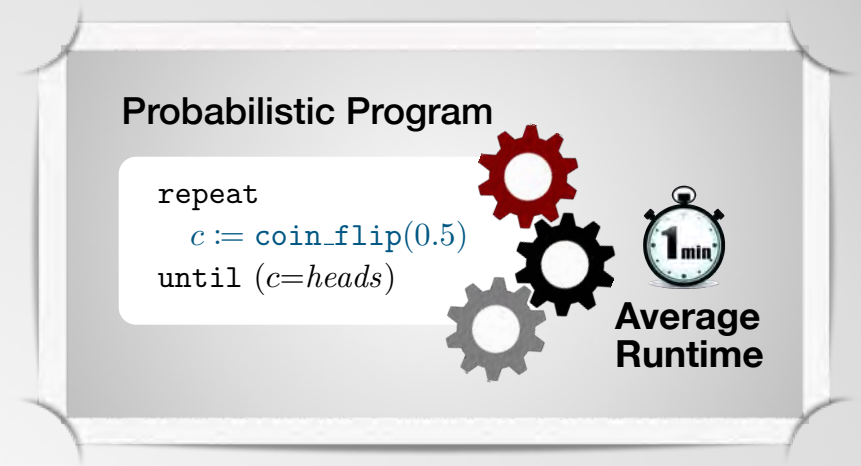


# Run-Time Analysis of Probabilistic Programs



**Joost-Pieter Katoen**

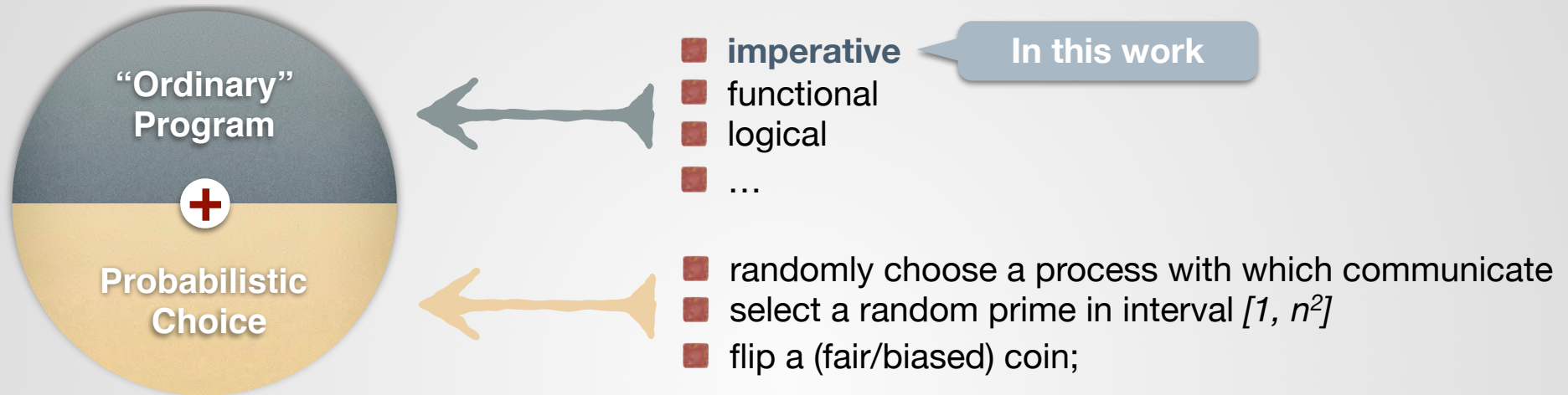
*joint work with: Benjamin Kaminski,  
Christoph Matheja, and Federico Olmedo*



IFIP WG2.2 Meeting Singapore, September 2016

# Probabilistic Programs — Basics

What is a probabilistic program?



Program behaviour (**input-output relation** + **runtime**) is determined by the outcome of its probabilistic choices.

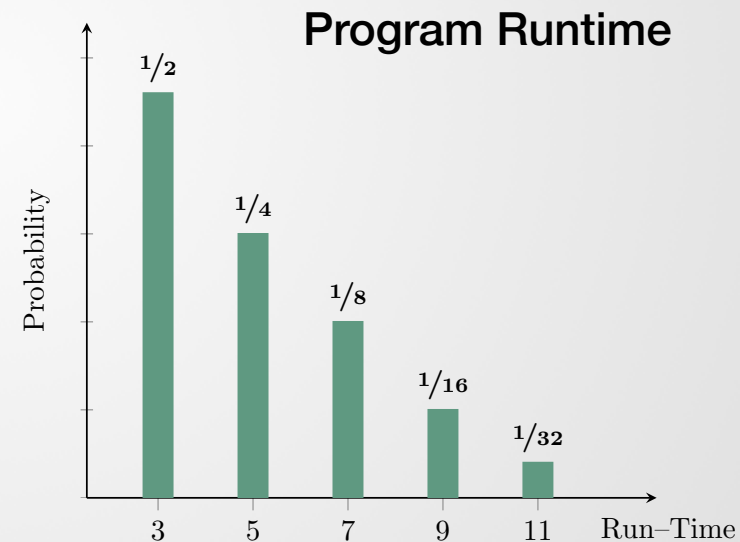
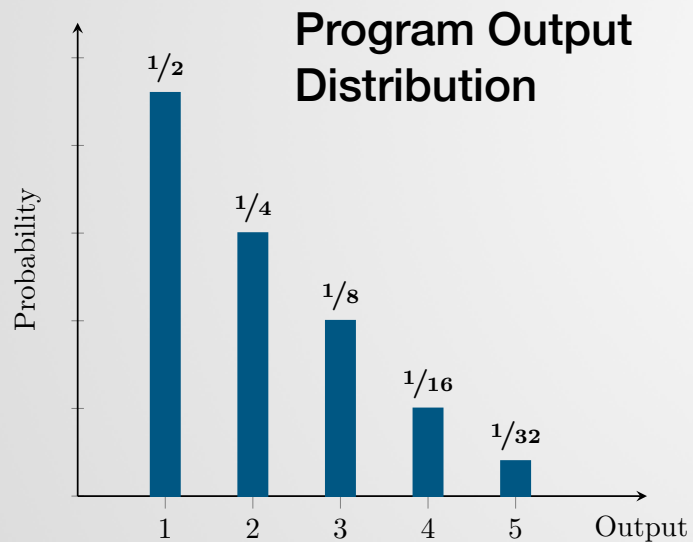
- Program **output** is a **probability distribution**:  $v_1$  with probability  $p_1$ ,  $v_2$  with probability  $p_2$ , etc
- Program **runtime** is a **random variable**:  $t_1$  with probability  $p_1$ ,  $t_2$  with probability  $p_2$ , etc

# Probabilistic Programs — Example

Probabilistic program  
that simulates a  
geometric distribution



```
 $C_{\text{geo}} :$    $n := 0;$   
            repeat  
               $n := n + 1;$   
               $c := \text{coin\_flip}(0.5)$   
            until  $(c = \text{heads});$   
            return  $n$ 
```



**Average Runtime:**

$$3 \cdot \frac{1}{2} + 5 \cdot \frac{1}{4} + \cdots + (2n+1) \cdot \frac{1}{2^n} + \cdots = 5$$

# Randomisation Allows Speeding Up Algorithms

## Quicksort:

```
QS(A)  $\triangleq$   
  if ( $|A| \leq 1$ ) then return (A);  
   $i := \lfloor |A|/2 \rfloor$ ;  
   $A_{<} := \{a' \in A \mid a' < A[i]\}$ ;  
   $A_{>} := \{a' \in A \mid a' > A[i]\}$ ;  
  return (QS( $A_{<}$ ) ++ A[i] ++ QS( $A_{>}$ ))
```

Worst case complexity:

*$O(n^2)$  comparisons*

## Randomised Quicksort:

```
rQS(A)  $\triangleq$   
  if ( $|A| \leq 1$ ) then return (A);  
   $i := \text{rand}[1 \dots |A|]$ ;  
   $A_{<} := \{a' \in A \mid a' < A[i]\}$ ;  
   $A_{>} := \{a' \in A \mid a' > A[i]\}$ ;  
  return (QS( $A_{<}$ ) ++ A[i] ++ QS( $A_{>}$ ))
```

Worst case complexity:

*$O(n \log(n))$  expected comparisons*

# Run-Time Analysis of Probabilistic Programs is Intricate

- Probabilistic programs may admit infinite runs, but finite expected run-time

```
 $C_{\text{geo}}:$    $n := 0;$   
           repeat  
              $n := n+1; c := \text{coin\_flip}(0.5)$   
           until  $(c=\text{heads})$ 
```

- Positive almost sure termination is not closed under sequential composition:

```
 $C_1:$    $x := 1;$   
       repeat  
          $c := \text{coin\_flip}(0.5); x := 2x;$   
       until  $(c=\text{heads})$ 
```

```
 $C_2:$   repeat  
         $x := x-1;$   
      until  $(x \leq 0)$ 
```

*$C_1$  and  $C_2$  both terminate in finite expected time, while  $C_1;C_2$  does not.*

- (Positive) almost-sure termination is “more undecidable” than ordinary termination

# This talk

- **wp-Calculus** for bounding the expected runtime of probabilistic programs
- **Soundness** of the calculus w.r.t. an operational program semantics
- **Consistency** w.r.t. Nielson's logic for bounding runtime of ordinary programs
- Runtime analysis of **random walk** and the **coupon collector's problem**

# Probabilistic Programming Language

$C ::=$	empty	empty program
	skip	effectless operation
	halt	immediate termination
	$x \approx \mu$	probabilistic assignment
	$C; C$	sequential composition
	$\{C\} \square \{C\}$	non-deterministic choice
	if ( $\eta$ ) $\{C\}$ else $\{C\}$	probabilistic conditional
	while ( $\eta$ ) $\{C\}$	probabilistic while loop

## ■ Truncated geometric distribution:

```
if ( $\frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle$ ) {succ  $\approx$  true} else  
  {if ( $\frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle$ ) {succ  $\approx$  true}  
   else {succ  $\approx$  false}}
```

## ■ Race between tortoise and hare:

```
h  $\approx$  0; t  $\approx$  30;  
while (h  $\leq$  t)  
  t  $\approx$  t + 1;  
  if ( $\frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle$ ) {h  $\approx$  h + Unif[0..10]}  
  else {empty}
```

# The Expected Runtime Transformer — Basics

**Our aim:**

$$\text{program } C \quad \longrightarrow \quad h_C : \overbrace{\mathcal{S} \rightarrow \mathbb{R}_{\geq 0}^{\infty}}^{\mathbb{T}}$$

$h_C(s) \mapsto$  number of skips, assignments and guard evaluations in the execution of  $C$  from state  $s$

**Our approach:**

We use a **continuation passing style** through transformer

$$\text{ert}[C] : \mathbb{T} \rightarrow \mathbb{T}$$

$f \mapsto$  runtime of the computation following program  $C$



$\text{ert}[C](f) \mapsto$  runtime of  $C$ , plus the computation following  $C$

In particular,

$\text{ert}[C](\mathbf{0})(s) \mapsto$  runtime of  $C$ , when started in state  $s$ .



# The Expected Runtime Transformer — Inductive Definition

$$\begin{aligned}\text{ert}[\text{empty}](f) &= f \\ \text{ert}[\text{skip}](f) &= \mathbf{1} + f \\ \text{ert}[\text{halt}](f) &= \mathbf{0} \\ \text{ert}[x : \approx \sum_i p_i \cdot \langle v_i \rangle](f) &= \mathbf{1} + \lambda s \bullet \sum_i p_i \cdot f(s[x \mapsto v_i]) \\ \text{ert}[C_1; C_2](f) &= \text{ert}[C_1](\text{ert}[C_2](f)) \\ \text{ert}[\{C_1\} \square \{C_2\}](f) &= \max\{\text{ert}[C_1](f), \text{ert}[C_2](f)\} \\ \text{ert}[\text{if } (\eta) \{C_1\} \text{ else } \{C_2\}](f) &= \mathbf{1} + \Pr[\eta=\text{true}] \cdot \text{ert}[C_1](f) + \Pr[\eta=\text{false}] \cdot \text{ert}[C_2](f) \\ \text{ert}[\text{while } (\eta) \{C\}](f) &= \text{lfp}(F_f^{\langle \eta, C \rangle}) \quad \text{where} \\ &F_f^{\langle \eta, C \rangle}(X) = \mathbf{1} + \Pr[\eta=\text{false}] \cdot f + \Pr[\eta=\text{true}] \cdot \text{ert}[C](X)\end{aligned}$$

Characteristic functional

# The Expected Runtime Transformer — Elementary Properties

**Monotonicity:**

$$f \preceq g \implies \text{ert}[C](f) \preceq \text{ert}[C](g)$$

**Propagation  
of constants:**

$$\text{ert}[C](\mathbf{k} + f) = \mathbf{k} + \text{ert}[C](f)$$

provided  $C$  is `halt`-free

**Preservation of  $\infty$ :**

$$\text{ert}[C](\infty) = \infty$$

provided  $C$  is `halt`-free

**Sub-additivity:**

$$\text{ert}[C](f + g) \preceq \text{ert}[C](f) + \text{ert}[C](g);$$

$C$  is fully probabilistic

**Scaling:**

$$\begin{aligned} \text{ert}[C](r \cdot f) &\succeq \min\{1, r\} \cdot \text{ert}[C](f) \\ \text{ert}[C](r \cdot f) &\preceq \max\{1, r\} \cdot \text{ert}[C](f) \end{aligned}$$

# The Expected Runtime Transformer — Application Example

$C_{\text{trunc}}$ :    **if**  $(\frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle)$   $\{succ : \approx \text{true}\}$  **else**  
                   $\{\text{if } (\frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle) \{succ : \approx \text{true}\}$   
                                   $\text{else } \{succ : \approx \text{false}\}\}$

$$\begin{aligned} \text{ert}[x : \approx \sum_i p_i \cdot \langle v_i \rangle](f) &= \\ & \mathbf{1} + \lambda s \cdot \sum_i p_i \cdot f(s[x \mapsto v_i]) \\ \text{ert}[\text{if } (\eta) \{C_1\} \text{else } \{C_2\}](f) &= \\ & \mathbf{1} + \text{Pr}[\eta = \text{true}] \cdot \text{ert}[C_1](f) + \text{Pr}[\eta = \text{false}] \cdot \text{ert}[C_2](f) \end{aligned}$$

$$\begin{aligned} \text{ert}[C_{\text{trunc}}](\mathbf{0}) &= \mathbf{1} + \frac{1}{2} \cdot \text{ert}[succ : \approx \text{true}](\mathbf{0}) \\ & \quad + \frac{1}{2} \cdot \text{ert}[\text{if } (\dots) \{succ : \approx \text{true}\} \text{else } \{succ : \approx \text{false}\}](\mathbf{0}) \\ &= \mathbf{1} + \frac{1}{2} \cdot \mathbf{1} + \frac{1}{2} \cdot \left( \mathbf{1} + \frac{1}{2} \cdot \text{ert}[succ : \approx \text{true}](\mathbf{0}) + \frac{1}{2} \cdot \text{ert}[succ : \approx \text{false}](\mathbf{0}) \right) \\ &= \mathbf{1} + \frac{1}{2} \cdot \mathbf{1} + \frac{1}{2} \cdot \left( \mathbf{1} + \frac{1}{2} \cdot \mathbf{1} + \frac{1}{2} \cdot \mathbf{1} \right) = \frac{5}{2} \end{aligned}$$

∴ The execution of  $C_{\text{trunc}}$  takes, on average, 2.5 units of time

# The Expected Runtime of Loops — Proof Rules

$$\frac{F_f^{\langle \eta, C \rangle}(I) \leq I}{\text{ert}[\text{while}(\eta) \{C\}](f) \leq I} \quad [\text{while}]$$

$$F_f^{\langle \eta, C \rangle}(X) = \mathbf{1} + \Pr[\eta = \text{false}] \cdot f + \Pr[\eta = \text{true}] \cdot \text{ert}[C](X)$$

$$\frac{F_f^{\langle \eta, C \rangle}(\mathbf{0}) \geq I_0 \quad F_f^{\langle \eta, C \rangle}(I_n) \geq I_{n+1}}{\text{ert}[\text{while}(\eta) \{C\}](f) \geq \lim_{n \rightarrow \infty} I_n} \quad [\omega\text{-while}^{\geq}]$$

$$\frac{F_f^{\langle \eta, C \rangle}(\mathbf{0}) \leq I_0 \quad F_f^{\langle \eta, C \rangle}(I_n) \leq I_{n+1}}{\text{ert}[\text{while}(\eta) \{C\}](f) \leq \lim_{n \rightarrow \infty} I_n} \quad [\omega\text{-while}^{\leq}]$$

## Theorem

The above proof rules are **sound** and **complete**.

# The Expected Runtime of Loops — Example

$$C_{\text{geo}^*} : \text{while } (b = 1) \{ b := \frac{1}{2} \cdot \langle 0 \rangle + \frac{1}{2} \cdot \langle 1 \rangle \}$$

$$\frac{1 + \Pr[\eta = \text{false}] \cdot f + \Pr[\eta = \text{true}] \cdot \text{ert}[C](I) \leq I}{\text{ert}[\text{while } (\eta) \{C\}](f) \leq I} \quad [\text{while}]$$

$$\text{ert}[x := \mu](f) = 1 + \lambda s \cdot E_{\mu}(\lambda v. f(s[x/v]))$$

To upper-bound the runtime of  $C_{\text{geo}^*}$  we apply rule [while] with continuation  $f = 0$  and invariant  $I = 1 + [b = 1] \cdot 4$

$$\begin{aligned} & 1 + [b \neq 1] \cdot 0 + [b = 1] \cdot \text{ert}[b := \frac{1}{2} \cdot \langle 0 \rangle + \frac{1}{2} \cdot \langle 1 \rangle](I) \\ &= 1 + [b = 1] \cdot \left( 1 + \frac{1}{2} \cdot I[b/0] + \frac{1}{2} \cdot I[b/1] \right) \\ &= 1 + [b = 1] \cdot \left( 1 + \frac{1}{2} \cdot \underbrace{(1 + [0 = 1] \cdot 4)}_{=1} + \frac{1}{2} \cdot \underbrace{(1 + [1 = 1] \cdot 4)}_{=5} \right) \\ &= 1 + [b = 1] \cdot 4 = I \leq I \end{aligned}$$

and conclude that  $\text{ert}[C_{\text{geo}^*}](0) \leq 1 + [b=1] \cdot 4$

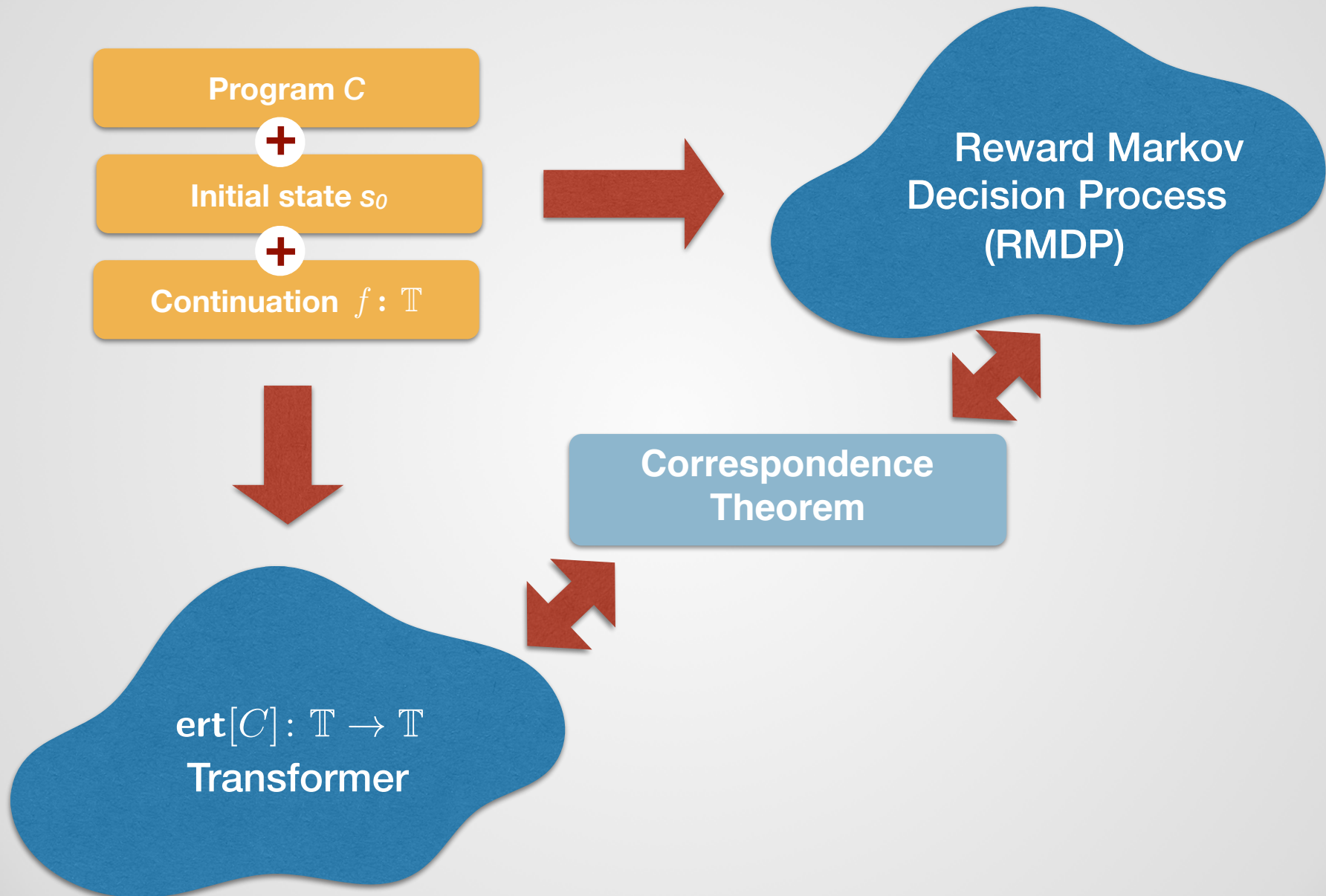
- The expected runtime of  $C_{\text{geo}^*}$  is at most 5 from any initial state where  $b=1$
- and at most 1 from all other states.

# The Expected Runtime of Loops — Bound Refinement

$$\frac{\text{ert}[\text{while}(\eta)\{C\}](f) \leq g \quad F_f^{\langle\eta, C\rangle}(g) \leq g}{\text{ert}[\text{while}(\eta)\{C\}](f) \leq F_f^{\langle\eta, C\rangle}(g) \leq g}$$

$$\frac{\text{ert}[\text{while}(\eta)\{C\}](f) \geq g \quad F_f^{\langle\eta, C\rangle}(g) \geq g}{\text{ert}[\text{while}(\eta)\{C\}](f) \geq F_f^{\langle\eta, C\rangle}(g) \geq g}$$

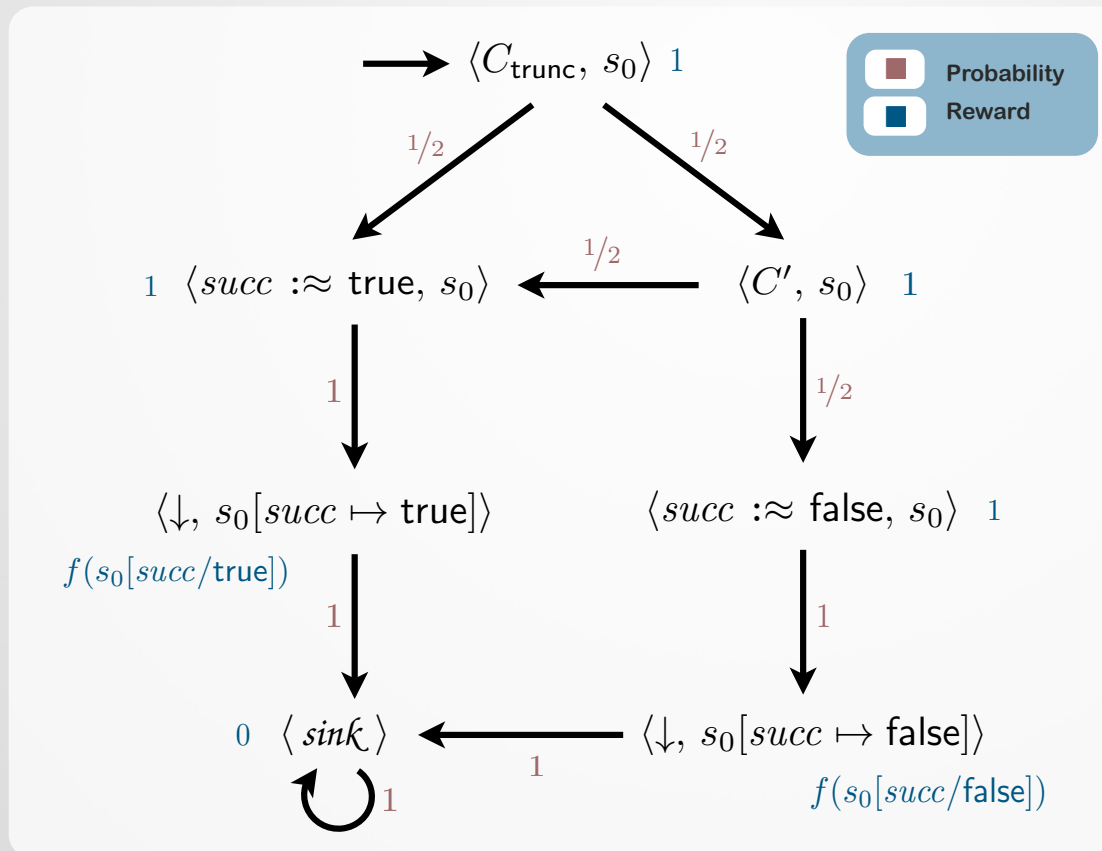
# Operational Program Semantics — The Big Picture



# Operational Program Semantics

$C_{\text{trunc}} :$      $\text{if } \left( \frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle \right) \{succ \approx \text{true}\} \text{ else}$   
 $\underbrace{\left\{ \text{if } \left( \frac{1}{2} \cdot \langle \text{true} \rangle + \frac{1}{2} \cdot \langle \text{false} \rangle \right) \{succ \approx \text{true}\} \text{ else } \{succ \approx \text{false}\} \right\}}_{C'}$

RMDP for program  $C_{\text{trunc}}$ , initial state  $s_0$  and continuation  $f$





# Operational Program Semantics — Reward MDP Construction

## RMDP State

## Interpretation

## RMDP Reward

$\langle C, s \rangle, \langle \downarrow; C, s \rangle$

intermediate execution point  
(C remaining computation from  
intermediate state s)

0 or 1

$\langle \downarrow, s \rangle$

normal termination  
(s final program state)

$f(s)$

$\langle \text{sink} \rangle$

termination  
(normal or halt)

0

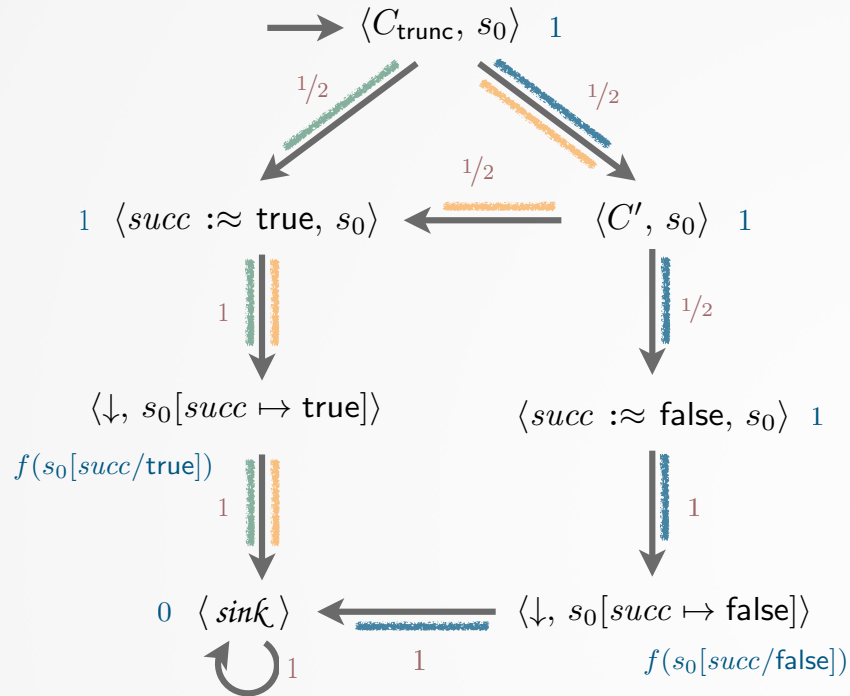
## Sample Construction Rules

$$\frac{\Pr[\mu(s)=v] = p > 0}{\langle x : \approx \mu, s \rangle \xrightarrow{\tau} \langle \downarrow, s[x/v] \rangle \vdash p} \text{ [pr-assgn]}$$

$$\frac{\Pr[\eta(s)=\text{true}] = p > 0}{\langle \text{if } (\eta) \{ C_1 \} \text{ else } \{ C_2 \}, s \rangle \xrightarrow{\tau} \langle C_1, s \rangle \vdash p} \text{ [if-true]}$$

$$\frac{}{\langle \text{while } (\eta) \{ C \}, s \rangle \xrightarrow{\tau} \langle \text{if } (\eta) \{ C; \text{while } (\eta) \{ C \} \} \text{ else } \{ \text{empty} \}, s \rangle \vdash 1} \text{ [while]}$$

# Operational Program Semantics — Relation to the $\text{ert}[\cdot]$ Transformer



$$\begin{aligned}
 \text{ExpRew}^{\mathcal{M}_s^f[C]}(\langle \text{sink} \rangle) &= \text{Pr}[\pi_{\text{true}}] \cdot \text{rew}(\pi_{\text{true}}) \\
 &+ \text{Pr}[\pi_{\text{false true}}] \cdot \text{rew}(\pi_{\text{false true}}) \\
 &+ \text{Pr}[\pi_{\text{false false}}] \cdot \text{rew}(\pi_{\text{false false}}) \\
 &= \left(\frac{1}{2} \cdot 1 \cdot 1\right) \cdot (1 + 1 + f(s_0[\text{succ}/\text{true}])) \\
 &+ \left(\frac{1}{2} \cdot \frac{1}{2} \cdot 1 \cdot 1\right) \cdot (1 + 1 + 1 + f(s_0[\text{succ}/\text{true}])) \\
 &+ \left(\frac{1}{2} \cdot \frac{1}{2} \cdot 1 \cdot 1\right) \cdot (1 + 1 + 1 + f(s_0[\text{succ}/\text{false}])) \\
 &= \frac{5}{2} + \frac{3}{4} \cdot f(s_0[\text{succ}/\text{true}]) + \frac{1}{4} \cdot f(s_0[\text{succ}/\text{false}]).
 \end{aligned}$$

$$\therefore \text{ExpRew}^{\mathcal{M}_{s_0}^0[C]}(\langle \text{sink} \rangle) = \frac{5}{2} = \text{ert}[C](\mathbf{0})(s_0)$$

## Theorem (Soundness)

Let  $\text{ExpRew}^{\mathcal{M}_{s_0}^f[C]}(\langle \text{sink} \rangle)$  be the expected reward to reach the sink in the RMDP associated to program  $C$ , initial state  $s_0$  and continuation  $f$ . Then

$$\text{ert}[C](f)(s_0) = \text{ExpRew}^{\mathcal{M}_{s_0}^f[C]}(\langle \text{sink} \rangle).$$

# Nielson's Logic for Deterministic Programs — Basics

## Judgments

(Total Correctness)

(Runtime Bound)

$$\{P\} C \{E \Downarrow Q\} \triangleq \{P\} C \{\Downarrow Q\} + C \text{ terminates from } s \text{ in (at most a mult. of) } \llbracket E \rrbracket(s) \text{ steps if } s \models P$$

deterministic  
program

numeric expression  
over program variables

Eg.  $\{\text{true}\} \text{ while } (x \geq 0) \{x := x - 1\} \{x \Downarrow x < 0\}$

## Sample proof rules

$$\frac{}{\vdash_E \{Q[x/e]\} x := e \{1 \Downarrow Q\}} [\text{Assgn}]$$

$$\frac{\vdash_E \{P \wedge B\} C_1 \{E \Downarrow Q\} \quad \vdash_E \{P \wedge \neg B\} C_2 \{E \Downarrow Q\}}{\vdash_E \{P\} \text{ if } (B) \{C_1\} \text{ else } \{C_2\} \{E \Downarrow Q\}} [\text{if}]$$

## Theorem $\text{ert}[\cdot]$ generalises Nielson's logic to probabilistic programs

For any deterministic program  $C$ ,

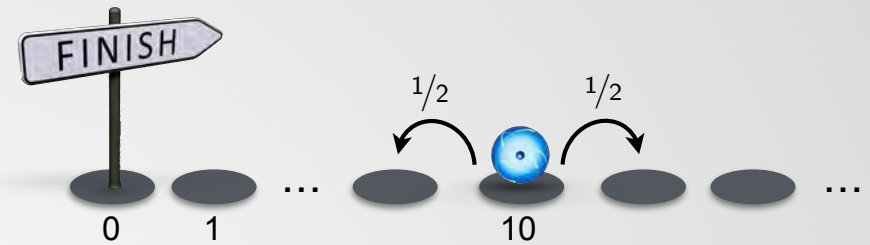
$$\vdash \{P\} C \{\Downarrow Q\} \implies \vdash_E \{P\} C \{\text{ert}[C](\mathbf{0}) \Downarrow Q\} \quad (\text{soundness})$$

$$\vdash_E \{P\} C \{E \Downarrow Q\} \implies \exists k \bullet \text{ert}[C](\mathbf{0})(s) = k \cdot \llbracket E \rrbracket(s) \quad (\text{completeness})$$

# Runtime Analysis of a Random Walk

A particle starts at position  $x=10$  and moves with equal probability to the left or to the right in each turn. The random walk stops when the particle reaches position  $x=0$ .

**What is the expected number of moves to termination?**



$C_{rw}: \quad x := 10; \text{ while } (x > 0) \{ x := \frac{1}{2} \cdot \langle x-1 \rangle + \frac{1}{2} \cdot \langle x+1 \rangle \}$

It can be shown that  $C_{rw}$  terminates with probability one. Using our expected runtime calculus one can show that it takes an expected infinite time to do so:

$$\text{ert}[C_{rw}](0) = \infty$$

# Coupon Collector's Problem



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## Coupon collector

From Wikipedia

### ON A CLASSICAL PROBLEM OF PROBABILITY THEORY

by  
P. ERDŐS and A. RÉNYI

We consider the following classical "urn-problem". Suppose that there are  $n$  urns given, and that balls are placed at random in these urns one after the other. Let us suppose that the urns are labelled with the numbers  $1, 2, \dots, n$  and let  $\xi_j$  be equal to  $k$  if the  $j$ -th ball is placed into the  $k$ -th urn. We suppose that the random variables  $\xi_1, \xi_2, \dots, \xi_N, \dots$  are independent, and  $P(\xi_j = k) = \frac{1}{n}$  for  $j = 1, 2, \dots$  and  $k = 1, 2, \dots, n$ . By other words each ball may be placed in any of the urns with the same probability and the choices of the urns for the different balls are independent. We continue this process so long till there are at least  $m$  balls in every urn ( $m = 1, 2, \dots$ ). What can be said about the number of balls which are needed to achieve this goal?

We denote the number in question (which is of course a random variable) by  $v_m(n)$ . The "dixie cup"-problem considered in [1] is clearly equivalent with the above problem. In [1] the mean value  $M(v_m(n))$  of  $v_m(n)$  has been evaluated (here and in what follows  $M(\cdot)$  denotes the mean value of the random variable in the brackets) and it has been shown that

$$M(v_m(n)) = n \log n + (m-1) n \log \log n + n \cdot C_m + o(n)$$

(1)  $C_m$  is a constant, depending on  $m$ . (The value of  $C_m$  is not given in [1]). To estimate we shall go a step further and determine asymptotically the distribution of  $v_m(n)$ ; we shall prove that for every

$$P(v_m(n) \leq x) = \frac{e^{-x}}{(x-1)!}.$$

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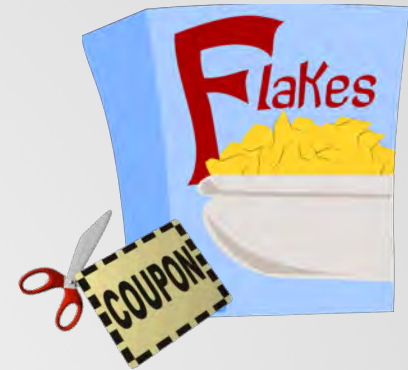


250  $E(n)$



# Coupon Collector's Problem

Suppose each box of cereal contains one of  $N$  different coupons and once a consumer has collected a coupon of each type, he can trade them for a prize. The aim of the problem is **determining the average number of cereal boxes the consumer should buy to collect all coupon types**, assuming that each coupon type occurs with the same probability in the cereal boxes.



Using our expected runtime calculus we showed that the expected number of necessary cereal boxes is in  $\mathcal{O}(N \log(N))$ .

```
 $C_{\text{ccp}}:$    $cp := [0, \dots, 0]; i := 1; x := N;$   
            while  $(x > 0)$  {  
                while  $(cp[i] \neq 0)$  {  $i := \text{Unif}[1 \dots N]$  };  
                 $cp[i] := 1; x := x - 1$   
            }
```

$$\text{ert}[C_{\text{ccp}}](\mathbf{0}) = 4 + 2N \cdot (2 + \mathcal{H}_{N-1}) \in \mathcal{O}(N \log(N))$$

# Summary

- Reasoning about the expected runtime of probabilistic programs a la Dijkstra
  - Handles **finite and infinite runtimes**
  - Establishes both **bounds** and **exact values** of the program runtimes
  - Includes several **sound and complete proof rules** for reasoning about **loops**
  - Extension with **recursion** has recently been provided

■ Soundness of the technique w.r.t. an operational program semantics

■ Extends Hoare logic for bounding the runtime of deterministic programs

Certified in Isabelle (courtesy Johannes Hölzl)

■ Cases: **random walk, coupon collector's problem, randomised binary search**

Further details: see ESOP'16 paper; for recursion see LICS'16