k_0(\text{tail}(s)) \text{ else } k_1(\text{tail}(s)).$$

Here  $k_0$  and  $k_1$  range over  $\phi(F\sigma) = \text{Cantor} \rightarrow \phi(\sigma)$ .

# Translation of the modal operators: may

Typing:

$$\begin{aligned}\diamond & : (\sigma \rightarrow \mathbf{S}) \rightarrow (F\sigma \rightarrow \mathbf{S}), \\ \phi(\diamond) & : (\phi(\sigma) \rightarrow \mathbf{S}) \rightarrow ((\text{Cantor} \rightarrow \phi(\sigma)) \rightarrow \mathbf{S}).\end{aligned}$$

We define

$$\phi(\diamond) = \lambda u. \lambda k. \exists s. u(k(s)).$$

Here

$$\underbrace{(\phi(\sigma) \rightarrow \mathbf{S})}_u \rightarrow \underbrace{((\underbrace{\text{Cantor}}_s \rightarrow \phi(\sigma)) \rightarrow \mathbf{S})}_k.$$

The quantification is over the Cantor space.

# Translation of the modal operators: must

Typing:

$$\begin{aligned}\Box & : (\sigma \rightarrow \mathbf{S}) \rightarrow (F\sigma \rightarrow \mathbf{S}), \\ \phi(\Box) & : (\phi(\sigma) \rightarrow \mathbf{S}) \rightarrow ((\text{Cantor} \rightarrow \phi(\sigma)) \rightarrow \mathbf{S}).\end{aligned}$$

We define

$$\phi(\Box) = \lambda u. \lambda k. \forall s. u(k(s)).$$

Here

$$\underbrace{(\phi(\sigma) \rightarrow \mathbf{S})}_u \rightarrow \underbrace{((\underbrace{\text{Cantor} \rightarrow \phi(\sigma)}_s) \rightarrow \mathbf{S})}_k.$$

The quantification is over the Cantor space.

# Translation of the modal operators: probabilistic

Typing:

$$\begin{aligned}\bigcirc & : (\sigma \rightarrow \mathbf{I}) \rightarrow (\forall \sigma \rightarrow \mathbf{I}), \\ \phi(\bigcirc) & : (\phi(\sigma) \rightarrow \mathbf{I}) \rightarrow ((\text{Cantor} \rightarrow \phi(\sigma)) \rightarrow \mathbf{I}).\end{aligned}$$

We define

$$\phi(\bigcirc) = \lambda u. \lambda k. \int u(k(s)) \mathfrak{s}.$$

Here

$$\underbrace{(\phi(\sigma) \rightarrow \mathbf{I})}_u \rightarrow \underbrace{((\underbrace{\text{Cantor}}_s \rightarrow \phi(\sigma)) \rightarrow \mathbf{I})}_k.$$

The integration is over the Cantor space.

# Translation of the monad constructions: functor

$$\phi(Ff) = \lambda k. \lambda s. f(k(s)).$$



# Translation of the monad constructions: unit

$$\phi(\eta_F) = \lambda x. \lambda s. x.$$

# Translation of the monad constructions: multiplication

We consider PCF terms

$\text{evens}, \text{odds}: \text{Cantor} \rightarrow \text{Cantor}$

that take sub-sequences at even and odd indices.

Define:

$$\phi(\mu_F) = \lambda k. \lambda s. k(\text{evens}(s))(\text{odds}(s)).$$

# Translation of the monad constructions: strength

Left as an exercise to the audience.

# Ground evaluation

For **MMP** terms  $M: \sigma$  with  $\gamma \neq \mathbf{I}$  ground, define

$$M \Downarrow v \iff \phi(M) \Downarrow v.$$

# Denotational semantics of the executable logic

As predicted by the audience.

## Types:

- ① Hoare powertype  $\mapsto$  Hoare powerdomain.
- ② Smyth powertype  $\mapsto$  Smyth powerdomain.
- ③ Plotkin powertype  $\mapsto$  Plotkin powerdomain.
- ④ Probabilistic powertype  $\mapsto$  probabilistic powerdomain.

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## Terms:

- 1 These are monads, which have the binary choice operators we need.
- 2 The modal operators correspond to the usual descriptions of the open sets of the powerdomains.
- 3 The probabilistic operator is interpreted by integration.

# Computational adequacy

To establish semi-decidability of may, must and probabilistic testing, we first prove *computational adequacy* of the model:

## Lemma

For any closed MMP-term  $M$  of ground type other than  $\mathbf{I}$ , and all syntactical values  $v$ ,

$$\llbracket M \rrbracket = \llbracket v \rrbracket \iff M \Downarrow v.$$

In particular, for  $M: \mathbf{I}$  closed and  $r \in \mathbb{Q}$ ,

$$r < \llbracket M \rrbracket \iff r < M \Downarrow \top.$$

# Computational adequacy: technical aspects

Because the model is already known to be computationally adequate for the deterministic sub-language  $\text{PCF} + \mathbf{S} + \mathbf{I}$ :

## Lemma

*Computational adequacy holds if and only if  $\llbracket M \rrbracket = \llbracket \phi(M) \rrbracket$  for every closed term  $M$  of ground type.*



# Correctness of the semi-decision procedures

Follows directly from computational adequacy.

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- ② For the proof of correctness, we rely on the **concrete descriptions** of the powerdomains:
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  - ② Set of compact sets (Smyth).
  - ③ Lenses (Plotkin).
  - ④ Continuous valuations with total mass 1 (Probabilistic).

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  - ④ Continuous valuations with total mass 1 (Probabilistic).
- ③ The abstract and concrete descriptions agree only for special kinds of domains.

# Partial results

## Theorem

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- 2 For any *continuous* type  $\sigma$ , *must* testing on terms of type  $S\sigma$  is semi-decidable.
- 3 For any *RSFP* type  $\sigma$ , *may* and *must* testing on terms of type  $P\sigma$  are semi-decidable.

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- 3 For any *RSFP* type  $\sigma$ , *may* and *must* testing on terms of type  $\mathbb{P}\sigma$  are semi-decidable.
- 4 For any *continuous* type  $\sigma$ , *probabilistic testing* on terms of type  $\mathbb{V}\sigma$  is semi-decidable.



## Remark

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- 1 If we hadn't included the probabilistic powertype in our language, we wouldn't have had any of the above difficulties.
- 2 May and must testing would be semi-decidable for all types.
- 3 What causes the restrictions is the presence of the probabilistic powertype.
- 4 But still the restrictions are not severe in practice.
- 5 For example, probabilistic computations on any PCF type of any order have semi-decidable probabilistic testing.

# Syntactical description of some types we account for

Define:

$$\begin{aligned} S &::= \gamma \mid S \times S \mid (C \rightarrow S) \mid \mathsf{H} C \mid \mathsf{S} C, \\ R &::= S \mid R \times R \mid (R \rightarrow R) \mid \mathsf{P} R, \\ C &::= R \mid C \times C \mid \mathsf{V} C. \end{aligned}$$

By a *continuous Scott domain* we mean a bounded complete continuous dcpo.

## Proposition

- ① *The interpretation of an  $S$  type is a continuous Scott domain.*
- ② *The interpretation of an  $R$  type is an RSFP domain.*
- ③ *The interpretation of a  $C$  type is a continuous dcpo.*

# End and summary

May, must and probabilistic testing are semi-decidable for a large class of higher types.



# Additional slides

# Example

Let  $(\sigma, f_1, \dots, f_n, p_1, \dots, p_n)$  be an IFS with probabilities.

Its invariant measure  $\nu: \mathbf{V} \sigma$  can be defined as

$$\nu = \text{weighted-choice}(p_1, \dots, p_n)(\mathbf{V}(f_1)(\nu), \dots, \mathbf{V}(f_n)(\nu)),$$

Scriven (MFPS 2008) developed a PCF program for computing integrals of functions  $u: \sigma \rightarrow \mathbf{I}$  with respect to the invariant measure.

Here we get the alternative algorithm  $\int_{\nu} u = \mathbf{O}(u)(\nu)$  in the program logic MMP instead.