



# Formalising Semantics for Expected Running Time of Probabilistic Programs

(Rough Diamond)

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clarified semantics

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clarified semantics – different proofs – fixed proofs

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| ITE  $g$   $p_1$   $p_2$                    $g :: \sigma \Rightarrow \text{bool}$

# Probabilistic Guarded Command Language (pGCL)

$$\begin{array}{lcl} \sigma \text{ pgcl} & = & \perp \mid \epsilon \mid \sharp \\ & | & x : \sim \mathcal{D} \text{ or } x := \text{expr} \quad \text{"Assign } (\sigma \Rightarrow \sigma \text{ pmf)"} \\ & | & p_1 ; p_2 \\ & | & p_1 \mid p_2 \\ & | & \text{ITE } g \ p_1 \ p_2 \qquad \qquad g :: \sigma \Rightarrow \text{bool} \\ & | & \text{WHILE } g \text{ DO } p \end{array}$$

## Denotational Semantics (Expected Running Time)

$\text{ert} :: \sigma \text{ pgcl} \Rightarrow (\sigma \Rightarrow \overline{\mathbb{R}}_{\geq 0}) \Rightarrow (\sigma \Rightarrow \overline{\mathbb{R}}_{\geq 0})$

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Values we want assigned to a *terminal state*

# Denotational Semantics (Expected Running Time)

Values computed for the a *starting state*

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$$\text{ert (WHILE } g \text{ DO } p) \quad c = \text{lfp } (\lambda W x. 1 + \text{if } g x \text{ then } \text{ert } p W x \text{ else } c x)$$

# Interjection: Markov decision processes

Automata with probabilistic and non-deterministic choice

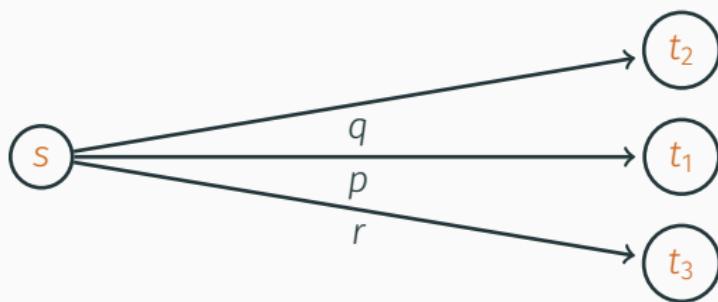
$$K :: \sigma \Rightarrow \sigma \text{ pmf set} \quad Ks \neq \emptyset$$



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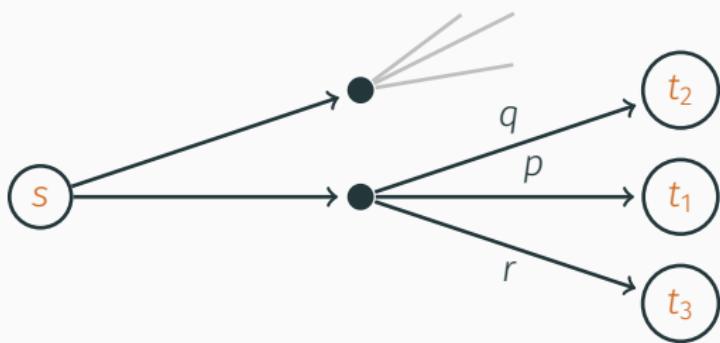
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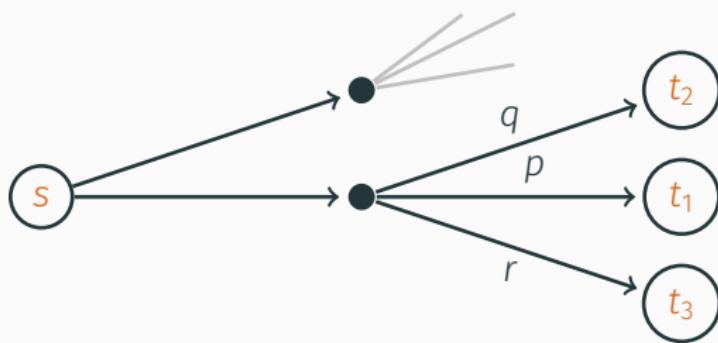
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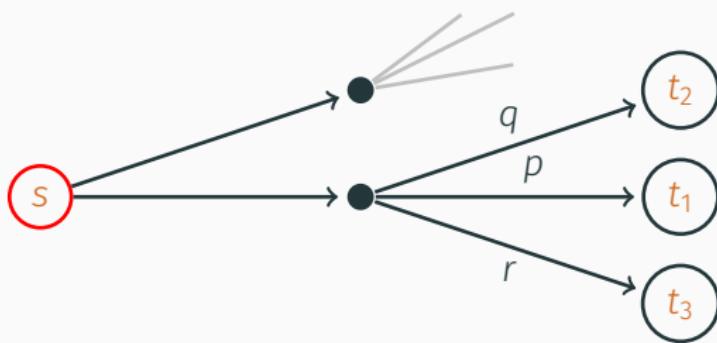
Construct maximal expectation  $\hat{\mathbb{E}}_s[f]$  of a (cost) function  $f$ :

$$\hat{\mathbb{E}}_s[f] = \bigsqcup_{D \in K_s} \int_t \hat{\mathbb{E}}_t[\lambda \omega. f(t \cdot \omega)] dD$$

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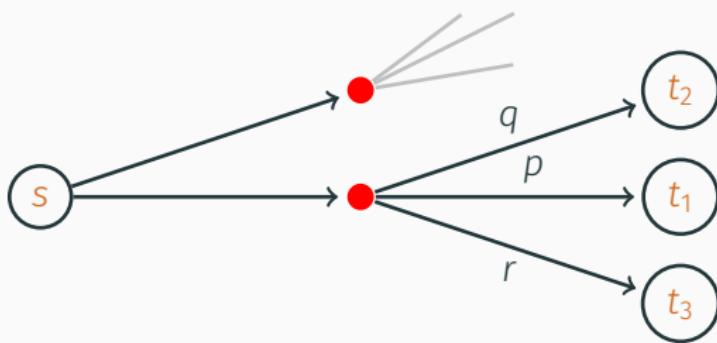
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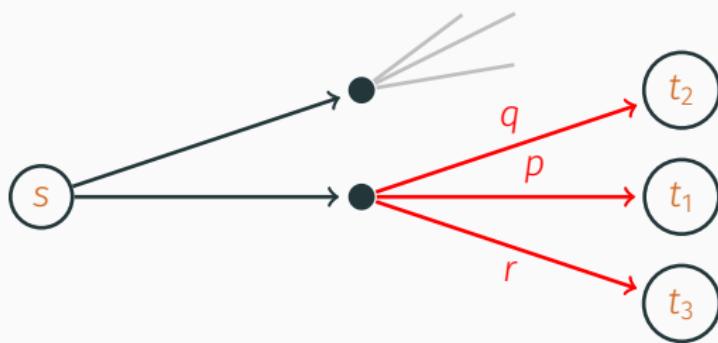
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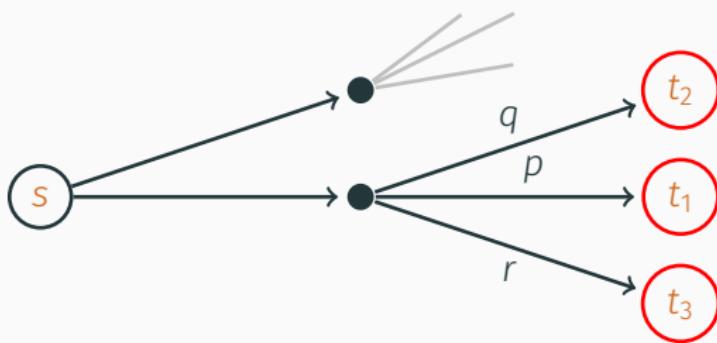
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$K(\text{WHILE } g \text{ DO } p, s) = \text{if } g \ s \text{ then } \ll p; \text{ WHILE } g \text{ DO } p, s \gg \text{ else } \ll\perp, s\gg$

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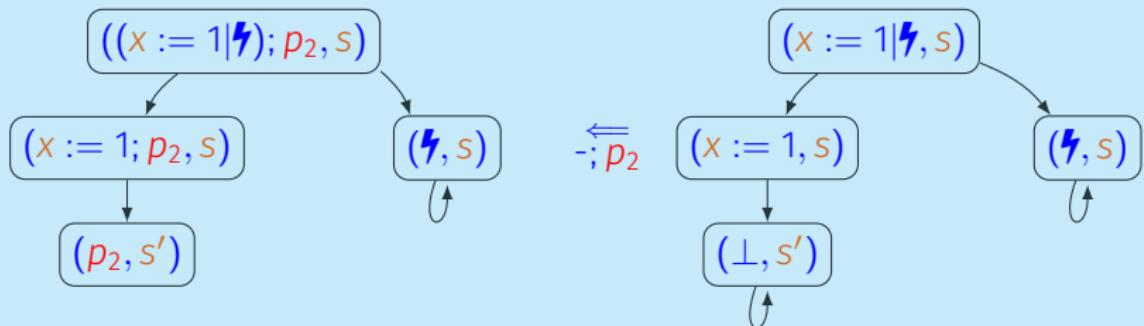
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$$K(p_1; p_2, s) = \left[ \begin{array}{cc} (\perp, s'). & (p_2, s') \\ \lambda (\lnot, s'). & (\lnot, s') \\ (p', s'). & (p'; p_2, s') \end{array} \right] K(p_1, s)$$

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# Traces

Trace	cost <sub>stream</sub>
$((x := 1   \sharp); y := 0, s)$	0

Cost per stream:

$$\text{cost}_{\text{stream}} f (s \cdot \omega) \stackrel{\text{lfp}}{=} \text{cost } fs (\text{cost}_{\text{stream}} f \omega)$$

# Traces

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$((x := 1 \#); y := 0, s)$	0
$(x := 1; y := 0, s)$	+
	1

Cost per stream:

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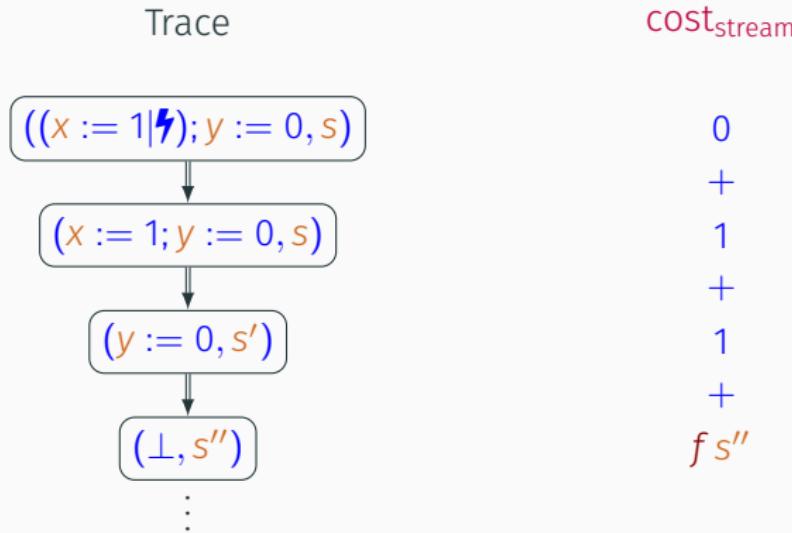
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Trace	cost <sub>stream</sub>
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$(y := 0, s')$	1
	+
	1

Cost per stream:

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# Interjection: Least Fixed Points

Theorem (Transfer rule for least fixed points)

$$\frac{\sqcup\text{-continuous } \alpha, f, g \quad \alpha \perp = \perp \quad \alpha \circ f = g \circ \alpha}{\alpha(\text{lfp } f) = \text{lfp } g}$$

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$$\begin{aligned}\alpha(\text{lfp } f) &= \alpha \circ f \circ f \circ f \circ f \circ f \circ \dots \circ \perp \\ &= g \circ g \circ g \circ \alpha \circ f \circ f \circ \dots \circ \perp \\ &= g \circ g \circ g \circ g \circ g \circ \dots \circ \alpha \perp \\ &= \text{lfp } g\end{aligned}$$

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$$\alpha f = \hat{\mathbb{E}}_s[f] \quad \text{for } f \text{ Borel-measurable}$$

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$$\hat{\mathbb{E}}_{(p,s)}[\text{cost}_{\text{stream}} f] = \text{lfp} \left( \bigsqcup_{K s} \int \text{cost} \right) p s$$

# Correspondence Proof

## Theorem

$$\hat{\mathbb{E}}_{(p,s)}[\text{cost}_{\text{stream } f}] = \text{ert } p \text{ } f \text{ } s$$

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Proof by induction on  $p$ :

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Proof by induction on  $p$ :

$p_1; p_2$  – Antisymmetry then fixed point induction

$$\begin{aligned}\hat{\mathbb{E}}_{(p_1,s)}[\text{cost}_{\text{stream}} (\hat{\mathbb{E}}_{(p_2,.)}[\text{cost}_{\text{stream}} f])] &= \\ \hat{\mathbb{E}}_{(p_1;p_2,s)}[\text{cost}_{\text{stream}} f]\end{aligned}$$

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WHILE  $g$  DO  $p_1$  – Fixed point massaging

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Proof by induction on  $p$ :

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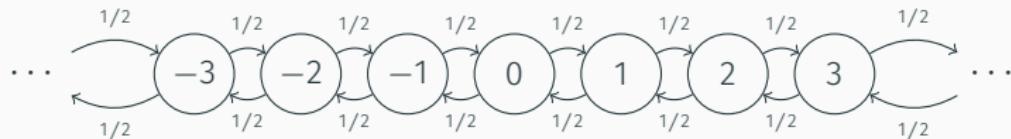
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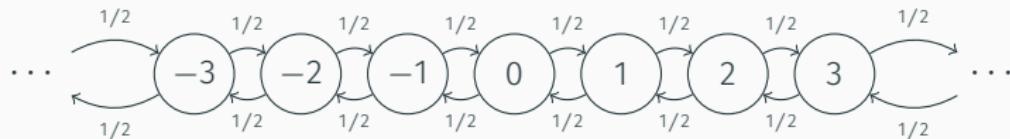
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# Simple Symmetric Random Walk (ssrw)



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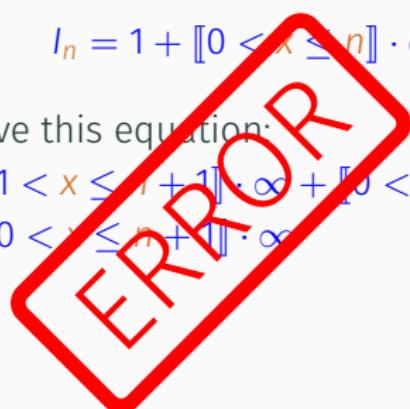
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Fails for  $x = 1$  and  $n = 0$ .

# Simple Symmetric Random Walk – our Solution

Operational semantics – trace representation

Use  $H i j = \hat{\mathbb{E}}_{(ssrw\ j,i)}[\text{cost}_{\text{stream}}]$  to prove

$$H i j = H i k + H k j \quad \text{for} \quad i \leq k \leq j$$

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# Coupon Collector

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x := 0, i := 0, cp := [F, ..., F];  
N times  
WHILE x < N DO  
    WHILE cp[i] DO  
        i := U({1, ..., N});  
        cp[i] := T, x := x + 1
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c := 0, b := F;  
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KKMO use loop invariants to prove running time

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