

Automatic Verification of Probabilistic Programs

Joost-Pieter Katoen & **Kevin Batz**



Lecture on Probabilistic Programming

Automatic Verification of Probabilistic Loops

How and in which cases can we automate the wp-calculus?

Automatic Verification of Probabilistic Loops

How and in which cases can we automate the wp-calculus?

What about the ert-calculus for expected runtimes?

Automatic Verification of Probabilistic Loops

How and in which cases can we automate the wp-calculus?

What about the ert-calculus for expected runtimes?

How: Quantitative Loop Invariants.

Automatic Verification of Probabilistic Loops

How and in which cases can we automate the wp-calculus?

What about the ert-calculus for expected runtimes?

How: Quantitative Loop Invariants.

In which cases: Restrict to a certain class of probabilistic loops and expectations.

Automatic Verification of Probabilistic Loops

How and in which cases can we automate the wp-calculus?

What about the ert-calculus for expected runtimes?

How: Quantitative Loop Invariants.

In which cases: Restrict to a certain class of probabilistic loops and expectations.

Motivation:

Verification of **Randomized Algorithms** and **Stochastic Processes**

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

Transmitting a single packet fails with some probability, say 0.01.

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

Transmitting a single packet fails with some probability, say 0.01.

Allow for at most F retransmission tries per packet; otherwise fail.

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

Transmitting a single packet fails with some probability, say 0.01.

Allow for at most F retransmission tries per packet; otherwise fail.

sent := 0; *fail* := 0; *totalFail* := 0;

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

Transmitting a single packet fails with some probability, say 0.01.

Allow for at most F retransmission tries per packet; otherwise fail.

```
sent := 0; fail := 0; totalFail := 0;
```

```
while (sent <  $N \wedge$  fail <  $F$ ) {
```

The Bounded Retransmission Protocol [Helmink *et al.* '93, D'Argenio *et al.* '97]

Goal: Sent file of N packets via lossy channel.

Transmitting a single packet fails with some probability, say 0.01.

Allow for at most F retransmission tries per packet; otherwise fail.

```
sent := 0; fail := 0; totalFail := 0;
```

```
while (sent <  $N \wedge$  fail <  $F$ ) {
```

```
  { fail := fail + 1; totalFail := totalFail + 1 } [0.01]
```

failed transmission

Optimal Discrete Uniform Generation from Coin Flips [Lumbroso '13]

Sample uniformly from $\{0, \dots, n - 1\}$ using **fair coin flips only**.

Optimal Discrete Uniform Generation from Coin Flips [Lumbroso '13]

Sample uniformly from $\{0, \dots, n - 1\}$ using **fair coin flips only**.

```
v := 1; c := 0; term := 0;
while (term = 0) {
  v := 2 · v;
  { c := 2 · c } [1/2] { c := 2 · c + 1 };
  if (v ≥ n) {
    if (c < n) { term := 1 }
    else { v := v − n; c := c − n }
  }
}
```

Optimal Discrete Uniform Generation from Coin Flips [Lumbroso '13]

Sample uniformly from $\{0, \dots, n - 1\}$ using **fair coin flips only**.

```
 $v := 1; c := 0; term := 0;$   
while ( $term = 0$ ) {  
   $v := 2 \cdot v;$   
   $\{c := 2 \cdot c\} [1/2] \{c := 2 \cdot c + 1\};$   
  if ( $v \geq n$ ) {  
    if ( $c < n$ ) {  $term := 1$  }  
    else {  $v := v - n; c := c - n$  }  
  }  
}
```

Is $\Pr(c = i) = 1/n$ for all $i \in \{0, \dots, n - 1\}$?

Probabilistic Programs

Recap: Weakest Preexpectations & Loop Invariants

Automatic Verification of Probabilistic Loops

Improved Automatic Verification of Probabilistic Loops

Invariant Synthesis

Conclusion

Weakest Preexpectation Reasoning

Weakest Preexpectation Reasoning

What is the expected value of program variable x after termination?

Weakest Preexpectation Reasoning

What is the expected value of program variable x after termination?

What is the probability of reaching a state $fail = F$?

Weakest Preexpectation Reasoning

What is the expected value of program variable x after termination?

What is the probability of reaching a state $fail = F$?

What is the termination probability?

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\}$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E}$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[C]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[C](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[C]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[C](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) = 1/2 \cdot x + 1/2 \cdot (x + 2)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) = 1/2 \cdot x + 1/2 \cdot (x + 2) = x + 1$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\begin{aligned} \text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) &= 1/2 \cdot x + 1/2 \cdot (x + 2) = x + 1 \\ \text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]]([x = 4]) & \end{aligned}$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\begin{aligned} \text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) &= 1/2 \cdot x + 1/2 \cdot (x + 2) = x + 1 \\ \text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]]([x = 4]) &= 1/2 \cdot [x = 4] + 1/2 \cdot [x = 2] \end{aligned}$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) = 1/2 \cdot x + 1/2 \cdot (x + 2) = x + 1$$

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]]([x = 4]) = 1/2 \cdot [x = 4] + 1/2 \cdot [x = 2]$$

$$\text{wp}[[\text{while}(c = 1) \{c := 0 [1/2] x := x + 1\}]](x)$$

Recap: Weakest Preexpectations & Loop Invariants

Consider the complete lattice $(\mathbb{E}, \sqsubseteq)$ of *expectations*:

$$\mathbb{E} = \{f \mid f: \text{States} \rightarrow \mathbb{R}_{\geq 0}^{\infty}\} \quad \text{with} \quad f \sqsubseteq g \quad \text{iff} \quad \forall \sigma \in \text{States}: f(\sigma) \leq g(\sigma)$$

Weakest preexpectation transformer:

$$\text{wp}[[C]]: \mathbb{E} \rightarrow \mathbb{E} \quad \text{wp}[[C]](f)(\sigma) = \text{expected value of } f \text{ w.r.t. distribution of final states reached after executing } C \text{ on } \sigma$$

In this lecture, we restrict to **computable** postexpectations $f: \text{States} \rightarrow \mathbb{Q}_{\geq 0}^{\infty}$.

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]](x) = 1/2 \cdot x + 1/2 \cdot (x + 2) = x + 1$$

$$\text{wp}[[\{\text{skip}\} [1/2] \{x := x + 2\}]]([x = 4]) = 1/2 \cdot [x = 4] + 1/2 \cdot [x = 2]$$

$$\text{wp}[[\text{while}(c = 1) \{c := 0 [1/2] x := x + 1\}]](x) = [c = 1] \cdot (x + 1) + [c \neq 1] \cdot x$$

Recap: Weakest Preexpectations & Loop Invariants

| C | $\text{wp} \llbracket C \rrbracket (f)$ |
|---|---|
| skip | f |
| $x := a$ | $f[x/a]$ |
| $C_1 ; C_2$ | $\text{wp} \llbracket C_1 \rrbracket (\text{wp} \llbracket C_2 \rrbracket (f))$ |
| $\{ C_1 \} [p] \{ C_2 \}$ | $p \cdot \text{wp} \llbracket C_1 \rrbracket (f) + (1 - p) \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| if (φ) { C_1 } else { C_2 } | $[\varphi] \cdot \text{wp} \llbracket C_1 \rrbracket (f) + [\neg\varphi] \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| while (φ) { <i>body</i> } | $\text{lfp } h. [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp} \llbracket \textit{body} \rrbracket (h)$ |

The lfp is taken w.r.t. \sqsubseteq on expectations.

Recap: Weakest Preexpectations & Loop Invariants

| C | $\text{wp} \llbracket C \rrbracket (f)$ |
|---|---|
| skip | f |
| $x := a$ | $f[x/a]$ |
| $C_1 ; C_2$ | $\text{wp} \llbracket C_1 \rrbracket (\text{wp} \llbracket C_2 \rrbracket (f))$ |
| $\{ C_1 \} [p] \{ C_2 \}$ | $p \cdot \text{wp} \llbracket C_1 \rrbracket (f) + (1 - p) \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| if (φ) { C_1 } else { C_2 } | $[\varphi] \cdot \text{wp} \llbracket C_1 \rrbracket (f) + [\neg\varphi] \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| while (φ) { <i>body</i> } | $\text{lfp } h. [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp} \llbracket \textit{body} \rrbracket (h)$ |

The lfp is taken w.r.t. \sqsubseteq on expectations.

Recap: Weakest Preexpectations & Loop Invariants

```
if (y = 5) {  
  
    { skip } [1/2] { x := x + 2 }  
  
} else {  
  
    x := 5  
  
}
```

Recap: Weakest Preexpectations & Loop Invariants

```
if (y = 5) {
```

```
    { skip } [1/2] { x := x + 2 }
```

```
} else {
```

```
    x := 5
```

```
}
```

```
/// x
```

x is the postexpectation

Recap: Weakest Preexpectations & Loop Invariants

```
if (y = 5) {
```

```
    { skip } [1/2] { x := x + 2 }
```

```
    /// x
```

```
} else {
```

```
    x := 5
```

```
    /// x
```

```
}
```

```
/// x
```

x is the postexpectation

Recap: Weakest Preexpectations & Loop Invariants

```
if (y = 5) {  
    { skip } [1/2] { x := x + 2 }  
    /// x  
} else {  
    /// 5  
    x := 5  
    /// x  
}  
/// x
```

x is the postexpectation

Recap: Weakest Preexpectations & Loop Invariants

```
if (y = 5) {  
  /// x + 1  
  { skip } [1/2] { x := x + 2 }  
  /// x  
} else {  
  /// 5  
  x := 5  
  /// x  
}  
/// x
```

x is the postexpectation

Recap: Weakest Preexpectations & Loop Invariants

$\lll [y = 5] \cdot (x + 1) + [y \neq 5] \cdot 5$

this is $\text{wp}[C](x)$

if ($y = 5$) {

$\lll x + 1$

{ skip } [1/2] { $x := x + 2$ }

$\lll x$

} else {

$\lll 5$

$x := 5$

$\lll x$

}

$\lll x$

x is the postexpectation

Recap: Weakest Preexpectations & Loop Invariants

$\lll [y = 5] \cdot (x + 1) + [y \neq 5] \cdot 5$

this is $\text{wp}[C](x)$

if ($y = 5$) {

$\lll x + 1$

{ skip } [1/2] { $x := x + 2$ }

$\lll x$

} else {

$\lll 5$

$x := 5$

$\lll x$

}

$\lll x$

x is the postexpectation

Observation: Determining wp for loop-free programs is automatable.

Recap: Weakest Preexpectations & Loop Invariants

| C | $\text{wp} \llbracket C \rrbracket (f)$ |
|---|---|
| skip | f |
| $x := a$ | $f [x/a]$ |
| $C_1 ; C_2$ | $\text{wp} \llbracket C_1 \rrbracket (\text{wp} \llbracket C_2 \rrbracket (f))$ |
| $\{ C_1 \} [p] \{ C_2 \}$ | $p \cdot \text{wp} \llbracket C_1 \rrbracket (f) + (1 - p) \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| if (φ) { C_1 } else { C_2 } | $[\varphi] \cdot \text{wp} \llbracket C_1 \rrbracket (f) + [\neg\varphi] \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| while (φ) { $body$ } | $\text{lfp } h. [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp} \llbracket body \rrbracket (h)$ |

Observation: Determining wp for loop-free programs is automatable.

Recap: Weakest Preexpectations & Loop Invariants

| C | $\text{wp} \llbracket C \rrbracket (f)$ |
|---|---|
| skip | f |
| $x := a$ | $f[x/a]$ |
| $C_1 ; C_2$ | $\text{wp} \llbracket C_1 \rrbracket (\text{wp} \llbracket C_2 \rrbracket (f))$ |
| $\{ C_1 \} [p] \{ C_2 \}$ | $p \cdot \text{wp} \llbracket C_1 \rrbracket (f) + (1 - p) \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| if (φ) { C_1 } else { C_2 } | $[\varphi] \cdot \text{wp} \llbracket C_1 \rrbracket (f) + [\neg\varphi] \cdot \text{wp} \llbracket C_2 \rrbracket (f)$ |
| while (φ) { $body$ } | $\text{lfp } h. [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp} \llbracket body \rrbracket (h)$ |

Observation: Determining wp for loop-free programs is automatable.

This does not hold for loops (Lec. 13). How to tackle the verification of loops?

Verifying Probabilistic Loops

... by means of quantitative loop invariants.

Recap: Weakest Preexpectations & Loop Invariants

Goal: Given computable expectations f, g , verify that $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$.

Recap: Weakest Preexpectations & Loop Invariants

Goal: Given computable expectations f, g , verify that $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$.

Example: $\text{wp}[\text{bounded retransmission protocol}](\text{fail} = F) \sqsubseteq 0.0001$

Recap: Weakest Preexpectations & Loop Invariants

Goal: Given computable expectations f, g , verify that $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$.

Example: $\text{wp}[\text{bounded retransmission protocol}](\text{fail} = F) \sqsubseteq 0.0001$

Problem: for $\text{wp}[\text{while}(\varphi)\{body\}](f) = \text{lfp } \Phi_f$, where

$$\Phi_f: \mathbb{E} \rightarrow \mathbb{E}, \quad \Phi_f(h) = [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp}[body](h) ,$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: Given computable expectations f, g , verify that $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$.

Example: $\text{wp}[\text{bounded retransmission protocol}](\text{fail} = F) \sqsubseteq 0.0001$

Problem: for $\text{wp}[\text{while}(\varphi)\{body\}](f) = \text{lfp } \Phi_f$, where

$$\Phi_f: \mathbb{E} \rightarrow \mathbb{E}, \quad \Phi_f(h) = [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp}[body](h),$$

$\text{lfp } \Phi_f$ is uncomputable in general (Lec. 13).

Goal: Given computable expectations f, g , verify that $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$.

Example: $\text{wp}[\text{bounded retransmission protocol}](\text{fail} = F) \sqsubseteq 0.0001$

Problem: for $\text{wp}[\text{while}(\varphi)\{body\}](f) = \text{lfp } \Phi_f$, where

$$\Phi_f: \mathbb{E} \rightarrow \mathbb{E}, \quad \Phi_f(h) = [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp}[body](h),$$

$\text{lfp } \Phi_f$ is uncomputable in general (Lec. 13).

Solution: Quantitative Loop Invariants

Recap: Weakest Preexpectations & Loop Invariants

$\text{wp}[\text{while}(\varphi)\{body\}](f) = \text{lfp } \Phi_f$, where

$$\Phi_f: \mathbb{E} \rightarrow \mathbb{E}, \quad \Phi_f(h) = [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp}[\text{body}](h)$$

Theorem (Superinvariants for Loops)

Let $I \in \mathbb{E}$. We have

$$\underbrace{\Phi_f(I) \sqsubseteq I}_{I \text{ is wp-superinvariant}} \quad \text{implies} \quad \text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I.$$

Recap: Weakest Preexpectations & Loop Invariants

$\text{wp}[\text{while}(\varphi)\{body\}](f) = \text{lfp } \Phi_f$, where

$$\Phi_f: \mathbb{E} \rightarrow \mathbb{E}, \quad \Phi_f(h) = [\neg\varphi] \cdot f + [\varphi] \cdot \text{wp}[\text{body}](h)$$

Theorem (Superinvariants for Loops)

Let $I \in \mathbb{E}$. We have

$$\underbrace{\Phi_f(I) \sqsubseteq I}_{I \text{ is wp-superinvariant}} \quad \text{implies} \quad \text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I.$$

Loop invariants reduce reasoning about loops to **reasoning about one loop iteration.**

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x)$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

$$\Phi_x(I) = [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\textit{body}](I)$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

`while (c = 1) {`

`{ c := 0 } [1/2] { x := x + 1 }`

`/// [c = 1] · (x + 1) + [c ≠ 1] · x`

`}`

$$\Phi_x(I) = [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\textit{body}](I)$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

```
while (c = 1) {  
  // 1/2 · (x + [c = 1] · (x + 2) + [c ≠ 1] · (x + 1))  
  { c := 0 } [1/2] { x := x + 1 }  
  // [c = 1] · (x + 1) + [c ≠ 1] · x  
}
```

$$\Phi_x(I) = [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\text{body}](I)$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

```
while (c = 1) {  
  // 1/2 · (x + [c = 1] · (x + 2) + [c ≠ 1] · (x + 1))  
  { c := 0 } [1/2] { x := x + 1 }  
  // [c = 1] · (x + 1) + [c ≠ 1] · x  
}
```

$$\begin{aligned}\Phi_x(I) &= [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\text{body}](I) \\ &= [c \neq 1] \cdot x + [c = 1] \cdot (1/2 \cdot (x + [c = 1] \cdot (x + 2) + [c \neq 1] \cdot (x + 1))) \sqsubseteq I\end{aligned}$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

```
while (c = 1) {  
  // 1/2 · (x + [c = 1] · (x + 2) + [c ≠ 1] · (x + 1))  
  { c := 0 } [1/2] { x := x + 1 }  
  // [c = 1] · (x + 1) + [c ≠ 1] · x  
}
```

$$\begin{aligned}\Phi_x(I) &= [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\text{body}](I) \\ &= [c \neq 1] \cdot x + [c = 1] \cdot (1/2 \cdot (x + [c = 1] \cdot (x + 2) + [c \neq 1] \cdot (x + 1))) \sqsubseteq I \checkmark\end{aligned}$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

```
while (c = 1) {  
  // 1/2 · (x + [c = 1] · (x + 2) + [c ≠ 1] · (x + 1))  
  { c := 0 } [1/2] { x := x + 1 }  
  // [c = 1] · (x + 1) + [c ≠ 1] · x  
}
```

$$\begin{aligned}\Phi_x(I) &= [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\text{body}](I) \\ &= \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (1/2 \cdot (x + [c = 1] \cdot (x + 2) + [c \neq 1] \cdot (x + 1)))}_{\text{When and how is this decidable?}} \sqsubseteq I \quad \checkmark\end{aligned}$$

Recap: Weakest Preexpectations & Loop Invariants

Goal: $\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (x + 1)}_{\text{choose as superinvariant } I}$

```
while (c = 1) {  
  // 1/2 · (x + [c = 1] · (x + 2) + [c ≠ 1] · (x + 1))  
  { c := 0 } [1/2] { x := x + 1 }  
  // [c = 1] · (x + 1) + [c ≠ 1] · x  
}
```

$$\begin{aligned}\Phi_x(I) &= [c \neq 1] \cdot x + [c = 1] \cdot \text{wp}[\text{body}](I) \\ &= \underbrace{[c \neq 1] \cdot x + [c = 1] \cdot (1/2 \cdot (x + [c = 1] \cdot (x + 2) + [c \neq 1] \cdot (x + 1)))}_{\text{When and how is this decidable?}} \sqsubseteq I \quad \checkmark\end{aligned}$$

Being able to decide $\Phi_f(I) \sqsubseteq I$ enables automatic verification of loops.

Automatic Verification of Loops

Automatic Verification of Loops

... by restricting expectations f, I and loops $\text{while}(\varphi) \{ \text{body} \}$ such that the wp-superinvariant condition $\Phi_f(I) \sqsubseteq I$ is **decidable**.

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

- ▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,

▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,

► $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

► $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

- ▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$
- ▶ $[c \neq 1] \cdot (x \cdot y) + [c = 1] \cdot (x + 1) \notin \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

- ▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$
- ▶ $[c \neq 1] \cdot (x \cdot y) + [c = 1] \cdot (x + 1) \notin \text{LinExp}$
- ▶ $[x \geq 1] \cdot x + [x \geq 2] \cdot y \notin \text{LinExp}$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

- ▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$
- ▶ $[c \neq 1] \cdot (x \cdot y) + [c = 1] \cdot (x + 1) \notin \text{LinExp}$
- ▶ $[x \geq 1] \cdot x + [x \geq 2] \cdot y \notin \text{LinExp}$ but
 $[x \geq 1 \wedge x \geq 2] \cdot (x + y)$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

- ▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$
- ▶ $[c \neq 1] \cdot (x \cdot y) + [c = 1] \cdot (x + 1) \notin \text{LinExp}$
- ▶ $[x \geq 1] \cdot x + [x \geq 2] \cdot y \notin \text{LinExp}$ but
 $[x \geq 1 \wedge x \geq 2] \cdot (x + y) + [x \geq 1 \wedge x \not\geq 2] \cdot x$

Definition

The set LinExp of *piecewise linear expectations* consists of all expectations f of the form

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad , \text{ where}$$

1. each φ_i is a Boolean combination of linear (in)equalities over Vars ,
2. each e_i is a linear expression over Vars with $\forall \sigma \in \text{States}: \sigma \models \varphi_i \implies e_i(\sigma) \geq 0$,
3. the φ_i partition the state space: for all $\sigma \in \text{States}$ we have $\sigma \models \varphi_j$ for exactly one j .

▶ $[c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \in \text{LinExp}$

▶ $[c \neq 1] \cdot (x \cdot y) + [c = 1] \cdot (x + 1) \notin \text{LinExp}$

▶ $[x \geq 1] \cdot x + [x \geq 2] \cdot y \notin \text{LinExp}$ but

$$[x \geq 1 \wedge x \geq 2] \cdot (x + y) + [x \geq 1 \wedge x \not\geq 2] \cdot x + [x \not\geq 1 \wedge x \not\geq 2] \cdot 0 \in \text{LinExp}$$

Definition

A loop $\text{while}(\varphi)\{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

Definition

A loop $\text{while}(\varphi) \{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

The following loop is linear:

```
while ( term = 0 ) {  
    v := 2 · v;  
    { c := 2 · c } [1/2] { c := 2 · c + 1 };  
    if ( v ≥ n ) {  
        if ( c < n ) { term := 1 }  
        else { v := v - n ; c := c - n }  
    }  
}
```

Definition

A loop $\text{while}(\varphi)\{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

Definition

A loop $\text{while}(\varphi)\{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

Lemma

Let $f, I \in \text{LinExp}$ and let $\text{while}(\varphi)\{ \text{body} \}$ be linear. Then $\Phi_f(I) \in \text{LinExp}$.

Definition

A loop $\text{while}(\varphi)\{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

Lemma

Let $f, I \in \text{LinExp}$ and let $\text{while}(\varphi)\{ \text{body} \}$ be linear. Then $\Phi_f(I) \in \text{LinExp}$.

Lemma

For $f, g \in \text{LinExp}$, it is decidable whether $f \sqsubseteq g$ holds.

Automatic Verification of Probabilistic Loops

Definition

A loop $\text{while}(\varphi)\{ \text{body} \}$ is called *linear* if *body* is loop-free and all expressions occurring in φ and *body* are linear.

Lemma

Let $f, I \in \text{LinExp}$ and let $\text{while}(\varphi)\{ \text{body} \}$ be linear. Then $\Phi_f(I) \in \text{LinExp}$.

Lemma

For $f, g \in \text{LinExp}$, it is decidable whether $f \sqsubseteq g$ holds.

Theorem

Let $f, I \in \text{LinExp}$ and let $\text{while}(\varphi)\{ \text{body} \}$ be linear. Then $\Phi_f(I) \sqsubseteq I$ is decidable.

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

Idea: Compute logical formula Θ over Vars such that

$$f \sqsubseteq g \quad \text{iff}$$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

Idea: Compute logical formula Θ over Vars such that

$$f \sqsubseteq g \quad \text{iff} \quad \underbrace{\forall \sigma \in \text{States}: \sigma \models \Theta}_{\text{decidable using existing tools}} .$$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j :

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j : $\left(\underbrace{\sigma \models \varphi_i}_{f(\sigma)=e_i(\sigma)} \right)$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j : $\left(\underbrace{\sigma \models \varphi_i}_{f(\sigma)=e_i(\sigma)} \text{ and } \underbrace{\sigma \models \psi_j}_{g(\sigma)=a_j(\sigma)} \right)$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j : $(\underbrace{\sigma \models \varphi_i}_{f(\sigma)=e_i(\sigma)} \text{ and } \underbrace{\sigma \models \psi_j}_{g(\sigma)=a_j(\sigma)})$ implies $e_i(\sigma) \leq a_j(\sigma)$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j : $(\underbrace{\sigma \models \varphi_i}_{f(\sigma)=e_i(\sigma)} \text{ and } \underbrace{\sigma \models \psi_j}_{g(\sigma)=a_j(\sigma)})$ implies $e_i(\sigma) \leq a_j(\sigma)$

iff for all $\sigma \in \text{States}$: $\sigma \models \underbrace{\bigwedge_{i=1}^n \bigwedge_{j=1}^m (\varphi_i \wedge \psi_j)}_{\text{formula of linear inequalities}} \longrightarrow e_i \leq a_j$

Automatic Verification of Probabilistic Loops

Given: $f, g \in \text{LinExp}$ with

$$f = [\varphi_1] \cdot e_1 + \dots + [\varphi_n] \cdot e_n \quad \text{and} \quad g = [\psi_1] \cdot a_1 + \dots + [\psi_m] \cdot a_m$$

Goal: Decide whether $f \sqsubseteq g$ holds.

$f \sqsubseteq g$ iff for all $\sigma \in \text{States}$: $f(\sigma) \leq g(\sigma)$

iff for all $\sigma \in \text{States}$ and all φ_i, ψ_j : $(\underbrace{\sigma \models \varphi_i}_{f(\sigma)=e_i(\sigma)} \text{ and } \underbrace{\sigma \models \psi_j}_{g(\sigma)=a_j(\sigma)})$ implies $e_i(\sigma) \leq a_j(\sigma)$

iff for all $\sigma \in \text{States}$: $\sigma \models \underbrace{\bigwedge_{i=1}^n \bigwedge_{j=1}^m (\varphi_i \wedge \psi_j)}_{\text{formula of linear inequalities}} \longrightarrow e_i \leq a_j$

The latter problem is decidable using **Satisfiability Modulo Theories** techniques.^a

^a<https://www.microsoft.com/en-us/research/project/z3-3/>

Automatic Verification of Probabilistic Loops

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Given $f, I \in \text{LinExp}$, and linear $\text{while } (\varphi) \{ \text{body} \}$:

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Given $f, I \in \text{LinExp}$, and linear $\text{while } (\varphi) \{ \text{body} \}$:

1. Compute $\Phi_f(I) \in \text{LinExp}$.

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Given $f, I \in \text{LinExp}$, and linear $\text{while } (\varphi) \{ \text{body} \}$:

1. Compute $\Phi_f(I) \in \text{LinExp}$.
2. If $\Phi_f(I) \sqsubseteq I$, then return **verify**;

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Given $f, I \in \text{LinExp}$, and linear $\text{while } (\varphi) \{ \text{body} \}$:

1. Compute $\Phi_f(I) \in \text{LinExp}$.
2. If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise return **inconclusive**.

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Given $f, I \in \text{LinExp}$, and linear $\text{while } (\varphi) \{ \text{body} \}$:

1. Compute $\Phi_f(I) \in \text{LinExp}$.
2. If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise return **inconclusive**.

Question: What to do if $\Phi_f(I) \sqsubseteq I$ does **not** hold?

Automatic Verification of Probabilistic Loops

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**.

Automatic Verification of Probabilistic Loops

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**. **Problem: Not every upper bound is a superinvariant:**

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1$$

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**. **Problem: Not every upper bound is a superinvariant:**

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1$$

We have

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5)$$

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**. **Problem: Not every upper bound is a superinvariant:**

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1$$

We have

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1.$$

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**. **Problem: Not every upper bound is a superinvariant:**

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1$$

We have

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1.$$

Even though $\text{lfp } \Phi_x \sqsubseteq x + 1$ we have $\Phi_x(x + 1) \not\sqsubseteq x + 1!$

Hence, our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

can be verified **automatically**. **Problem: Not every upper bound is a superinvariant:**

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1$$

We have

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1.$$

Even though $\text{lfp } \Phi_x \sqsubseteq x + 1$ we have $\Phi_x(x + 1) \not\sqsubseteq x + 1!$

Mitigate this problem using a more powerful proof rule: k -induction.

Improved Automatic Verification of Loops

... by means of k -induction.

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda\sigma. \min\{h(\sigma), h'(\sigma)\}$.

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda\sigma. \min\{h(\sigma), h'(\sigma)\}$.

Example:

$$[c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1$$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda\sigma. \min\{h(\sigma), h'(\sigma)\}$.

Example:

$$[c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1 = [c \neq 1] \cdot x$$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda\sigma. \min\{h(\sigma), h'(\sigma)\}$.

Example:

$$[c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1 = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1)$$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda\sigma. \min\{h(\sigma), h'(\sigma)\}$.

Theorem

For $f, I \in \mathbb{E}$ and loop $C = \text{while}(\varphi)\{ \text{body} \}$, it holds:

- ▶ 1-induction: $\Phi_f(I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda \sigma. \min\{h(\sigma), h'(\sigma)\}$.

Theorem

For $f, I \in \mathbb{E}$ and loop $C = \text{while}(\varphi)\{ \text{body} \}$, it holds:

- ▶ 1-induction: $\Phi_f(I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$
- ▶ 2-induction: $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda \sigma. \min\{h(\sigma), h'(\sigma)\}$.

Theorem

For $f, I \in \mathbb{E}$ and loop $C = \text{while}(\varphi)\{body\}$, it holds:

- ▶ 1-induction: $\Phi_f(I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{body\} \rrbracket(f) \sqsubseteq I$
- ▶ 2-induction: $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{body\} \rrbracket(f) \sqsubseteq I$
- ▶ 3-induction: $\Phi_f(\Phi_f(\Phi_f(I) \sqcap I) \sqcap I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{body\} \rrbracket(f) \sqsubseteq I$

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda \sigma. \min\{h(\sigma), h'(\sigma)\}$.

Theorem

For $f, I \in \mathbb{E}$ and loop $C = \text{while}(\varphi)\{body\}$, it holds:

- ▶ 1-induction: $\Phi_f(I) \sqsubseteq I$ implies $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$
- ▶ 2-induction: $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$ implies $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$
- ▶ 3-induction: $\Phi_f(\Phi_f(\Phi_f(I) \sqcap I) \sqcap I) \sqsubseteq I$ implies $\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$
- ▶ ...

Improved Automatic Verification of Probabilistic Loops

Given $h, h' \in \mathbb{E}$, define their *pointwise minimum* by $h \sqcap h' = \lambda \sigma. \min\{h(\sigma), h'(\sigma)\}$.

Theorem

For $f, I \in \mathbb{E}$ and loop $C = \text{while}(\varphi)\{ \text{body} \}$, it holds:

- ▶ 1-induction: $\Phi_f(I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$
- ▶ 2-induction: $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$
- ▶ 3-induction: $\Phi_f(\Phi_f(\Phi_f(I) \sqcap I) \sqcap I) \sqsubseteq I$ implies $\text{wp}\llbracket \text{while}(\varphi)\{ \text{body} \} \rrbracket(f) \sqsubseteq I$
- ▶ ...

Theorem

For $f, I \in \text{LinExp}$, linear C , and $k \geq 1$, the condition for k -induction is decidable.

Follows from the fact that for $h, h' \in \text{LinExp}$, one can compute $h \sqcap h' \in \text{LinExp}$.

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c \neq 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c \neq 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\Phi_x(\Phi_x(x + 1) \sqcap x + 1)$$

Improved Automatic Verification of Probabilistic Loops

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\Phi_x(\Phi_x(x + 1) \sqcap x + 1) = \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1)$$

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\begin{aligned} \Phi_x(\Phi_x(x + 1) \sqcap x + 1) &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1) \\ &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1)) \end{aligned}$$

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\begin{aligned} \Phi_x(\Phi_x(x + 1) \sqcap x + 1) &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1) \\ &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1)) \\ &= [c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \sqsubseteq x + 1 \end{aligned}$$

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\begin{aligned} \Phi_x(\Phi_x(x + 1) \sqcap x + 1) &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1) \\ &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1)) \\ &= [c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \sqsubseteq x + 1 \checkmark \end{aligned}$$

Back to our example

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

1-induction fails:

$$\Phi_x(x + 1) = [c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \not\sqsubseteq x + 1$$

Now consider 2-induction:

$$\begin{aligned} \Phi_x(\Phi_x(x + 1) \sqcap x + 1) &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1.5) \sqcap x + 1) \\ &= \Phi_x([c \neq 1] \cdot x + [c = 1] \cdot (x + 1)) \\ &= [c \neq 1] \cdot x + [c = 1] \cdot (x + 1) \sqsubseteq x + 1 \checkmark \end{aligned}$$

***k*-induction is more powerful and decidable in the linear setting.**

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

¹<https://github.com/moves-rwth/kipro2>

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

Given piecewise linear f, I and linear $\text{while}(\varphi)\{body\}$, `kipro2` aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$$

by checking:

¹<https://github.com/moves-rwth/kipro2>

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

Given piecewise linear f, I and linear $\text{while}(\varphi)\{body\}$, `kipro2` aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$$

by checking:

1-induction: If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise

¹<https://github.com/moves-rwth/kipro2>

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

Given piecewise linear f, I and linear $\text{while}(\varphi)\{body\}$, kipro2 aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$$

by checking:

1-induction: If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise

2-induction: If $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$, then return **verify**; otherwise

¹<https://github.com/moves-rwth/kipro2>

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

Given piecewise linear f, I and linear $\text{while}(\varphi)\{body\}$, `kipro2` aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$$

by checking:

1-induction: If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise

2-induction: If $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$, then return **verify**; otherwise

3-induction: If $\Phi_f(\Phi_f(\Phi_f(I) \sqcap I) \sqcap I) \sqsubseteq I$, then return **verify**; otherwise

¹<https://github.com/moves-rwth/kipro2>

We have implemented k -induction for probabilistic loops:

KIPRO2: k -Induction for PRObabilistic PROgrams¹

Given piecewise linear f, I and linear $\text{while}(\varphi)\{body\}$, `kipro2` aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq I$$

by checking:

1-induction: If $\Phi_f(I) \sqsubseteq I$, then return **verify**; otherwise

2-induction: If $\Phi_f(\Phi_f(I) \sqcap I) \sqsubseteq I$, then return **verify**; otherwise

3-induction: If $\Phi_f(\Phi_f(\Phi_f(I) \sqcap I) \sqcap I) \sqsubseteq I$, then return **verify**; otherwise

...

¹<https://github.com/moves-rwth/kipro2>

For C_{brp} given by

$$\text{while} (sent < N \wedge fail < F) \{$$
$$\underbrace{\{ fail := fail + 1; totalFail := totalFail + 1 \}}_{\text{failed transmission}} [0.1] \underbrace{\{ fail := 0; sent := sent + 1 \}}_{\text{successful transmission}}$$

For C_{brp} given by

$$\text{while} (sent < N \wedge fail < F) \{$$
$$\underbrace{\{ fail := fail + 1; totalFail := totalFail + 1 \}}_{\text{failed transmission}} [0.1] \underbrace{\{ fail := 0; sent := sent + 1 \}}_{\text{successful transmission}} \}$$

We fully automatically **verify** that

$$\text{wp}[\![C_{\text{brp}}]\!] (totalFail) \sqsubseteq [N \leq 3] \cdot (totalFail + 1) + [N > 3] \cdot \infty$$

by means of **is 4-induction**.

Optimal Discrete Uniform Generation from Coin Flips [Lumbroso '13]

Sample uniformly from $\{0, \dots, n - 1\}$ using **fair coin flips only**.

```
 $v := 1; c := 0; term := 0;$   
while ( $term = 0$ ) {  
   $v := 2 \cdot v;$   
   $\{c := 2 \cdot c\} [1/2] \{c := 2 \cdot c + 1\};$   
  if ( $v \geq n$ ) {  
    if ( $c < n$ ) {  $term := 1$  }  
    else {  $v := v - n; c := c - n$  }  
  }  
}
```

Optimal Discrete Uniform Generation from Coin Flips [Lumbroso '13]

Sample uniformly from $\{0, \dots, n - 1\}$ using **fair coin flips only**.

```
 $v := 1; c := 0; term := 0;$   
while ( $term = 0$ ) {  
   $v := 2 \cdot v;$   
   $\{c := 2 \cdot c\} [1/2] \{c := 2 \cdot c + 1\};$   
  if ( $v \geq n$ ) {  
    if ( $c < n$ ) {  $term := 1$  }  
    else {  $v := v - n; c := c - n$  }  
  }  
}
```

For $n \in \{2, 3, 4, 5\}$, we automatically **verify**
 $\Pr(c = i) \leq 1/n$ for all $i \in \{0, \dots, n - 1\}$
by means of **2-, 3-, and 5-induction**.

***k*-induction is more powerful and decidable in the linear setting.**

The property

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

is provable by 2-induction.

***k*-induction is more powerful and decidable in the linear setting.**

The property

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

is provable by 2-induction. **But: Not every upper bound is provable by *k*-induction:**

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

***k*-induction is more powerful and decidable in the linear setting.**

The property

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] x := x + 1 \}] (x) \sqsubseteq x + 1,$$

is provable by 2-induction. **But: Not every upper bound is provable by *k*-induction:**

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

We have $\Phi_x(2x + 1) \not\sqsubseteq 2x + 1$, $\Phi_x((\Phi_x(2x + 1) \sqcap 2x + 1)) \not\sqsubseteq 2x + 1, \dots$

***k*-induction is more powerful and decidable in the linear setting.**

The property

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq x + 1,$$

is provable by 2-induction. **But: Not every upper bound is provable by *k*-induction:**

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

We have $\Phi_x(2x + 1) \not\sqsubseteq 2x + 1$, $\Phi_x((\Phi_x(2x + 1) \sqcap 2x + 1)) \not\sqsubseteq 2x + 1, \dots$

Solution: Invariant Synthesis

Invariant Synthesis

... i.e., computing superinvariants.

Idea: To verify

$$\text{wp}[\text{while}(c = 1) \{ c := 0 [1/2] x := x + 1 \}](x) \sqsubseteq 2x + 1$$

Idea: To verify

$$\text{wp}[\text{while} (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

we **compute** an expectation I such that

1. I is a superinvariant: $\Phi_x(I) \sqsubseteq I$

Idea: To verify

$$\text{wp}[\text{while} (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

we **compute** an expectation I such that

1. I is a superinvariant: $\Phi_x(I) \sqsubseteq I$
2. I is not too large: $I \sqsubseteq 2x + 1$.

Idea: To verify

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

we **compute** an expectation I such that

1. I is a superinvariant: $\Phi_x(I) \sqsubseteq I$
2. I is not too large: $I \sqsubseteq 2x + 1$.

We then have

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \quad \underbrace{\sqsubseteq}_{I \text{ is superinvariant}} \quad I$$

Idea: To verify

$$\text{wp}[\text{while}(c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

we **compute** an expectation I such that

1. I is a superinvariant: $\Phi_x(I) \sqsubseteq I$
2. I is not too large: $I \sqsubseteq 2x + 1$.

We then have

$$\text{wp}[\text{while}(c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \underbrace{\sqsubseteq}_I I \underbrace{\sqsubseteq}_{I \text{ is not too large}} 2x + 1 \checkmark$$

I is superinvariant I is not too large

Idea: To verify

$$\text{wp}[\text{while}(c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

we **compute** an expectation I such that

1. I is a superinvariant: $\Phi_x(I) \sqsubseteq I$
2. I is not too large: $I \sqsubseteq 2x + 1$.

We then have

$$\text{wp}[\text{while}(c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \underbrace{\sqsubseteq}_I I \underbrace{\sqsubseteq}_{I \text{ is not too large}} 2x + 1 \checkmark$$

I is superinvariant I is not too large

Question: How to compute I ?

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



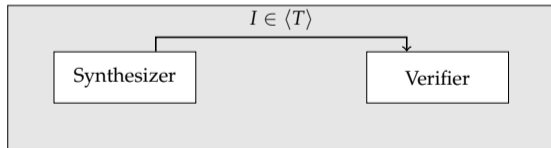
$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



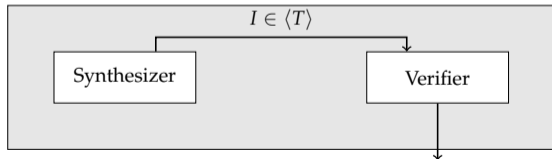
$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



I is superinvariant and not too large ✓

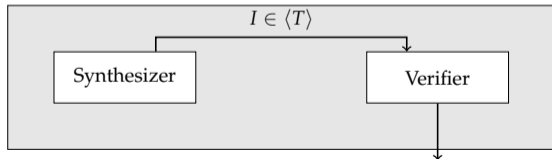
$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



I is superinvariant and not too large ✓

Remember:

$$\Phi_x(I) \sqsubseteq I \text{ iff } \forall \sigma: \Phi_x(I)(\sigma) \leq I(\sigma)$$

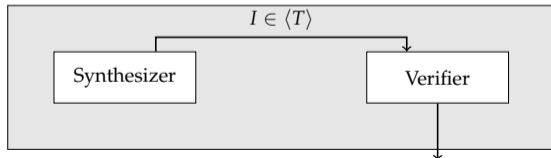
$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



I is superinvariant and not too large ✓

Remember:

$$\Phi_x(I) \sqsubseteq I \text{ iff } \forall \sigma: \Phi_x(I)(\sigma) \leq I(\sigma)$$

$$I \sqsubseteq 2x + 1 \text{ iff } \forall \sigma: I(\sigma) \leq 2 \cdot \sigma(x) + 1$$

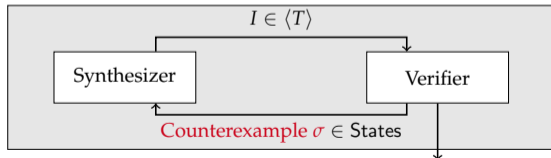
$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

Goal: Compute $I \in \text{LinExp}$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Idea: Set up a *template*

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

New goal: Find appropriate $\alpha, \beta, \gamma \in \mathbb{Q}$.



I is superinvariant and not too large ✓

Remember:

$$\Phi_x(I) \sqsubseteq I \text{ iff } \forall \sigma: \Phi_x(I)(\sigma) \leq I(\sigma)$$

$$I \sqsubseteq 2x + 1 \text{ iff } \forall \sigma: I(\sigma) \leq 2 \cdot \sigma(x) + 1$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

|

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$ |

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$ |

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$ |

Rule out the cex. σ in all future iterations.

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$ |

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$ |

$$\Phi_x(T) = [c \neq 1] \cdot x + [c = 1] \cdot \frac{1}{2} \cdot (\alpha \cdot c + \beta \cdot (x + 1) + \gamma + x)$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$ |

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$ |

$$\Phi_x(T) = [c \neq 1] \cdot x + [c = 1] \cdot \frac{1}{2} \cdot (\alpha \cdot c + \beta \cdot (x + 1) + \gamma + x)$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$ |

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$

$$\Phi_x(T)(\sigma) = \frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot (1 + 1) + \gamma + 1)$$

$$T(\sigma) = \alpha \cdot 1 + \beta \cdot 1 + \gamma$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 0) + [c \neq 1] \cdot x$

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 1$ with $\Phi_x(I)(\sigma) > I(\sigma)$

$$\Phi_x(T)(\sigma) = \frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot (1 + 1) + \gamma + 1)$$

$$T(\sigma) = \alpha \cdot 1 + \beta \cdot 1 + \gamma$$

Learn constraint $\frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot 2 + \gamma + 1) \leq \alpha \cdot 1 + \beta \cdot 1 + \gamma$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Learned constraints:

$$\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

$$\text{Synthesizer: } I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x \quad \left| \quad \begin{array}{l} \text{Learned constraints:} \\ \frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma \end{array} \right.$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x$

Verifier: **cex.** $\sigma(c) = 1, \sigma(x) = 2$ with $\Phi_x(I)(\sigma) > I(\sigma)$

Learned constraints:

$$\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

| | |
|---|--|
| Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x$ | Learned constraints: $\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$ |
| Verifier: cex. $\sigma(c) = 1, \sigma(x) = 2$ with $\Phi_x(I)(\sigma) > I(\sigma)$ | |

$$\Phi_x(T) = [c \neq 1] \cdot x + [c = 1] \cdot \frac{1}{2} \cdot (\alpha \cdot c + \beta \cdot (x + 1) + \gamma + x)$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

| | |
|---|--|
| Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x$ | Learned constraints: $\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$ |
| Verifier: cex. $\sigma(c) = 1, \sigma(x) = 2$ with $\Phi_x(I)(\sigma) > I(\sigma)$ | |

$$\Phi_x(T) = [c \neq 1] \cdot x + [c = 1] \cdot \frac{1}{2} \cdot (\alpha \cdot c + \beta \cdot (x + 1) + \gamma + x)$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

| | |
|---|--|
| Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x$ | Learned constraints: $\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$ |
| Verifier: cex. $\sigma(c) = 1, \sigma(x) = 2$ with $\Phi_x(I)(\sigma) > I(\sigma)$ | |

$$\begin{aligned} \Phi_x(T)(\sigma) &= \frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot (2 + 1) + \gamma + 1) \\ T(\sigma) &= \alpha \cdot 1 + \beta \cdot 2 + \gamma \end{aligned}$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

| | |
|--|---|
| Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 0 \cdot x + 1) + [c \neq 1] \cdot x$ Verifier: cex. $\sigma(c) = 1, \sigma(x) = 2$ with $\Phi_x(I)(\sigma) > I(\sigma)$ | Learned constraints: $\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$ |
|--|---|

$$\Phi_x(T)(\sigma) = \frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot (2 + 1) + \gamma + 1)$$

$$T(\sigma) = \alpha \cdot 1 + \beta \cdot 2 + \gamma$$

Learn constraint $\frac{1}{2} \cdot (\alpha \cdot 1 + \beta \cdot 3 + \gamma + 1) \leq \alpha \cdot 1 + \beta \cdot 2 + \gamma$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

Learned constraints:

$$\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$$

$$\frac{1}{2}(\alpha + 3\beta + \gamma + 1) \leq \alpha + 2\beta + \gamma$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

$$\text{Synthesizer: } I = [c = 1] \cdot (0 \cdot c + 1 \cdot x + 1) + [c \neq 1] \cdot x \quad \left| \quad \begin{array}{l} \text{Learned constraints:} \\ \frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma \\ \frac{1}{2}(\alpha + 3\beta + \gamma + 1) \leq \alpha + 2\beta + \gamma \end{array} \right.$$

$$\text{wp}[\text{while } (c = 1) \{ c := 0 \ [1/2] \ x := x + 1 \}] (x) \sqsubseteq 2x + 1$$

$$T = [c \neq 1] \cdot x + [c = 1] \cdot (\alpha \cdot c + \beta \cdot x + \gamma)$$

Goal: Find $I \in \langle T \rangle$ such that $\Phi_x(I) \sqsubseteq I$ and $I \sqsubseteq 2x + 1$.

| | |
|---|----------------------|
| Synthesizer: $I = [c = 1] \cdot (0 \cdot c + 1 \cdot x + 1) + [c \neq 1] \cdot x$ | Learned constraints: |
| Verifier: verified! ✓ | |

$$\frac{1}{2}(\alpha + 2\beta + \gamma + 1) \leq \alpha + \beta + \gamma$$

$$\frac{1}{2}(\alpha + 3\beta + \gamma + 1) \leq \alpha + 2\beta + \gamma$$

We have implemented invariant synthesis for probabilistic loops:

CEGISRO2: CounterExample Guided Synthesis for PRObabilistic PROgrams

We have implemented invariant synthesis for probabilistic loops:

CEGISRO2: CounterExample Guided Synthesis for PRObabilistic PROgrams

Given piecewise linear f, g and linear $\text{while}(\varphi)\{body\}$, cegispro2 aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$$

by **computing** a piecewise linear expectation I with $\Phi_f(I) \sqsubseteq I$ and $I \sqsubseteq g$.

We have implemented invariant synthesis for probabilistic loops:

CEGISRO2: CounterExample Guided Synthesis for PRObabilistic PROgrams

Given piecewise linear f, g and linear $\text{while}(\varphi)\{body\}$, cegispro2 aims to verify

$$\text{wp}[\text{while}(\varphi)\{body\}](f) \sqsubseteq g$$

by **computing** a piecewise linear expectation I with $\Phi_f(I) \sqsubseteq I$ and $I \sqsubseteq g$.

Using the same technique cegispro2 can also compute upper bounds on

$$\text{ert}[\text{while}(\varphi)\{body\}](0)$$

by computing an ert-superinvariant.

Invariant synthesis enables verifying programs with gigantic state spaces.

For C_{brp} given by

$$\text{while} \left(\text{sent} < 8 \cdot 10^9 \wedge \text{fail} < 10 \right) \{$$
$$\underbrace{\{ \text{fail} := \text{fail} + 1; \text{totalFail} := \text{totalFail} + 1 \}}_{\text{failed transmission}} [0.01] \underbrace{\{ \text{fail} := 0; \text{sent} := \text{sent} + 1 \}}_{\text{successful transmission}} \}$$

Invariant synthesis enables verifying programs with gigantic state spaces.

For C_{brp} given by

$$\text{while} \left(\text{sent} < 8 \cdot 10^9 \wedge \text{fail} < 10 \right) \{$$
$$\underbrace{\{ \text{fail} := \text{fail} + 1; \text{totalFail} := \text{totalFail} + 1 \}}_{\text{failed transmission}} [0.01] \underbrace{\{ \text{fail} := 0; \text{sent} := \text{sent} + 1 \}}_{\text{successful transmission}} \}$$

We fully automatically **verify** that

$$\text{wp}[\![C_{\text{brp}}]\!] ([\text{fail} = 10]) \sqsubseteq [\text{sent} = 0 \wedge \text{fail} = 0] \cdot 0.0001 + [\neg(N = 0 \wedge \text{fail} = 0)] \cdot \infty.$$

in 58 seconds using 106 counterexamples.

Invariant synthesis enables upper bounding **expected runtimes**.

Consider the program C modeling a race between a tortoise and a hare:

```
while ( $h \leq t$ ) {  
   $t := t + 1$ ;  
  { skip } [1/2] {  $h := 1/11 \cdot [h + 0] + \dots + 1/11 \cdot [h + 10]$  } }
```

²counting only the number of loop iterations

Invariant synthesis enables upper bounding **expected runtimes**.

Consider the program C modeling a race between a tortoise and a hare:

```
while ( $h \leq t$ ) {  
   $t := t + 1$ ;  
  { skip } [1/2] {  $h := 1/11 \cdot [h + 0] + \dots + 1/11 \cdot [h + 10]$  } }
```

We fully automatically **compute** an ert-superinvariant² and obtain

$$\text{ert}\llbracket C \rrbracket(0) \sqsubseteq [h \leq t] \cdot (2/3 \cdot t - 2/3 \cdot h + 13/3) + [h > t] \cdot 1$$

in 1.3 seconds using 32 counterexamples.

²counting only the number of loop iterations

The concepts of this lecture series provide the foundations for the **automatic verification of probabilistic programs.**

The concepts of this lecture series provide the foundations for the **automatic verification of probabilistic programs.**

Are you interested in writing a Bachelor/Master Thesis? Contact us!

The concepts of this lecture series provide the foundations for the **automatic verification of probabilistic programs.**

Are you interested in writing a Bachelor/Master Thesis? Contact us!

- ▶ Lutz Klinkenberg (Algebraic Techniques): `lutz.klinkenberg@cs.rwth-aachen.de`
- ▶ Tobias Winkler (Automata Theoretic Techniques):
`tobias.winkler@cs.rwth-aachen.de`
- ▶ Ira Fesefeldt (Concurrency): `fesefeldt@cs.rwth-aachen.de`
- ▶ Philipp Schröder (Deductive Verification): `phisch@cs.rwth-aachen.de`
- ▶ Bahar Salmani (Bayesian Networks): `salmani@cs.rwth-aachen.de`
- ▶ Kevin Batz (Automatic Verificaton): `kevin.batz@cs.rwth-aachen.de`

`thesis@i2.informatik.rwth-aachen.de`