

# Probabilistic Programming

Lecture #16+#17: Expected Runtime Analysis

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RWTH Lecture Series on Probabilistic Programming 2018

# Overview

- 1 Motivation
- 2 An unsound approach
- 3 The expected runtime transformer
- 4 Properties
- 5 Proof rules for runtimes of loops
- 6 Proving positive almost-sure termination
- 7 Case studies

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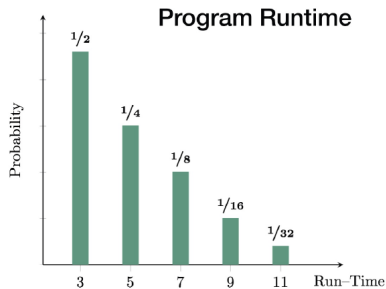
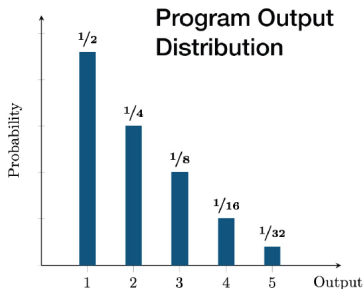
# The runtime of a probabilistic program

The **runtime** of a probabilistic program depends  
on the **input** and  
on the internal **randomness** of the program.



# The runtime of a probabilistic program is random

```
int i := 0;  
repeat {i++; (c := false [0.5] c := true)}  
until (c)
```



The **expected runtime** is  $1 + 3 \cdot 1/2 + 6 \cdot 1/4 + \dots (3n+1) \cdot 1/2^n = 5$ .

# Expected runtimes

Expected run-time of program  $P$  on input  $s$ :

$$\sum_{i=1}^{\infty} i \cdot Pr \left( \begin{array}{l} \text{"} P \text{ terminates after} \\ i \text{ steps on input } s \text{"} \end{array} \right)$$

# Efficiency of randomised algorithms

## Quicksort:

```
QS(A) =  
  if |A| <= 1 { return A; }  
  i := ceil(|A|/2);  
  A< := {a in A | a < A[i]};  
  A> := {a in A | a > A[i]};  
  return QS(A<) ++ A[i] ++ QS(A>)
```

Worst case complexity:

*$O(N^2)$  comparisons*



## Randomised Quicksort:

```
rQS(A) =  
  if |A| <= 1 { return A; }  
  i := Unif[1...|A|];  
  A< := {a in A | a < A[i]};  
  A> := {a in A | a > A[i]};  
  return rQS(A<) ++ A[i] ++ rQS(A>)
```

Worst case complexity:

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



# Coupon collector's problem

## ON A CLASSICAL PROBLEM OF PROBABILITY THEORY

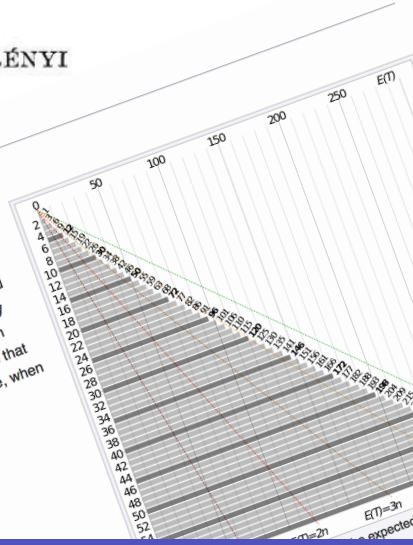
by

P. ERDŐS and A. RÉNYI

### Coupon collector's problem

From Wikipedia, the free encyclopedia

In [probability theory](#), the **coupon collector's problem** describes the "collect all coupons and win" contests. It asks the following question: Suppose that there is an [urn](#) of  $n$  different [coupons](#), from which coupons are being collected, equally likely, with replacement. What is the probability that more than  $t$  sample trials are needed to collect all  $n$  coupons? An alternative statement is: Given  $n$  coupons, how many coupons do you expect you need to draw with replacement before having drawn each coupon at least once? The mathematical analysis of the problem reveals that the [expected number](#) of trials needed grows as  $\Theta(n \log(n))$ .<sup>[1]</sup> For example, when about 225<sup>[2]</sup> trials to collect all 50 coupons.



# Coupon collector's problem

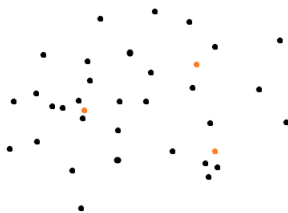
---

```
cp := [0,...,0]; // no coupons yet
i := 1; // coupon to be collected next
x := 0; // number of coupons collected
while (x < N) {
    while (cp[i] != 0) {
        i := uniform(1..N) // next coupon
    }
    cp[i] := 1; // coupon i obtained
    x++; // one coupon less to go
}
```

---

The expected runtime of this program is in  $\Theta(N \cdot \log N)$ .

# Closest-pair problem



**Closest-pair problem:** find two distinct points  $u, v \in \mathbb{R}^2$  among  $N$  points in the plane that minimise the Euclidean distance among all pairs of these points.

A naive deterministic approach takes  $O(N^2)$ . More efficient version in  $O(N \cdot \log N)$ .

Rabin's randomised algorithm has an **expected runtime in  $O(N)$** .

# Randomised primality test

Problem: is  $N$  prime or not?

Basic structure of a **randomised** primality test:

1. Randomly pick a number  $a$ , say
2. Do the **primality test**: Check some equality involving  $a$  and  $N$
3. If equality fails,  $N$  is composite (with witness  $a$ )
4. Otherwise repeat the process.

If after  $K > 0$  iterations,  $N$  is not found to be composite, then  $N$  is **probably prime**.

# Some primality tests

- ▶ **Fermat primality test:**

Select  $a \in \mathbb{Z}$  relative prime to  $N$ . If  $a^{N-1} \bmod N \neq 1$ , then  $N$  is composite.

- ▶ **Rabin-Miller test:**

Select  $0 < a < N$ . Let  $2^s \cdot d = N-1$  where  $d$  is odd. If  $a^d \not\equiv 1 \pmod{N}$  and  $a^{2^r \cdot d} \not\equiv -1 \pmod{N}$  for all  $0 \leq r \leq s-1$ , then  $N$  is composite.

- ▶ **Solovay and Strassen test:**

For  $N$  odd, pick  $a < N$ . If  $a^{N-1/2} \not\equiv \dots$ , then  $N$  is composite.

Adleman and Huang (1992) provided a randomised primality test that terminates with **expected polynomial runtime** and certainly provides the correct answer.<sup>1</sup>

---

<sup>1</sup>Decision problems with this characteristic constitute the complexity class ZPP (zero-error probabilistic polynomial time).



# The aim of this lecture

A wp-calculus to reason about runtimes at **the source code level**.

No “descend” into the underlying probabilistic model.

The calculus should be **compositional**.

$$\text{ert}(P ; Q) = \text{ert}(P) \overset{?}{\times} \text{ert}(Q)$$

# Proving **positive** almost-sure termination

- ▶ What? AST+termination in finite expected time

# Proving **positive** almost-sure termination

- ▶ What? AST+termination in finite expected time
- ▶ Generalise. How?
  - ▶ Provide an weakest-precondition calculus
  - ▶ ..... for expected runtimes
- ▶ Why?
  - ▶ Reason about the efficiency of randomised algorithms
  - ▶ Reason about simulation efficiency of Bayesian networks
  - ▶ Is compositional and reasons at the program's code

# Hurdles in runtime analysis

1. Programs may admit **diverging runs** while still having a **finite expected runtime**

```
while (x > 0) { x-- [1/2] skip }
```

admits a diverging run but has expected runtime  $O(x)$ .

2. Having a finite expected time is **not compositional** w.r.t. sequencing
3. Expected runtimes are **extremely sensitive** to variations in probabilities

```
while (x > 0) { x-- [1/2+p] x++ } // 0 <= p <= 1/2
```

- ▶ For  $p=0$ , the expected runtime is infinite.
- ▶ For arbitrary small  $p > 0$ , the expected runtime is  $1/2 \cdot p \cdot x$ , linear in  $x$ .

P::

$x := 1$

$c := 1;$

while (c) {

$c := 0 \quad [\frac{1}{2}] \quad c := 1;$

$x := 2 * x$

}

$$\sum_{i=1}^{\infty} \underbrace{\frac{1}{2^i} \cdot 2^i}_{\text{exp. value of } x \text{ after } i \text{ iteration}} = \infty$$

exp. value  
of  $x$  after  
 $i$  iteration

PAST(P)

Q::

while ( $x > 0$ ) {  $x--$  }

PAST(Q)

$\neg \text{PAST}(P; Q)$

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# Re-use weakest preconditions?

Idea: equip the program with a counter  $rc$  and use standard wp-reasoning to determine its expected value.

Determine  $wp(\underset{rc}{P}, rc)$  for program  $P$ .

$= P_{rc}$



Dexter Kozen

A probabilistic PDL

1983

# An example

Consider the program  $P$ :

---

```
x := 1;  
while (x > 0) { x := 0 [1/2] skip }
```

---



# An example

Consider the program  $P$ :

---

```
x := 1;  
while (x > 0) { x := 0 [1/2] skip }
```

---

Equipping  $P$  with a runtime counter yields  $P_{rc}$ :

---

```
x := 1; rc := 1;  
while (x > 0) { rc++; (x := 0 [1/2] skip) }
```

---

# An example

Consider the program  $P$ :

---

```
x := 1;
while (x > 0) { x := 0 [1/2] skip }
```

---

Equipping  $P$  with a runtime counter yields  $P_{rc}$ :

---

```
x := 1; rc := 0;
while (x > 0) { rc++; (x := 0 [1/2] skip) }
```

---

now  
next  
replace  
this  
fragment



It follows  $\Phi(I) \leq I$  for  $I = rc + [x > 0] \cdot 2$ .

In total, we thus obtain  $wp(P_{rc}, rc) = 2$ .

# An example

Consider the program  $Q$ :

---

```
x := 1;  
while (x > 0) { x := 0 [1/2] while(true) { skip } }
```

---

*diverge*

# An example

Consider the program  $Q$ :

---

```
x := 1;
while (x > 0) { x := 0 [1/2] while(true) { skip } }
```

---

Equipping  $Q$  with a runtime counter yields  $Q_{rc}$ :

---

```
x := 1; rc := 0;
while (x > 0) {
  rc++;
  (x := 0 [1/2] while(true) { rc++ ; skip})
}
```

---

# An example

Consider the program  $Q$ :

---

```
x := 1;
while (x > 0) { x := 0 [1/2] while(true) { skip } }
```

---

Equipping  $Q$  with a runtime counter yields  $Q_{rc}$ :

---

```
x := 1; rc := 0;
while (x > 0) {
  rc++;
  (x := 0 [1/2] while(true) { rc++ ; skip })
}
```

---

As  $wp(\text{inner loop}, f) = 0$  for every  $f$ , it follows  $\Phi_{Q_{rc}} \leq \Phi_{P_{rc}}$ .

↓  
 "skip" variant  
 of  $Q_{rc}$  (prev.  
 slide)

# An example

Consider the program  $Q$ :

---

```
x := 1;
while (x > 0) { x := 0 [1/2] while(true) { skip } }
```

---

Equipping  $Q$  with a runtime counter yields  $Q_{rc}$ :

---

```
x := 1; rc := 0;
while (x > 0) {
  rc++;
  (x := 0 [1/2] while(true) { rc++ ; skip })
}
```

---

As  $wp(\text{inner loop}, f) = 0$  for every  $f$ , it follows  $\Phi_{Q_{rc}} \leq \Phi_{P_{rc}}$ .

Thus,  $\Phi_{Q_{rc}}(l) \leq \Phi_{P_{rc}}(l) \leq l$  for  $l = rc + [x > 0] \cdot 2$ .

# An example

Consider the program  $Q$ :

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x := 1;
while (x > 0) { x := 0 [1/2] while(true) { skip } }
```

---

Equipping  $Q$  with a runtime counter yields  $Q_{rc}$ :

---

```
x := 1; rc := 0;
while (x > 0) {
  rc++;
  (x := 0 [1/2] while(true) { rc++ ; skip})
}
```

---

As  $wp(\text{inner loop}, f) = 0$  for every  $f$ , it follows  $\Phi_{Q_{rc}} \leq \Phi_{P_{rc}}$ .

Thus,  $\Phi_{Q_{rc}}(l) \leq \Phi_{P_{rc}}(l) \leq l$  for  $l = rc + [x > 0] \cdot 2$ .

This contradicts the fact that the true expected runtime of  $Q$  is  $\infty$ .

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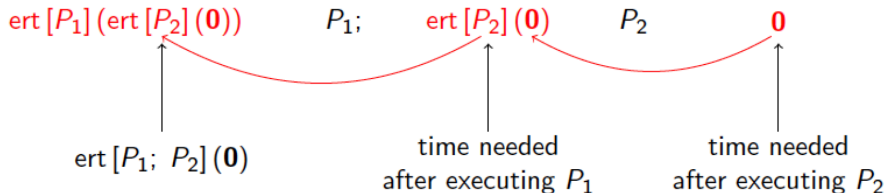


# The basic idea

function:  $S \rightarrow \mathbb{R}_{\geq 0} + \infty$

Let  $\text{ert}() : \text{pGCL} \rightarrow (\mathbb{T} \rightarrow \mathbb{T})$  where:

- ▶  $\text{ert}(P, t)(s)$  is the expected runtime of  $P$  on input state  $s$  if  $t$  captures the runtime of the computation following  $P$ .
- ▶  $\text{ert}(P, 0)(s)$  is the expected runtime of  $P$  on input state  $s$ .



# Runtimes

## Expectations

A **expectation**  $f : \mathbb{S} \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$ .

Let  $\mathbb{E}$  be the set of all expectations and let  $\sqsubseteq$  be defined for  $f, g \in \mathbb{E}$  by:

$$f \sqsubseteq g \quad \text{if and only if} \quad f(s) \leq g(s) \quad \text{for all } s \in \mathbb{S}.$$

## Runtimes

A **runtime**  $t : \mathbb{S} \rightarrow \mathbb{R}_{\geq 0} \cup \{\infty\}$ .

Let  $\mathbb{T}$  denote the set of all runtimes and let  $\leq$  be defined for  $t, u \in \mathbb{T}$  by:

$$t \leq u \quad \text{if and only if} \quad t(s) \leq u(s) \quad \text{for all } s \in \mathbb{S}.$$

A runtime **transformer** is defined in a similar way as an expectation transformer

# The runtime model

We assume the following runtimes:

- ▶ Executing a `skip`-statement takes a single time unit
- ▶ Executing an (ordinary or random) assignment takes a single time unit
- ▶ Evaluating a guard takes a single time unit
- ▶ Flipping a coin in a probabilistic choice takes a single time unit
- ▶ Sequential composition does not take time

The ert-calculus can be easily adapted to other runtime models.

# Expected runtime transformer for pGCL

$P$

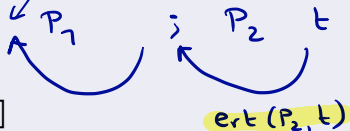
$ert(P, \bullet)$

## Syntax

- ▶ skip
- ▶ diverge
- ▶  $x := E$
- ▶  $x :r= \mu$
- ▶  $P_1 ; P_2$
- ▶ if (G)  $P_1$  else  $P_2$
- ▶  $P_1 [p] P_2$
- ▶ while(G)  $P$

## Expected runtime $ert(P, t)$

- ▶  $1 + t$
- ▶  $\infty$
- ▶  $1 + t[x := E]$
- ▶  $1 + \lambda s. \int_{\mathbb{Q}} (\lambda v. t[s[x := v]]) d\mu_s$
- ▶  $ert(P_1, ert(P_2, t))$
- ▶  $1 + [G] \cdot ert(P_1, t) + [\neg G] \cdot ert(P_2, t)$
- ▶  $1 + p \cdot ert(P_1, t) + (1-p) \cdot ert(P_2, t)$
- ▶  $\text{lfp } X. (1 + [G] \cdot ert(P, X) + [\neg G] \cdot t)$



# Expected runtime transformer for pGCL

## Syntax

- ▶ skip
- ▶ diverge
- ▶  $x := E$
- ▶  $x \text{ :r= } \mu$
- ▶  $P_1 ; P_2$
- ▶ if (G)  $P_1$  else  $P_2$
- ▶  $P_1 [p] P_2$
- ▶ while(G)  $P$

## Expected runtime $ert(P, t)$

- ▶  $1 + t$
- ▶  $\infty$
- ▶  $1 + t[x := E]$
- ▶  $1 + \lambda s. \int_{\mathbb{Q}} (\lambda v. t(s[x := v])) d\mu_s$
- ▶  $ert(P_1, ert(P_2, t))$
- ▶  $1 + [G] \cdot ert(P_1, t) + [\neg G] \cdot ert(P_2, t)$
- ▶  $1 + p \cdot ert(P_1, t) + (1 - p) \cdot ert(P_2, t)$
- ▶  $\text{lfp } X. (1 + [G] \cdot ert(P, X) + [\neg G] \cdot t)$

lfp is the least fixed point operator wrt. the ordering  $\leq$  on runtimes

**Examples**  $P:: \text{succ} := 1 \left[ \frac{1}{2} \right] (\text{succ} := 1 \left[ \frac{1}{2} \right] \text{succ} := 0)$

$$\begin{aligned}
 \text{ert}(P, \underline{0}) &= 1 + \frac{1}{2} \text{ert}(\text{succ} := 1, \underline{0}) \\
 &\quad + \frac{1}{2} \text{ert}(\text{succ} := 1 \left[ \frac{1}{2} \right] \text{succ} := 0, \underline{0}) \\
 &= 1 + \frac{1}{2} \left( 1 + \underbrace{\underline{0} [\text{succ} := 1]}_{= \underline{0}} \right) \\
 &\quad \underbrace{\hspace{10em}}_{= 1/2} \\
 &\quad + \frac{1}{2} \left( 1 + \underbrace{\frac{1}{2} \text{ert}(\text{succ} := 1, \underline{0})}_{\frac{1}{2}(1+\underline{0})} + \underbrace{\frac{1}{2} \text{ert}(\text{succ} := 0, \underline{0})}_{\frac{1}{2}(1+\underline{0})} \right) \\
 &= 1 + \frac{1}{2} + \frac{1}{2} \left( 1 + \frac{1}{2} + \frac{1}{2} \right) \\
 &= \frac{5}{2}
 \end{aligned}$$

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# Elementary properties

► **Continuity:**  $ert(P, t)$  is continuous on  $(\mathbb{T}, \leq)$

► **Monotonicity:**  $t \leq t'$  implies  $ert(P, t) \leq ert(P, t')$

► **Constant propagation:**  $ert(P, \mathbf{k} + t) = \mathbf{k} + ert(P, t)$

► **Preservation of  $\infty$ :**  $ert(P, \infty) = \infty$

► **Connection to wp:**  $ert(P, t) = ert(P, \mathbf{0}) + wp(P, t)$

► **Affinity:**  $ert(P, a \cdot t + t') = ert(P, \mathbf{0}) + a \cdot ert(P, t) + ert(P, t')$



$$\text{ert}(P, t+t') = \text{ert}(P, t) + \text{wp}(P, t')$$

Proof (sketch) by induction on the structure of  $P$ .

$$\begin{aligned} \text{ert}(P; Q, t+t') &= (* \text{ def. of } \text{ert} *) \\ &\quad \text{ert}(P, \text{ert}(Q, t+t')) \\ &= (* \text{ I.H. on } Q *) \end{aligned}$$

$$\begin{aligned} &\text{ert}(P, \underbrace{\text{ert}(Q, t) + \text{wp}(Q, t')}) \\ &= (* \text{ I.H. on } P *) \end{aligned}$$

$$\begin{aligned} &\underbrace{\text{ert}(P, \text{ert}(Q, t))} + \underbrace{\text{wp}(P, \text{wp}(Q, t'))} \\ &\quad \text{ert}(P; Q, t) + \text{wp}(P; Q, t') \end{aligned}$$

loop:  $\Phi_t = 1 + [\neg G] \cdot t + [G] \text{ert}(P, X)$   
while (G) {P}

$$\Phi'_{t'} = [\neg G] \cdot t' + [G] \text{wp}(P, X)$$

To show  $\text{ert}(\text{loop}, t)$

$$\begin{aligned} \underbrace{\text{lfp } X. \Phi_{t+t'}(X)}_{\text{ert}(\text{loop}, t+t')} &= \underbrace{\text{lfp } X. \Phi_t(X)}_{\text{ert}(\text{loop}, t)} \\ &\quad + \underbrace{\text{lfp } X. \Phi'_{t'}(X)}_{\text{wp}(\text{loop}, t')} \end{aligned}$$

This is equivalent to prove:

$$\underbrace{\lim_{n \rightarrow \infty} \Phi_{t+t'}^n(\underline{0})}_{\text{ert}(\text{loop}, t+t')} = \underbrace{\lim_{n \rightarrow \infty} \Phi_t^n(\underline{0}) + \Phi_{t'}^{'n}(\underline{0})}_{\text{ert}(\text{loop}, t) + \text{wp}(\text{loop}, t')}$$

Proof:  $\forall n: \Phi_{t+t'}^n(\underline{0}) = \Phi_t^n(\underline{0}) + \Phi_{t'}^{'n}(\underline{0}) \quad (*)$

by induction on  $n$ . Base case  $n=0: \underline{0} = \underline{0} + \underline{0}$

Ind. step.

$$\begin{aligned} \Phi_{t+t'}^{n+1}(\underline{0}) &= 1 + [\neg G] \cdot (t+t') + [G] \text{ert}(P, \Phi_{t+t'}^n(\underline{0})) \\ &= (* \text{ I.H. on } n *) \end{aligned}$$

$$\begin{aligned} &1 + [\neg G](t+t') + [G] \cdot \underbrace{\text{ert}(P, \Phi_t^n(\underline{0}) + \Phi_{t'}^{'n}(\underline{0}))}_{\text{ert}(\text{loop}, t) + \text{wp}(\text{loop}, t')} \\ &= (* \text{ I.H. on loop body } P *) \end{aligned}$$

$$1 + [\neg G](t+t') + [G] \left( \underbrace{\text{ert}(P, \Phi_t^n(\underline{0}))}_{\text{ert}(\text{loop}, t)} + \underbrace{\text{wp}(P, \Phi_{t'}^{'n}(\underline{0}))}_{\text{wp}(\text{loop}, t')} \right)$$

$$\begin{aligned} &= 1 + [\neg G](t) + [G] \cdot \text{ert}(P, \Phi_t^n(\underline{0})) \\ &\quad + [\neg G](t') + [G] \cdot \text{wp}(P, \Phi_{t'}^{'n}(\underline{0})) \end{aligned}$$



# (Positive) almost-sure termination

For every pGCL program  $P$  and input state  $s$ :

$$\underbrace{ert(P, 0)(s) < \infty}_{\text{positive a.s-termination on } s} \quad \text{implies} \quad \underbrace{wp(P, 1)(s) = 1}_{\text{almost-sure termination on } s}$$

Moreover:

$$\underbrace{ert(P, 0) \leq \infty}_{\text{universal positive a.s-termination}} \quad \text{implies} \quad \underbrace{wp(P, 1) = 1}_{\text{universal almost-sure termination}}$$

# A Markov chain perspective on runtimes

- ▶ Consider  $\text{ert}(P, t)$  for pCGL program  $P$
- ▶ Consider the Markov chain  $\llbracket P \rrbracket$  of program  $P$ 
  - keep a track of the expected runtime of a program statement
- ▶ Attach rewards to each Markov chain state in  $\llbracket P \rrbracket$ :
  - ▶ State  $\langle \downarrow, s \rangle$  gets reward  $t(s)$
  - ▶ State  $\langle \text{skip}, s \rangle$  gets reward one
  - ▶ State  $\langle \text{diverge}, s \rangle$  gets reward  $\infty$
  - ▶ State  $\langle x := E, s \rangle$  gets reward one
  - ▶ State  $\langle x \approx \mu, s \rangle$  gets reward one
  - ▶ State  $\langle \text{if } G \dots, s \rangle$  gets reward one
  - ▶ State  $\langle P[p]Q, s \rangle$  gets reward one
  - ▶ State  $\langle \text{while}(G)P' \dots, s \rangle$  gets reward one
  - ▶ All other states get reward zero

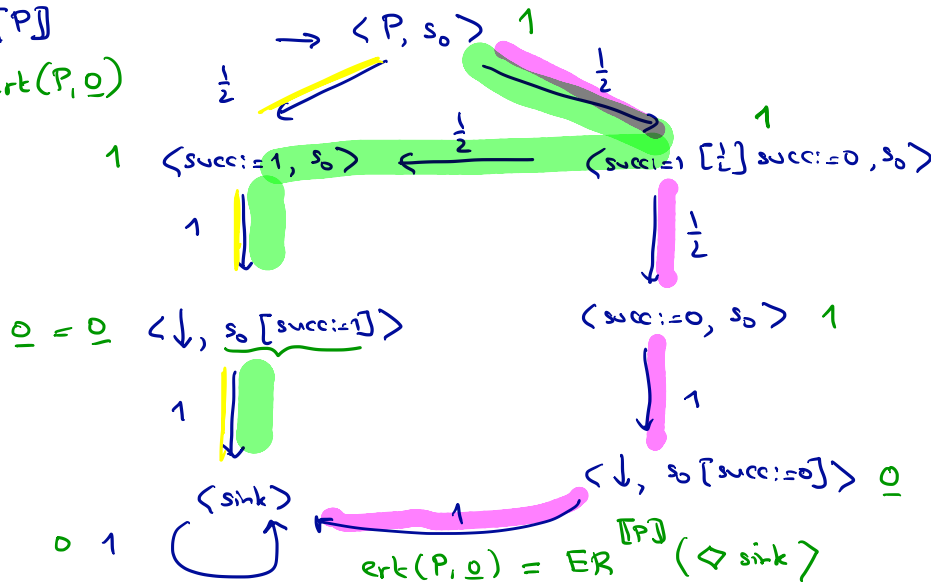
$$r: \Sigma \rightarrow \mathbb{R}_{\geq 0} + \infty$$

# Example

$$\text{succ} := 1 \quad \left[ \frac{1}{2} \right] \left( \text{succ} := 1 \quad \left[ \frac{1}{2} \right] \text{succ} := 0 \right) = P$$

$\llbracket P \rrbracket$

$\text{ert}(P, \underline{0})$



$$\text{ert}(P, \underline{0}) = \text{ER}^{\llbracket P \rrbracket} (\Diamond \text{sink})$$

# Correspondence between $ert()$ and Markov chains

## Compatibility theorem

For every pGCL program  $P$  and input  $s$ :

$$ert(P, \mathbf{0})(s) = ER^{\llbracket P \rrbracket}(s, \Diamond sink)$$

In words: the  $ert(P, \mathbf{0})$  for input  $s$  equals the expected reward to reach final state  $sink$  in MC  $\llbracket P \rrbracket$  where reward function  $r$  in  $\llbracket P \rrbracket$  is defined as defined on the previous slide.

# Backward compatibility

## Deterministic programs

For any GCL program  $P$ ,  $\text{ert}(P, \mathbf{0})$  equals the number of executed computational steps<sup>2</sup> of  $P$  until  $P$  terminates.

---

<sup>2</sup>This equals the number of skip statements, guard evaluations and assignments.

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# Loops

$$\begin{array}{ccc}
 & \text{ert} & \\
 lb \leq (\text{while } (G) \{P\}) & \leq & ub \\
 \uparrow & & \uparrow \\
 \text{subinvariant} & & \text{superinvariant}
 \end{array}$$

Reasoning about loops requires — like for wp — invariants.

# Runtime invariants

$$1 + [\neg G] \cdot t + [G] \cdot \text{ert}(P, X)$$

## Runtime invariants

Let  $\Phi_t$  be the wp-characteristic function of  $P' = \text{while}(G)\{P\}$  with respect to post-runtime  $t \in \mathbb{T}$  and let  $I \in \mathbb{T}$ . Then:

# Runtime invariants

## Runtime invariants

Let  $\Phi_t$  be the wp-characteristic function of  $P' = \text{while}(G)\{P\}$  with respect to post-runtime  $t \in \mathbb{T}$  and let  $I \in \mathbb{T}$ . Then:

1.  $I$  is a **runtime-superinvariant** of  $P'$  w.r.t.  $t$  iff  $\Phi_t(I) \leq I$ .

ert (loop,  $t$ )

# Runtime invariants

## Runtime invariants

Let  $\Phi_t$  be the wp-characteristic function of  $P' = \text{while}(G)\{P\}$  with respect to post-runtime  $t \in \mathbb{T}$  and let  $I \in \mathbb{T}$ . Then:

1.  $I$  is a **runtime-superinvariant** of  $P'$  w.r.t.  $t$  iff  $\Phi_t(I) \leq I$ .
2.  $I$  is a **runtime-subinvariant** of  $P'$  w.r.t.  $t$  iff  $I \leq \Phi_t(I)$ .

If  $I$  is a **runtime-superinvariant** of  $\text{while}(G)\{P\}$  with respect to  $t \in \mathbb{T}$ , then:

$$\text{ert}(\text{while}(G)\{P\}, t) \leq I$$

# Example

while ( $c=1$ ) {  $c:=0$   $\left[\frac{1}{2}\right]$   $c:=1$  }

claim:  $\boxed{I = 1 + [c=1] b}$  is a runtime  
super  
invariant.  
wrt 0

proof:  $\Phi_0(I) \leq I$

$$\Phi_0(I) = 1 + \underbrace{[c \neq 1] \cdot 0}_{=0} + [c=1] \text{ert}(\text{body}, I)$$

$$= 1 + [c=1] \left( 1 + \frac{1}{2} \text{ert}(c:=0, I) + \frac{1}{2} \text{ert}(c:=1, I) \right)$$

$$= 1 + [c=1] \left( 1 + \frac{1}{2} \left( 1 + \underbrace{I(c:=0)}_1 \right) + \frac{1}{2} \left( 1 + \underbrace{I(c:=1)}_1 \right) \right)$$

$$= 1 + [c=1] \underbrace{(1 + 1 + 1)}_{=3}$$

Theorem says that  $I = 1 + [c=1] \cdot b$   
is an upperbound of  
ert (program)

# A **wrong** proof rule for lower bonds

Probabilistic programs do **not** satisfy:

if  $I \leq \Phi_t(I)$  then  $I \leq \text{ert}(\text{while}(G) P, t)$ .

$I$  is a subinvariant       $I$  is a lb of  $\text{ert}(\text{loop})$

These “metering” functions  $I$  do work for ordinary programs

[Frohn *et al.*, IJCAR 2016]

why? see lecture on loop invariants  
(part on Park's lemma)

# A counterexample

---

```
while (true) { skip [1/2] x++ }
```

---

- ▶ Characteristic functional  $F(X) = 1 + 1/2 (1 + 1 + X[x/x+1])$
- ▶ Least fixed point is **4** as  $F(4) = 2 + 1/2 \cdot 4 = 4$
- ▶  $4 + 2^i$  is a fixed point of  $F$  too:

$$F(4 + 2^i) = 2 + \frac{1}{2} (4 + 2^{i+1}) = 4 + 2^i$$

- ▶ Thus:  $4 + 2^i \leq F(4 + 2^i)$  but  $4 + 2^i \not\leq 4 = \text{lfp } F$
- ▶ In fact,  $4 + 2^{i+c}$  is a fixed point of  $F$  for any  $c$ :

$$F(4 + 2^{i+c}) = 2 + \frac{1}{2} (4 + 2^{i+c+1}) = 4 + 2^{i+c}$$



# Runtime $\omega$ -invariants

$$I_0, I_1, I_2, \dots$$

$$I_0 \leq I_1 \leq I_2 \dots$$

## Runtime $\omega$ -invariants

Let  $n \in \mathbb{N}$ ,  $t \in \mathbb{T}$  and  $\Phi_t$  the cert-characteristic function of  $\text{while}(G)\{P\}$ .

The monotonically increasing<sup>3</sup> sequence  $(I)_{n \in \mathbb{N}}$  is a runtime- $\omega$ -subinvariant of the loop w.r.t. runtime  $t$  iff

lb

$$I_0 \leq \Phi_t(\mathbf{0}) \quad \text{and} \quad I_{n+1} \leq \Phi_t(I_n) \quad \text{for all } n.$$

In a similar way, runtime  $\omega$ -superinvariants can be defined, but we will not use them here.

<sup>3</sup>But not necessarily strictly increasing.

# Lower bounds

$$I_0 \leq I_1 \leq I_2 \leq I_3 \quad \text{s.t.} \quad \begin{cases} I_0 \leq \Phi_t(\underline{0}) \\ I_{n+1} \leq \Phi_t(I_n) \end{cases} \quad \forall n$$

## Runtime lower bounds

If  $I_n$  is a runtime  $\omega$ -subinvariant of  $\text{while}(G)\{P\}$  with respect to  $t$ , then:

$$\sup_n I_n \leq \text{ert}(\text{while}(G) P, t)$$

## Example

Consider the same program as for proving an upper bound on the expected runtime.

$P:: \boxed{\text{while } (c=1) \{ c:=0 \mid c:=1 \}} \text{ert}(P, \underline{0})$

guess  $I_n$ 's structure. How?

$$a_0 \leq a_1 \leq \dots$$

ub =

$$1 + [c=1] \cdot b$$

$a_i$  are monotonically increasing

let  $I_n = 1 + [c=1] \cdot a_n$

↑ variable of the invariant

In order for  $I_n$  to be a w-subinvariant we have to show:

①  $I_0 \leq \underline{\Phi}_0(\underline{0})$

$$\cancel{1 + [c=1] \cdot a_0} \leq \cancel{1 + [c=1] \cdot 0} + \underbrace{[c=1]}_{=0} \cdot (\text{ert}(\text{body}, \underline{0}))$$

$$a_0 \leq 1 + \frac{1}{2} \text{ert}(c:=0, \underline{0}) + \frac{1}{2} \text{ert}(c:=1, \underline{0})$$

$$a_0 \leq 1 + \frac{1}{2} (1 + \underline{0}) + \frac{1}{2} (1 + \underline{0})$$

$$\boxed{a_0 \leq 2}$$

(\*)

②  $I_{n+1} \leq \underline{\Phi}_0(I_n)$

$$\cancel{1 + [c=1] \cdot a_{n+1}} \leq \cancel{1 + [c=1] \cdot 0} + \underbrace{[c=1]}_{=0} \cdot \text{ert}(\text{body}, I_n)$$

↑  
 $\text{body } I_n$

$$= \boxed{a_{n+1} \leq 3 + \frac{1}{2} a_n} \quad (**)$$

A possible solution is

$$a_n = 5 - \frac{3}{2^n}$$

$$n=0$$

$$a_0 = 2$$

So it follows that

$$I_n = 1 + [c-1] \left( 5 - \frac{3}{2^n} \right)$$

is a  $\omega$ -subinvariant.

According to the theorem

$$\lim_{n \rightarrow \infty} 1 + [c-1] \left( 5 - \frac{3}{2^n} \right) =$$

$$1 + [c-1] \cdot 5$$

is a lower bound to the runtime  
of our program

# Overview

- 1 Motivation
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- 3 The expected runtime transformer
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- 5 Proof rules for runtimes of loops
- 6 Proving positive almost-sure termination
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$\Pi_3$ -complete



# PAST is not compositional

Consider the two probabilistic programs:

---

```
int x := 1;
bool c := true;
while (c) {
  c := false [0.5] c := true;
  x := 2*x
}
```

---

Finite expected termination time

# PAST is not compositional

Consider the two probabilistic programs:

---

```
int x := 1;
bool c := true;
while (c) {
  c := false [0.5] c := true;
  x := 2*x
}
```

---

Finite expected termination time

PAST

---

```
while (x > 0) {
  x--
}
```

---

Finite termination time

PAST

# PAST is not compositional

$$\text{ert}(P; Q, \underline{0}) = \infty$$

$$\Rightarrow \neg \text{PAST}(P; Q)$$

Consider the two probabilistic programs:

P

---

```
int x := 1;
bool c := true;
while (c) {
  c := false [0.5] c := true;
  x := 2*x
}
```

---

Finite expected termination time

Q

---

```
while (x > 0) {
  x--
}
```

---

Finite termination time

$$\neg \text{PAST}(P; Q)$$

Running the right after the left program  
yields an **infinite** expected termination time



ert (P; Q, 0)

= ert (P, ert (Q, 0))

①

②

Q: while (x > 0)  
{ x-- }

function depending  
on 2.X

focus on lower bounds.

# Proving that PAST is not compositional (1)

$Q :$   $\frac{\text{while } (x > 0) \{ x := x-1 \}}{\text{ }}$

lower bound on  $\text{ert}(Q, \mathcal{Q})$

$\rightarrow$   $\omega$ -subinvariant

$$J_n \quad (*) \quad J_0 \leq \Phi_{Q, \mathcal{Q}}(\mathcal{Q})$$

$$J_0, J_1, J_2, J_3, \dots \quad (**) \quad J_{n+1} \leq \Phi_{Q, \mathcal{Q}}(J_n)$$

$$J_0 \leq J_1 \leq J_2 \leq \dots$$

# Proving that PAST is not compositional (1)

$n = 10$

$Q: \frac{}{\text{while } (x > 0) \{ x := x - 1 \}}$

— (\*) and (\*\*)

It is easy to check that a lower  $\omega$ -invariant is:

$$J_n = 1 + \underbrace{[0 < x < n] \cdot 2x}_{\text{on iteration}} + \underbrace{[x \geq n] \cdot (2n-1)}_{\text{on termination}}$$

check  
whether  
 $x > 0$

$x < n$

intuition:  $J_n$  is a lower b.

on running the loop  
 $n$  times

every iteration of  $Q$   
takes 2 time units

(check  $x > 0$   
and  $x--$ )

$x < n \leadsto$   
or  $\text{run } Q$   
 $x$  times

$x \geq n \leadsto$   
 $\text{run } Q$   
 $n$  times

# Proving that PAST is not compositional (1)

$$\overline{\text{while } (x > 0) \{ x := x-1 \}}$$

It is easy to check that a **lower  $\omega$ -invariant** is:

$$J_n = 1 + \underbrace{[0 < x < n] \cdot 2x}_{\text{on iteration}} + \underbrace{[x \geq n] \cdot (2n-1)}_{\text{on termination}}$$

Thus we obtain that:  $0 < x$

$$\lim_{n \rightarrow \infty} (1 + [0 < x < n] \cdot 2x + [x \geq n] \cdot (2n-1)) = 1 + [x > 0] \cdot 2x$$

use this in  $\text{ert}(P, \bullet)$

is a **lower bound** on the runtime of the above program.

# Proving that PAST is not compositional (2)

---

$P = \text{while } (c) \{ \{c := \text{false} [0.5] \ c := \text{true}\}; x := 2*x \};$   
 $Q = \text{while } (x > 0) \{ x := x-1 \}$

---

aim: a lower bound on  $\text{ert}(P; Q, \underline{0})$

$$= \text{ert}(P, \underbrace{\text{ert}(Q, \underline{0})})$$

$$1 + [x > 0] \cdot 2x$$

$\omega$ -subinvariant

$$I_0, I_1, I_2, I_3, \dots \quad I_0 \leq I_1 \leq I_2 \leq \dots$$

$$(*) \quad I_0 \leq \Phi_P(\underline{0}) \quad \wedge \quad \underbrace{\forall n \in \mathbb{N}. I_{n+1} \leq \Phi_P(I_n)}_{(**)}$$

How to find  $I_n$ ?

# Proving that PAST is not compositional (2)

$\mathcal{P}$ : `while (c) { {c := false [0.5] c := true}; x := 2*x};`  
 $\mathcal{Q}$ : `while (x > 0) { x := x-1 }`

Template for a lower  $\omega$ -invariant of composed program:

$$I_n = \underbrace{1 + [c \neq 1] \cdot (1 + [x > 0] \cdot 2x)}_{\text{on termination}} + \underbrace{[c = 1] \cdot (a_n + b_n \cdot [x > 0] \cdot 2x)}_{\text{on iteration}}$$

check c

$\text{ert}(\mathcal{Q}, \underline{0})$

$1 + [x > 0] \cdot 2x$   
 lb on  
 $\text{ert}(\mathcal{Q}, \underline{0})$

# Proving that PAST is not compositional (2)

---

```
while (c) { {c := false [0.5] c := true}; x := 2*x};
while (x > 0) { x := x-1 }
```

---

Template for a lower  $\omega$ -invariant of composed program:

$$I_n = 1 + \underbrace{[c \neq 1] \cdot (1 + [x > 0] \cdot 2x)}_{\text{on termination}} + \underbrace{[c = 1] \cdot (a_n + b_n \cdot [x > 0] \cdot 2x)}_{\text{on iteration}}$$

(\*) and (\*\*)

The constraints on being a lower  $\omega$ -invariant yield:

$$a_0 \leq 2 \quad \text{and} \quad a_{n+1} \leq 7/2 + 1/2 \cdot a_n \quad \text{and} \quad b_0 \leq 0 \quad \text{and} \quad b_{n+1} \leq 1 + b_n$$

# Proving that PAST is not compositional (2)

$\mathcal{P}$  `while (c) { {c := false [0.5] c := true}; x := 2*x};`  
 $\mathcal{Q}$  `while (x > 0) { x := x-1 }`

Template for a lower  $\omega$ -invariant of composed program:

$$\lim_{n \rightarrow \infty} I_n = 1 + \underbrace{[c \neq 1] \cdot (1 + [x > 0] \cdot 2x)}_{\text{on termination}} + \underbrace{[c = 1] \cdot (7 - \frac{5}{2^n} + n \cdot [x > 0] \cdot 2x)}_{\text{on iteration}}$$

The constraints on being a lower  $\omega$ -invariant yield:

$$a_0 \leq 2 \quad \text{and} \quad a_{n+1} \leq 7/2 + 1/2 \cdot a_n \quad \text{and} \quad b_0 \leq 0 \quad \text{and} \quad b_{n+1} \leq 1 + b_n$$

This admits the solution  $a_n = 7 - 5/2^n$  and  $b_n = n$ .



# Proving that PAST is not compositional (2)

---

```
while (c) { {c := false [0.5] c := true}; x := 2*x};
while (x > 0) { x := x-1 }
```

---

Template for a lower  $\omega$ -invariant of composed program:

$$I_n = 1 + \underbrace{[c \neq 1] \cdot (1 + [x > 0] \cdot 2x)}_{\text{on termination}} + \underbrace{[c = 1] \cdot (a_n + b_n \cdot [x > 0] \cdot 2x)}_{\text{on iteration}}$$

The constraints on being a lower  $\omega$ -invariant yield:

$$a_0 \leq 2 \quad \text{and} \quad a_{n+1} \leq 7/2 + 1/2 \cdot a_n \quad \text{and} \quad b_0 \leq 0 \quad \text{and} \quad b_{n+1} \leq 1 + b_n$$

This admits the solution  $a_n = 7 - 5/2^n$  and  $b_n = n$ . Then:  $\lim_{n \rightarrow \infty} I_n = \infty$ .

# Proving PAST

The ert-transformer enables to prove  
that a program is positively almost-surely terminating  
in a [compositional manner](#),  
although PAST itself is not a compositional property.

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# Coupon collector's problem

**ON A CLASSICAL PROBLEM OF PROBABILITY THEORY**

by

P. ERDŐS and A. RÉNYI

# Coupon collector's problem

---

```
cp := [0,...,0]; i := 1; x := 0; // no coupons yet
while (x < N) {
  while (cp[i] != 0) {
    i := uniform(1..N) // next coupon
  }
  cp[i] := 1; // coupon i obtained
  x++; // one coupon less to go
}
```

---

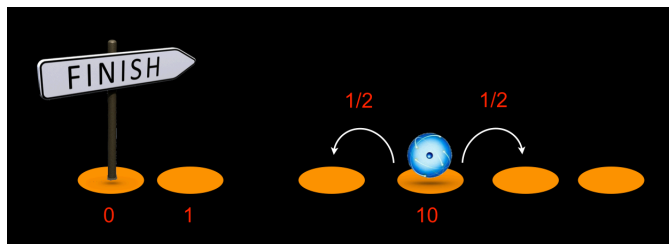
Using the ert-calculus one can prove that:

$$\text{ert}(\text{cpcl}, 0) = 4 + [N > 0] \cdot 2N \cdot (2 + H_{N-1}) \in \Theta(N \cdot \log N)$$

As Harmonic number  $H_{N-1} \in \Theta(\log N)$ .

By systematic program verification. Machine checkable.

# Random walk

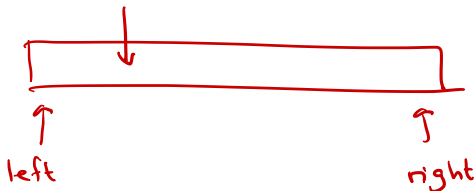


Using the ert-calculus one can prove that its expected runtime is  $\infty$ .

By systematic formal verification. Machine checkable.

# Randomised binary search

```
proc BinSearch {  
  mid := Unif(left, right); // pick mid uniformly  
  if (left < right) {  
    if (A[mid] < val) {  
      left := min(mid+1, right);  
      call BinSearch  
    } else {  
      if (A[mid] > val) {  
        right := max(mid-1, left);  
        call BinSearch  
      } else { skip }  
    } else { skip }  
  }  
}
```



# Randomised binary search

---

```
proc BinSearch {  
  mid := Unif(left, right); // pick mid uniformly  
  if (left < right) {  
    if (A[mid] < val) {  
      left := min(mid+1, right);  
      call BinSearch  
    } else {  
      if (A[mid] > val) {  
        right := max(mid-1, left);  
        call BinSearch  
      } else { skip }  
    } else { skip }  
  }  
}
```

---

Using the ert-calculus one can prove that its expected runtime is  $\Theta(\log N)$ .

By systematic formal verification. Machine checkable.