

Reasoning about Probabilistic Sequential Programs

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Motivation

- reasoning about non-deterministic programs
- new approach: truth values for formulas

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 - Language
 - Semantics
 - Calculus
 - Properties
- 2 The Programming Language
 - Syntax
 - Semantics
- 3 The Hoare Calculus
 - The calculus
 - Soundness
 - Completeness
- 4 Conclusions

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Why EPPL

- two-layered design (exogenous approach)
- classical propositional logic at the lower level
- probabilistic logic built at the higher level

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Real-closed fields

Definition

A *real closed field* is an ordered field \mathcal{K} where:

- every non-negative element of the \mathcal{K} has a square root in \mathcal{K} ;
- every polynomial of odd degree with coefficients in \mathcal{K} has at least one solution in \mathcal{K} .

Example

- the set of real numbers with the usual multiplication, addition and order relation;
- the set of computable real numbers with the same operations.

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Setting

- finite range D of real numbers
- finite set $\mathbf{m} = \{0, \dots, m - 1\}$ of indices
- registers $\mathbf{xM} = \{\mathbf{xm}_k \mid k \in \mathbf{m}\}$ containing real values
- registers $\mathbf{bM} = \{\mathbf{bm}_k \mid k \in \mathbf{m}\}$ containing booleans
- variables $B = \{B_k : k \in \mathbb{N}\}$ ranging over truth values
- variables $X = \{X_k : k \in \mathbb{N}\}$ ranging over D
- real-closed field \mathcal{K} with set of algebraic numbers \mathcal{A}
- logical variables $Y = \{y_k : k \in \mathbb{N}\}$ ranging over \mathcal{K}

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Language

Real terms (with $c \in D$)

$$t ::= c \mid \mathbf{xm} \mid X \mid (t + t) \mid (t t)$$

Classical state formulas

$$\gamma ::= \mathbf{bm} \mid B \mid (t \leq t) \mid \mathbf{ff} \mid (\gamma \Rightarrow \gamma)$$

Probability terms (with $r \in \mathcal{A}$)

$$p ::= r \mid y \mid \tilde{r} \mid (f\gamma) \mid (p + p) \mid (p p)$$

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$$\eta ::= (p \leq p) \mid \mathbf{fff} \mid (\eta > \eta)$$

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Useful notions

Definition

An *analytical term* is a term without occurrences of probability terms.

$$a ::= r \mid y \mid \tilde{r} \mid (a + a) \mid (aa)$$

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An *analytical formula* is a formula without occurrences of probability terms.

$$\kappa ::= (a \leq a) \mid \text{fff} \mid (\kappa \supset \kappa)$$

$(\Box\gamma)$ stands for the formula $((\int\gamma) = (\int\text{tt}))$

$(\Diamond\gamma)$ stands for the formula $(\ominus(\Box(\neg\gamma)))$

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Valuations

Definition

A *valuation* is a map that provides values to the memory variables and corresponding logical variables. The set of all valuations is denoted by \mathcal{V} .

The denotation $\llbracket t \rrbracket_v$ of a real term t given a valuation v is defined inductively as expected.

Satisfaction $v \Vdash_c \gamma$ of a classical state formula γ by a valuation v is also defined inductively as usual.

Definition

The *extent* of a classical state formula γ in a set V of valuations is

$$|\gamma|_V = \{v \in V \mid v \Vdash_c \gamma\}.$$

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Measure functions

Definition

A finitely additive, discrete and bounded \mathcal{K} -measure μ on a set X is a map from X to \mathcal{K}^+ such that:

- $\mu(\emptyset) = 0$;
- if $U_1 \cap U_2 = \emptyset$, then $\mu(U_1 \cup U_2) = \mu(U_1) + \mu(U_2)$.

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Interpretation

Definition

A *generalized probabilistic state* consists of a real closed field \mathcal{K} and a finitely additive, discrete and finite \mathcal{K} -measure over $\wp\mathcal{V}$.

Given a classical formula γ we define

$$\mu_\gamma = \lambda V. \mu(|\gamma|_V).$$

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Given a real closed field \mathcal{K} , a \mathcal{K} -assignment is a map $\rho : Y \rightarrow \mathcal{K}$.

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Denotation of probability terms

$$\llbracket r \rrbracket_{K,\mu}^\rho = r$$

$$\llbracket y \rrbracket_{K,\mu}^\rho = \rho(y)$$

$$\llbracket (\int \gamma) \rrbracket_{K,\mu}^\rho = \mu(|\gamma|v)$$

$$\llbracket p_1 + p_2 \rrbracket_{K,\mu}^\rho = \llbracket p_1 \rrbracket_{K,\mu}^\rho + \llbracket p_2 \rrbracket_{K,\mu}^\rho$$

$$\llbracket p_1 p_2 \rrbracket_{K,\mu}^\rho = \llbracket p_1 \rrbracket_{K,\mu}^\rho \times \llbracket p_2 \rrbracket_{K,\mu}^\rho$$

Satisfaction of probabilistic formulas

$$(K, \mu)\rho \Vdash (p_1 \leq p_2) \text{ iff } \llbracket p_1 \rrbracket_{K,\mu}^\rho \leq \llbracket p_2 \rrbracket_{K,\mu}^\rho$$

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$$\begin{aligned} \llbracket r \rrbracket_{K,\mu}^\rho &= r \\ \llbracket y \rrbracket_{K,\mu}^\rho &= \rho(y) \\ \llbracket (\int \gamma) \rrbracket_{K,\mu}^\rho &= \mu(|\gamma| \nu) \\ \llbracket p_1 + p_2 \rrbracket_{K,\mu}^\rho &= \llbracket p_1 \rrbracket_{K,\mu}^\rho + \llbracket p_2 \rrbracket_{K,\mu}^\rho \\ \llbracket p_1 p_2 \rrbracket_{K,\mu}^\rho &= \llbracket p_1 \rrbracket_{K,\mu}^\rho \times \llbracket p_2 \rrbracket_{K,\mu}^\rho \end{aligned}$$

Satisfaction of probabilistic formulas

$$\begin{aligned} (K, \mu)\rho \Vdash (p_1 \leq p_2) &\text{ iff } \llbracket p_1 \rrbracket_{K,\mu}^\rho \leq \llbracket p_2 \rrbracket_{K,\mu}^\rho \\ (K, \mu)\rho \not\Vdash \text{fff} & \\ (K, \mu)\rho \Vdash (\eta_1 \supset \eta_2) &\text{ iff } (K, \mu)\rho \Vdash \eta_2 \text{ or } (K, \mu)\rho \not\Vdash \eta_1 \end{aligned}$$

Auxiliary notions

Definition

A classical state formula γ is said to be *valid* if it holds for all valuations $v \in \mathcal{V}$.

Example

$$((x1 \leq x2) \wedge (x1 > 0)) \Rightarrow (x1^2 \leq x2^2)$$

Since D is finite, the set of valid classical state formulas is recursive.

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A probabilistic formula η is said to be a *probabilistic tautology* if there exists a propositional tautology β such that η is obtained from β by replacing all occurrences of \perp by fff , \rightarrow by \supset and each propositional symbol (uniformly) by a probabilistic state formula.

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$$((f(x_1 \leq x_2)) < 1) \supset (((f(x_1 \leq x_2)) < 1) \cap \text{fff})$$

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[CTaut] $\vdash (\Box\gamma)$ for each valid state formula γ

[PTaut] $\vdash \eta$ for each probabilistic tautology η

[RCF] $\vdash \kappa \frac{\gamma}{p}$ for any valid analytical formula κ

[Meas \emptyset] $\vdash ((\int \mathbb{f}) = 0)$

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Theorem

The axiom system of EPPL is sound: if $\vdash \eta$, then $\models \eta$.

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Straightforward from the definition of the semantics. □

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The central result is to show that if η is consistent then there is a model $(\mathcal{K}, \mu)\rho$ such that $(\mathcal{K}, \mu)\rho \models \eta$. The decidability follows by showing that the consistency of a formula is decidable. \square

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Construction of the model

- 1 compute the (finite) set of valuations over the memory cells and the logical variables in the sets B and X occurring in η and let this set of valuations be V ;
- 2 let κ_1 be the analytical formula obtained from η by effectively replacing measure terms ($\int \gamma$) by sums $\sum_{v \models \tau \gamma, v \in V} y_v$ where y_v represents the probability of the valuation v ;
- 3 let κ be the analytical formula $\kappa_1 \cap \bigcap_{y_v | v \in V} (0 \leq y_v)$;
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$s ::= \text{skip} \mid \mathbf{xm} \leftarrow t \mid \mathbf{bm} \leftarrow \gamma \mid \text{toss}(\mathbf{bm}, r) \mid s; s \mid \text{if } \gamma \text{ then } s \text{ else } s$

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An *expression* is either a term t or a classical state formula γ .

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$\llbracket \gamma \rrbracket_v = \text{tt}$ if $v \Vdash_c \gamma$ and $\llbracket \gamma \rrbracket_v = \text{ff}$ otherwise

if m is a memory cell and e is an expression of the same type, then $\delta_e^m(v)$ assigns the value $\llbracket e \rrbracket_v$ to the cell m and coincides with v elsewhere

$$(\mathcal{K}, \mu_1) + (\mathcal{K}, \mu_2) = (\mathcal{K}, \mu_1 + \mu_2)$$

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Denotation of programs

The denotation of a program s is a function on generalized probabilistic states.

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$$\Psi ::= \eta \mid \{\eta\} s \{\eta\}$$

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A Hoare assertion Ψ is *semantically valid* ($\models_h \Psi$) if $(\mathcal{K}, \mu)\rho \Vdash_h \Psi$ for every generalized probabilistic state (\mathcal{K}, μ) and any \mathcal{K} -assignment ρ .

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Tossed terms

Let \mathbf{bm} be a memory cell, $r \in \mathcal{A}$ be a constant and p be a probabilistic term.

The term $\text{toss}(\mathbf{bm}, r; p)$ is the term obtained from p by replacing every occurrence of each measure term $(\int \gamma)$ by $\tilde{r}(\int \gamma_{\#}^{\mathbf{bm}}) + (1 - \tilde{r})(\int \gamma_{\#}^{\mathbf{bm}})$.

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$$\text{toss}(\mathbf{bm}, r; (pp')) = (\text{toss}(\mathbf{bm}, r; p) \text{toss}(\mathbf{bm}, r; p'))$$

Tossed terms

Let \mathbf{bm} be a memory cell, $r \in \mathcal{A}$ be a constant and p be a probabilistic term.

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Conditioned terms

Let γ be classical state formula and p be a probabilistic term.

The term (p/γ) is the term obtained from p by replacing every occurrence of each measure term $(\int \gamma')$ by $(\int(\gamma' \wedge \gamma))$.

$$r/\gamma = r$$

$$y/\gamma = y$$

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Axioms

[TAUT] $\vdash \eta$ if η is an EPPL theorem

[FREE] $\vdash \{\kappa\} s \{\kappa\}$ if κ is an analytical formula

[SKIP] $\vdash \{\eta\} \text{skip} \{\eta\}$

[ASGR] $\vdash \{\eta_t^{xm}\} xm \leftarrow t \{\eta\}$

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Inference rules

$$[\text{SEQ}] \quad \{\eta_0\} s_1 \{\eta_1\}, \{\eta_1\} s_2 \{\eta_2\} \vdash \{\eta_0\} s_1; s_2 \{\eta_2\}$$

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$$[\text{ELIMV}] \quad \{\eta_1 \cap (y = p)\} s \{\eta_2\} \vdash \{\eta_1 \overset{y}{p}\} s \{\eta_2\}$$

y does not occur in p or η_2

$$[\text{CONS}] \quad \eta_0 \supset \eta_1, \{\eta_1\} s \{\eta_2\}, \eta_2 \supset \eta_3 \vdash \{\eta_0\} s \{\eta_3\}$$

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Substitution Lemma for classical valuations

Lemma

*For any valuation $v \in \mathcal{V}$, any classical state formula γ , any memory cell m (**xm** or **bm**) and term e of the same type,*

$$v_{[[e]]_v}^m \Vdash_c \gamma \text{ iff } v \Vdash_c \gamma_e^m.$$

Proof.

Induction on the structure of γ . □

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Substitution Lemma for assignment

Lemma

Let (\mathcal{K}, μ) be a generalized probabilistic structure and ρ be a \mathcal{K} -assignment. Given a memory cell m and a term e of the same type, let $\mu' = \mu \circ (\delta_e^m)^{-1}$. Then

$$\llbracket \int \gamma \rrbracket_{(\mathcal{K}, \mu')}^\rho = \llbracket \int \gamma_e^m \rrbracket_{(\mathcal{K}, \mu)}^\rho$$

for any classical state formula γ .

Furthermore, for any probabilistic term p ,

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$$(\delta_e^m)^{-1}(|\gamma|v) = |\gamma_e^m|v \text{ and hence } \mu((\delta_e^m)^{-1}(|\gamma|v)) = \mu(|\gamma_e^m|v).$$

Therefore, by definition,

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The result is extended to probabilistic terms and formulas by induction. □

Corollary

Axioms ASGB and ASGR are sound.

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*Axioms **ASGB** and **ASGR** are sound.*

Substitution Lemma for probabilistic tosses

Lemma

Let (K, μ) and ρ be as before, $r \in \mathcal{A}$ be a constant and $\mu' = \tilde{r}\mu \circ (\delta_{\dagger}^{\mathbf{bm}})^{-1} + (1 - \tilde{r})\mu \circ (\delta_{\ddagger}^{\mathbf{bm}})^{-1}$.

For any classical state formula γ ,

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Substitution Lemma for probabilistic tosses

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Let (K, μ) and ρ be as before, $r \in \mathcal{A}$ be a constant and $\mu' = \tilde{r}\mu \circ (\delta_{\#}^{\mathbf{bm}})^{-1} + (1 - \tilde{r})\mu \circ (\delta_{\#}^{\mathbf{bm}})^{-1}$.

For any classical state formula γ ,

$$\llbracket (\int \gamma) \rrbracket_{(K, \mu')}^{\rho} = \tilde{r} \llbracket (\int \gamma_{\#}^{\mathbf{bm}}) \rrbracket_{(K, \mu)}^{\rho} + (1 - \tilde{r}) \llbracket (\int \gamma_{\#}^{\mathbf{bm}}) \rrbracket_{(K, \mu)}^{\rho}.$$

Furthermore, for any probabilistic term p ,

$$\llbracket p \rrbracket_{(K, \mu')}^{\rho} = \llbracket \text{toss}(\mathbf{bm}, r; p) \rrbracket_{(K, \mu)}^{\rho},$$

and, for any probabilistic formula η ,

$$(K, \mu')\rho \Vdash \eta \text{ iff } (K, \mu)\rho \Vdash \text{toss}(\mathbf{bm}, r; \eta).$$

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Proof.

Let $\mu_1 = \mu \circ (\delta_{\text{tt}}^{\text{bm}})^{-1}$ and $\mu_2 = \mu \circ (\delta_{\text{ff}}^{\text{bm}})^{-1}$. Then

$$\llbracket (f\gamma) \rrbracket_{(\mathcal{K}, \mu')}^{\rho} = \tilde{r} \llbracket (f\gamma) \rrbracket_{(\mathcal{K}, \mu_1)}^{\rho} + (1 - \tilde{r}) \llbracket (f\gamma) \rrbracket_{(\mathcal{K}, \mu_2)}^{\rho}$$

by definition. Also

$$\llbracket (f\gamma) \rrbracket_{(\mathcal{K}, \mu_1)}^{\rho} = \llbracket (f\gamma_{\text{tt}}^{\text{bm}}) \rrbracket_{(\mathcal{K}, \mu)}^{\rho} \text{ and } \llbracket (f\gamma) \rrbracket_{(\mathcal{K}, \mu_2)}^{\rho} = \llbracket (f\gamma_{\text{ff}}^{\text{bm}}) \rrbracket_{(\mathcal{K}, \mu)}^{\rho}.$$

The claim for probabilistic terms and probabilistic formulas then follows by induction. \square

Corollary

Axiom TOSS is sound.

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Let $\mu_1 = \mu \circ (\delta_{\text{tt}}^{\text{bm}})^{-1}$ and $\mu_2 = \mu \circ (\delta_{\text{ff}}^{\text{bm}})^{-1}$. Then

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Corollary

Axiom **TOSS** is sound.

Soundness of \int FREE

Lemma

For any statement s , any analytical formula κ , any generalized state (\mathcal{K}, μ) and \mathcal{K} assignment ρ ,

$$(\llbracket s \rrbracket(\mathcal{K}, \mu))\rho \Vdash \kappa \text{ iff } (\mathcal{K}, \mu)\rho \Vdash \kappa.$$

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Lemma

For any generalized state (\mathcal{K}, μ) , \mathcal{K} -assignment ρ and classical state formulas γ and γ' ,

$$\llbracket (\int \gamma') / \gamma \rrbracket_{(\mathcal{K}, \mu)}^\rho = \llbracket (\int \gamma') \rrbracket_{(\mathcal{K}, \mu_\gamma)}^\rho.$$

Furthermore, for any probability term p ,

$$\llbracket p / \gamma \rrbracket_{(\mathcal{K}, \mu)}^\rho = \llbracket p \rrbracket_{(\mathcal{K}, \mu_\gamma)}^\rho,$$

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By definition,

$$\begin{aligned} \llbracket (f\gamma') \rrbracket_{(\mathcal{K}, \mu_\gamma)}^\rho &= \mu_\gamma(|\gamma'| \nu) = \mu(|\gamma'| \nu \cap |\gamma| \nu) = \mu(|\gamma' \wedge \gamma| \nu) = \\ &\llbracket (f\gamma')/\gamma \rrbracket_{(\mathcal{K}, \mu)}^\rho. \end{aligned}$$

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Corollary

Given probabilistic state formulas η_1 and η_2 , programs s_1 and s_2 , variables $y_1 \in Y$ and $y_2 \in Y$ and a classical state formula γ ,

$$\models_h \{ \eta_1 \} s_1 \{ y_1 = (\int \gamma) \} \text{ and } \models_h \{ \eta_2 \} s_2 \{ y_2 = (\int \gamma) \}$$

iff, for any classical state formula γ_0 ,

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Proof.

Suppose that $(\mathcal{K}, \mu)\rho \Vdash \eta_1 \vee_{\gamma_0} \eta_2$. Then $(\mathcal{K}, \mu)\rho \Vdash \eta_1/\gamma_0$ and $(\mathcal{K}, \mu)\rho \Vdash \eta_2/(\neg\gamma_0)$. Thus, $(\mathcal{K}, \mu_{\gamma_0})\rho \Vdash \eta_1$ and $(\mathcal{K}, \mu_{(\neg\gamma_0)})\rho \Vdash \eta_2$. Let $(\mathcal{K}, \mu_1) = \llbracket s_1 \rrbracket(\mathcal{K}, \mu_{\gamma_0})$, $(\mathcal{K}, \mu_2) = \llbracket s_2 \rrbracket(\mathcal{K}, \mu_{(\neg\gamma_0)})$ and $\mu' = \mu_1 + \mu_2$.

Since $\Vdash_h \{\eta_1\} s_1 \{y_1 = (f\gamma)\}$ and $(\mathcal{K}, \mu_{\gamma_0})\rho \Vdash \eta_1$, it follows that $(\mathcal{K}, \mu_1) \Vdash_h y_1 = (f\gamma)$. Thus, by definition $\rho(y_1) = \mu_1(|\gamma|v)$.

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Hence,

$\mu'(|\gamma|v) = \mu_1(|\gamma|v) + \mu_2(|\gamma|v) = \rho(y_1) + \rho(y_2) = \rho(y_1 + y_2)$ and $(\mathcal{K}, \mu')\rho \Vdash (y_1 + y_2 = (f\gamma))$ as required. \square

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Soundness of ELIMV

Lemma

Let $k = \llbracket p \rrbracket_{(\mathcal{K}, \mu)}^\rho$ and $\rho_1 = \rho_k^y$. Then:

- for any probabilistic term p_0 , $\llbracket p_0 \rrbracket_{(\mathcal{K}, \mu)}^{\rho_1} = \llbracket p_0 \rrbracket_{(\mathcal{K}, \mu)}^\rho$;
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Let p_0 be a variable y_0 .

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The rest follows by induction. \square

Soundness of ELIMV

Lemma

Let $k = \llbracket p \rrbracket_{(\mathcal{K}, \mu)}^\rho$ and $\rho_1 = \rho_k^y$. Then:

- for any probabilistic term p_0 , $\llbracket p_0 \rrbracket_{(\mathcal{K}, \mu)}^{\rho_1} = \llbracket p_0 \rrbracket_{(\mathcal{K}, \mu)}^\rho$;
- for any probabilistic formula η , $(\mathcal{K}, \mu)\rho_1 \Vdash \eta$ iff $(\mathcal{K}, \mu)\rho \Vdash \eta^y$.

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Given y not occurring in either p or in η ,

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Since $\Vdash_h \{\eta_1 \cap (y = p)\} s \{\eta_2\}$ and ρ_1 and ρ differ only in the value assigned to y , which does not occur in η_2 , $(\llbracket s \rrbracket(\mathcal{K}, \mu))\rho \Vdash \eta_2$ as required. □

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Soundness of the calculus

Theorem

If $\vdash \Psi$ then $\models_h \Psi$.

Proof.

By induction on the length of the derivation of $\vdash \Psi$ using the previous lemmas. □

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Preterms

$$\text{pt}(\text{skip}, p) = p$$

$$\text{pt}(\mathbf{bm} \leftarrow \gamma, p) = p_{\gamma}^{\mathbf{bm}}$$

$$\text{pt}(\mathbf{xm} \leftarrow t, p) = p_t^{\mathbf{xm}}$$

$$\text{pt}(\text{toss}(\mathbf{bm}, r), p) = \text{toss}(\mathbf{bm}, r; p)$$

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Properties of preterms

Lemma

$$\llbracket \text{pt}(s, \rho) \rrbracket_{(\mathcal{K}, \mu)}^\rho = \llbracket \rho \rrbracket_{\llbracket s \rrbracket}^\rho_{(\mathcal{K}, \mu)}.$$

Weakest preconditions

$$\begin{aligned}\text{wp}(s, \text{fff}) &= \text{fff} \\ \text{wp}(s, (p_1 \leq p_2)) &= (\text{pt}(s, p_1) \leq \text{pt}(s, p_2)) \\ \text{wp}(s, (\eta_1 \supset \eta_2)) &= (\text{wp}(s, \eta_1) \supset \text{wp}(s, \eta_2))\end{aligned}$$

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$$(\mathcal{K}, \mu)\rho \Vdash_h \text{wp}(s, \eta) \text{ iff } (\llbracket s \rrbracket(\mathcal{K}, \mu))\rho \Vdash_h \eta.$$

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Corollary

$$\models_h \{\eta'\} s \{\eta\} \text{ iff } \models (\eta' \supset \text{wp}(s, \eta)).$$

Proof.

(\Rightarrow) Suppose that $\models_h \{\eta'\} s \{\eta\}$ and $(\mathcal{K}, \mu)\rho \Vdash \eta'$.

Then $(\llbracket s \rrbracket)(\mathcal{K}, \mu)\rho \Vdash \eta$, hence $(\mathcal{K}, \mu)\rho \Vdash \text{wp}(s, \eta)$. Therefore $\models (\eta' \supset \text{wp}(s, \eta))$.

(\Leftarrow) Suppose that $\models (\eta' \supset \text{wp}(s, \eta))$ and $(\mathcal{K}, \mu)\rho \Vdash \eta'$.

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$$\models_h \{\eta'\} s \{\eta\} \text{ iff } \models (\eta' \supset \text{wp}(s, \eta)).$$

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Lemma

For any probabilistic term p , statement s and variable y ,

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For any statement s and any conditional-free formula η ,

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Decidability. By soundness and completeness, $\vdash \{\eta'\} s \{\eta\}$ iff $\models_h \{\eta'\} s \{\eta\}$. By completeness of EPPL and the properties of weakest preconditions, it follows that $\vdash \{\eta'\} s \{\eta\}$ iff $\vdash (\eta' \supset wp(s, \eta))$. The decidability is now a consequence of the decidability of EPPL and the fact that $wp(s, \eta)$ can be computed algorithmically. □

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