



# Policy Iterations as Traditional Abstract Domains

---

Pierre Roux<sup>1,2</sup>    Pierre-Loïc Garoche<sup>1</sup>

Synchron 2012

November 19

<sup>1</sup>ONERA, Toulouse, France

<sup>2</sup>ISAE, Toulouse, France



---

# A Static Analysis Talk

This talk is about **static analysis** and may seem weakly related to synchronous programming. However:

- we analyze control command programs,  
typically written in the synchronous paradigm;
- we present an implementation analyzing Lustre code.

- ① State of the Art – a Policy Iteration Primer
- ② An Abstract Control Flow Graph Domain
- ③ Application to Quadratic Invariants on Linear Systems
- ④ Experimental Results
- ⑤ Conclusion and Future Work

- 1 State of the Art – a Policy Iteration Primer
- 2 An Abstract Control Flow Graph Domain
- 3 Application to Quadratic Invariants on Linear Systems
- 4 Experimental Results
- 5 Conclusion and Future Work



# A Toy Imperative Language

## Definition

$$\begin{aligned} \textit{stm} ::= & v := \textit{expr} \mid v := ?(r, r) \mid \textit{stm}; \textit{stm} \\ & \mid \textbf{if } \textit{expr} \leq 0 \textbf{ then } \textit{stm} \textbf{ else } \textit{stm} \textbf{ fi} \\ & \mid \textbf{while } \textit{expr} \leq 0 \textbf{ do } \textit{stm} \textbf{ od} \end{aligned}$$
$$\begin{aligned} \textit{expr} ::= & v \mid r \\ & \mid \textit{expr} + \textit{expr} \mid \textit{expr} - \textit{expr} \mid \textit{expr} \times \textit{expr} \mid \textit{expr} / \textit{expr} \end{aligned}$$
 $v \in \mathbb{V}, r \in \mathbb{R}.$ 

First step of analysis is to compile synchronous programs to this kind of language.



# A Toy Imperative Language, Example

## Example

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi od
```



# A Toy Imperative Language, Semantics

## Definition (Collecting Semantics)

Expressions :  $\llbracket e \rrbracket : (\mathbb{V} \rightarrow \mathbb{R}) \rightarrow \mathbb{R}$

$$\llbracket v \rrbracket(\rho) = \rho(v)$$

$$\llbracket r \rrbracket(\rho) = r$$

$$\llbracket e_1 \diamond e_2 \rrbracket(\rho) = \llbracket e_1 \rrbracket(\rho) \diamond \llbracket e_2 \rrbracket(\rho) \text{ for } \diamond \in \{ +, -, \times, / \}$$

Statements :  $\llbracket s \rrbracket : 2^{(\mathbb{V} \rightarrow \mathbb{R})} \rightarrow 2^{(\mathbb{V} \rightarrow \mathbb{R})}$

$$\llbracket v := e \rrbracket(R) = \{\rho[v \leftarrow \llbracket e \rrbracket(\rho)] \mid \rho \in R\}$$

$$\llbracket v := ?(r_1, r_2) \rrbracket(R) = \{\rho[v \leftarrow r] \mid \rho \in R, r \in \mathbb{R}, r_1 \leq r \leq r_2\}$$

$$\llbracket e \bowtie 0 \rrbracket(R) = \{\rho \in R \mid \llbracket e \rrbracket(\rho) \bowtie 0\} \text{ for } \bowtie \in \{ >, \geq, <, \leq \}$$

$$\llbracket s_1 ; s_2 \rrbracket(R) = \llbracket s_2 \rrbracket(\llbracket s_1 \rrbracket(R))$$

$$\llbracket \textbf{if } e \leq 0 \textbf{ then } s_1 \textbf{ else } s_2 \textbf{ fi} \rrbracket(R) = \llbracket s_1 \rrbracket(\llbracket e \leq 0 \rrbracket(R)) \cup \llbracket s_2 \rrbracket(\llbracket e > 0 \rrbracket(R))$$

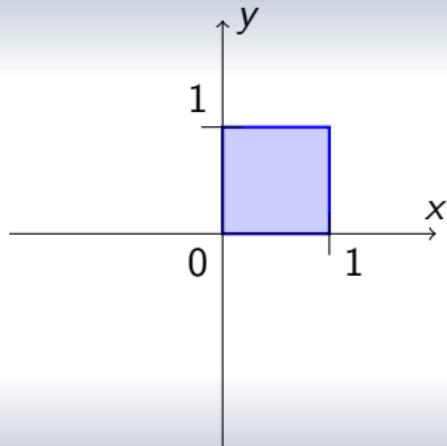
$$\llbracket \textbf{while } e \leq 0 \textbf{ do } s \textbf{ od} \rrbracket(R) = \llbracket e > 0 \rrbracket(\text{lfp}(X \mapsto R \cup X \cup \llbracket s \rrbracket(\llbracket e \leq 0 \rrbracket(X))))$$



## Kleene Iterations

With interval domain:

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



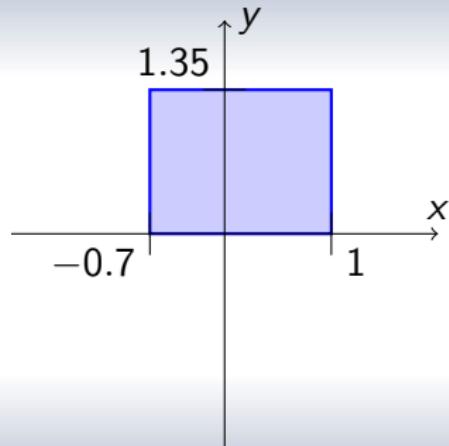
before entering the loop



## Kleene Iterations

With interval domain:

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



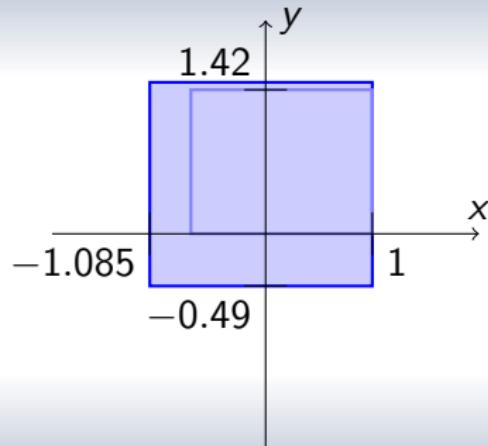
after a first iteration



## Kleene Iterations

With interval domain:

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



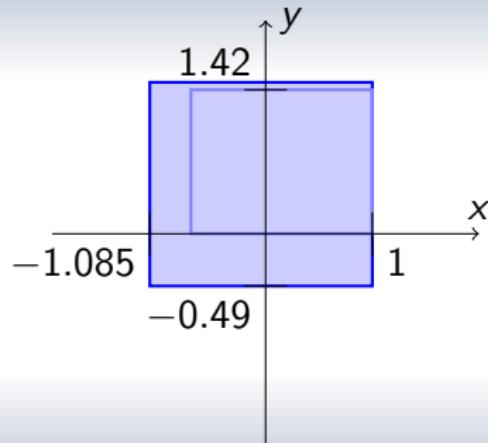
after a second iteration



## Kleene Iterations

With interval domain:

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



... does not converge



## Widening

### Basic Idea

Jump to ensure convergence in **finitely many** iterations.

### Example (Simplest Widening)

Jump to  $(-\infty, +\infty)$  as soon as bounds are not stable.



## Widening

### Basic Idea

Jump to ensure convergence in **finitely many** iterations.

### Example (Simplest Widening)

Jump to  $(-\infty, +\infty)$  as soon as bounds are not stable.

Leads to  $x \in (-\infty, +\infty) \wedge y \in (-\infty, +\infty)$



## Widening

### Basic Idea

Jump to ensure convergence in **finitely many** iterations.

### Example (Simplest Widening)

Jump to  $(-\infty, +\infty)$  as soon as bounds are not stable.

Leads to  $x \in (-\infty, +\infty) \wedge y \in (-\infty, +\infty)$ : useless result.



## Widening

### Basic Idea

Jump to ensure convergence in **finitely many** iterations.

### Example (Simplest Widening)

Jump to  $(-\infty, +\infty)$  as soon as bounds are not stable.

Leads to  $x \in (-\infty, +\infty) \wedge y \in (-\infty, +\infty)$ : useless result.

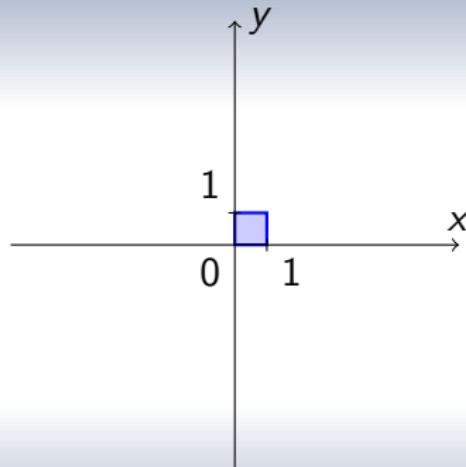
### Example (Widening with Thresholds)

Stop on a finite number of thresholds  $T$  on the way to infinity.  
For instance  $T = \{-5, 5\}$ .



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

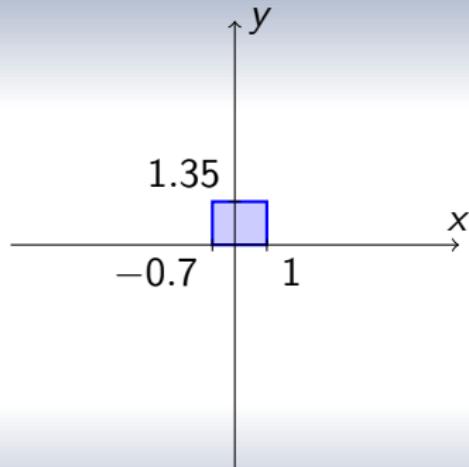


before entering the loop



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

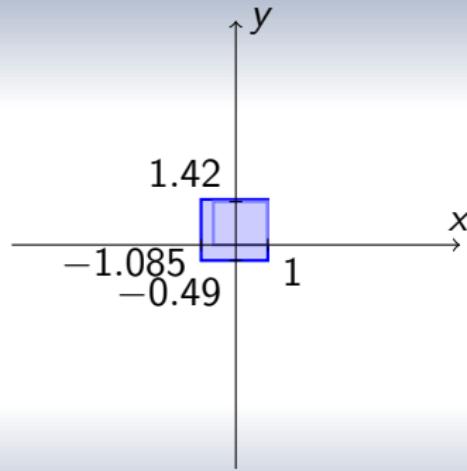


after a first iteration



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

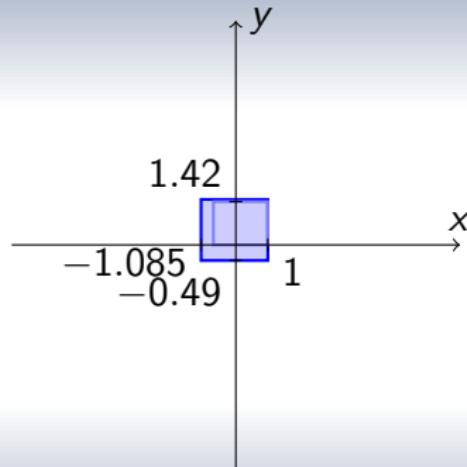


after a second iteration



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

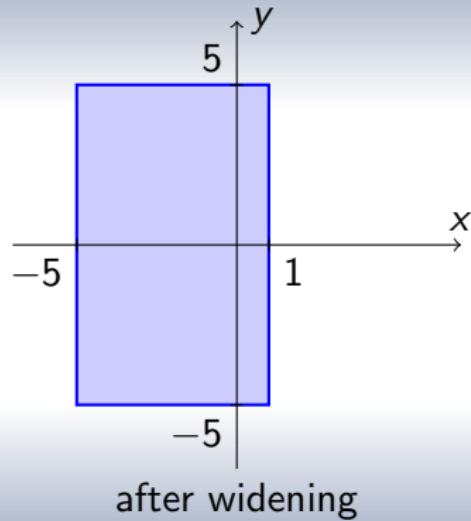


not stable → widening



# Widening with Thresholds

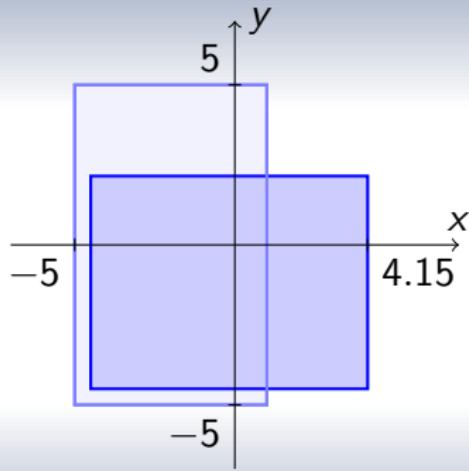
```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```





# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

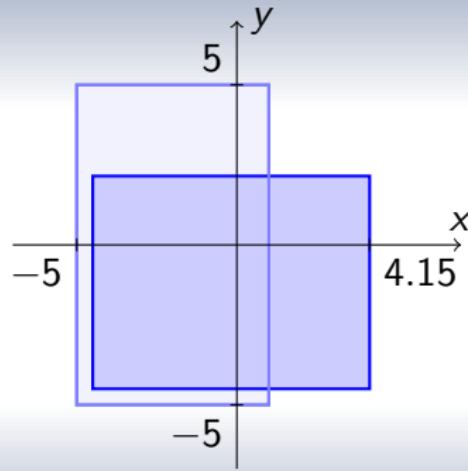


after another iteration



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

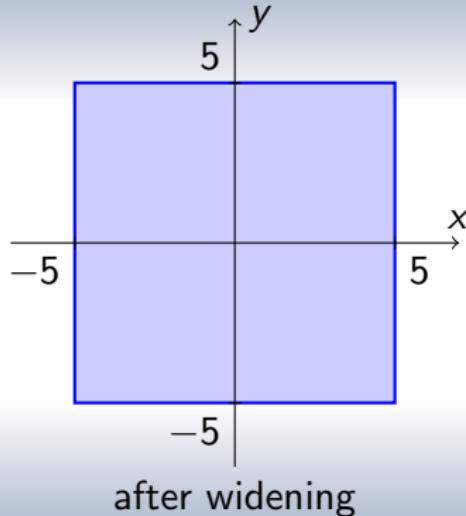


not stable → widening



## Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

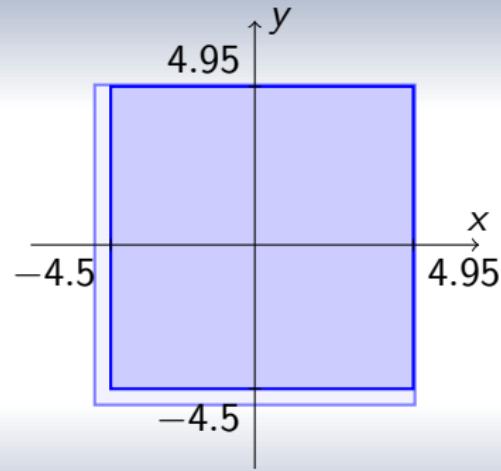


after widening



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

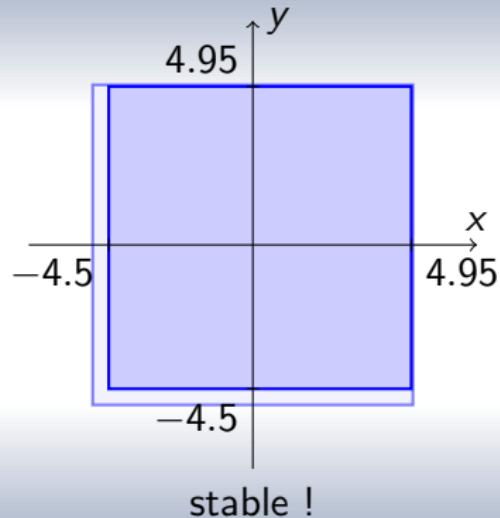


after another iteration



# Widening with Thresholds

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```





## Descending Iterations with Narrowing

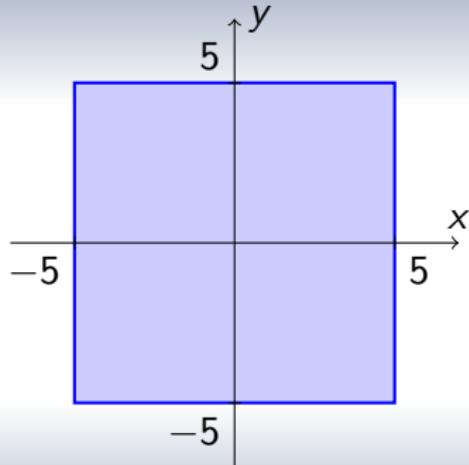
Perform a few descending iterations:

- allows to retrieve precision lost in widening;
- cannot always regain everything.



# Descending Iterations with Narrowing

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

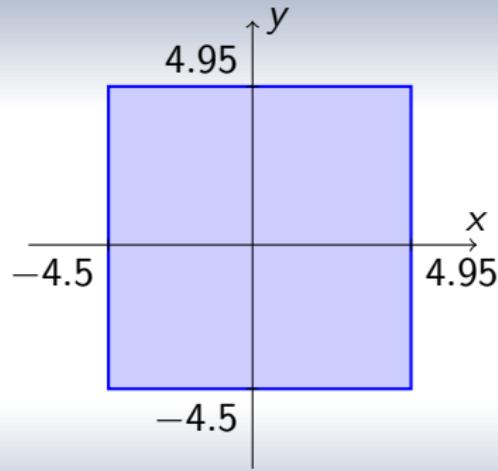


fixpoint reached with widening



# Descending Iterations with Narrowing

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

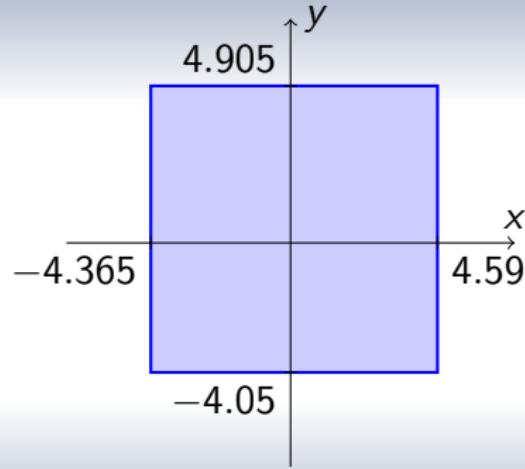


after a descending iteration



# Descending Iterations with Narrowing

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

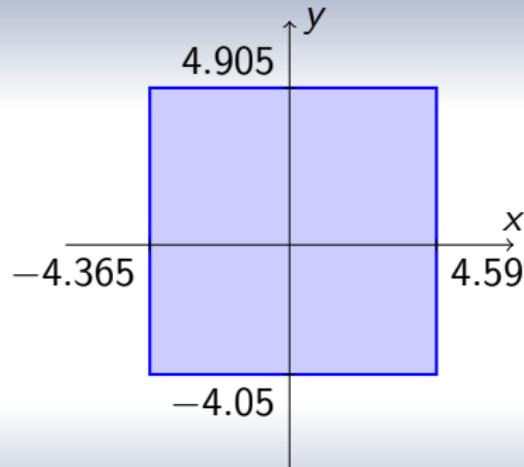


after a second iteration



# Descending Iterations with Narrowing

```
x := ?(0, 1); y := ?(0, 1);
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



not converging, we stop here



## Widening/Narrowing

- Often works well.
- But sometime incurs dramatic losses of precision.
- Lot of work to improve on this.
- Among which **policy iterations**.



# Policy Iterations

## Basic Idea

Use **numerical optimization tools** to compute bounds that are hard to guess for widening.



# Policy Iterations

## Basic Idea

Use **numerical optimization tools** to compute bounds that are hard to guess for widening.

## Example

- linear programming
- semidefinite programming

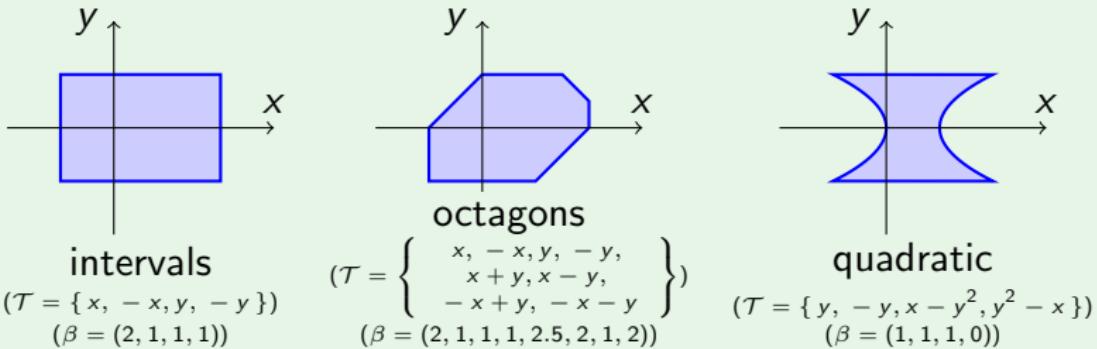


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



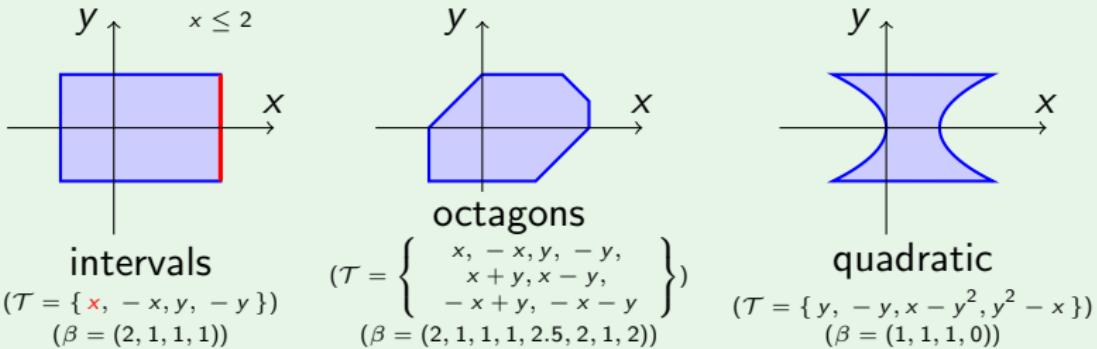


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



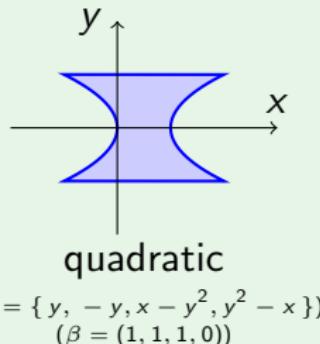
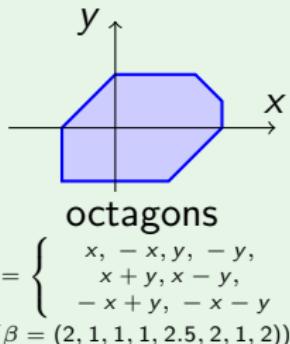
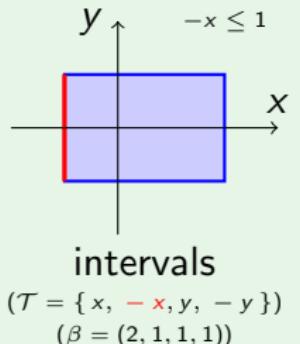


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



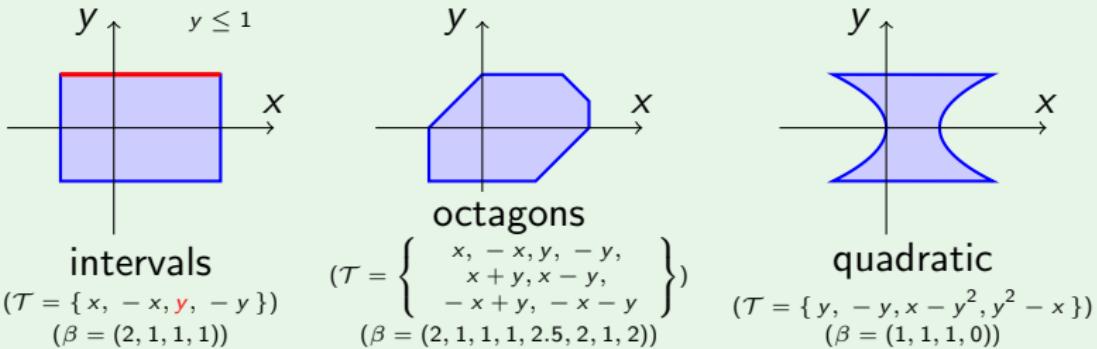


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



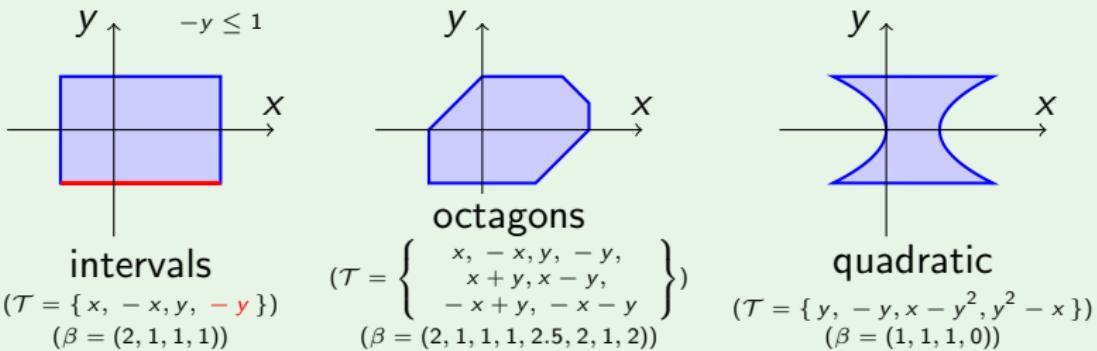


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



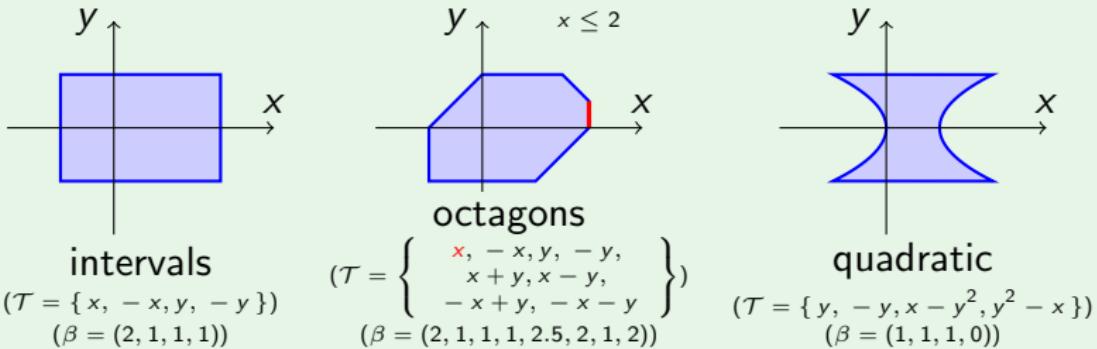


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



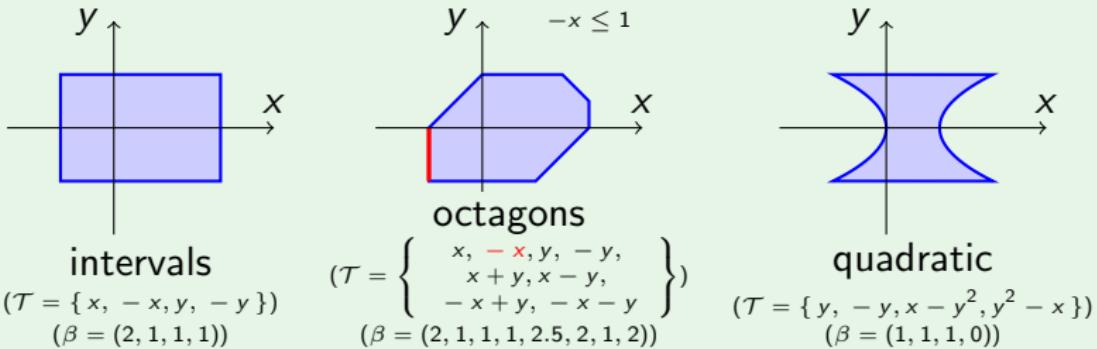


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



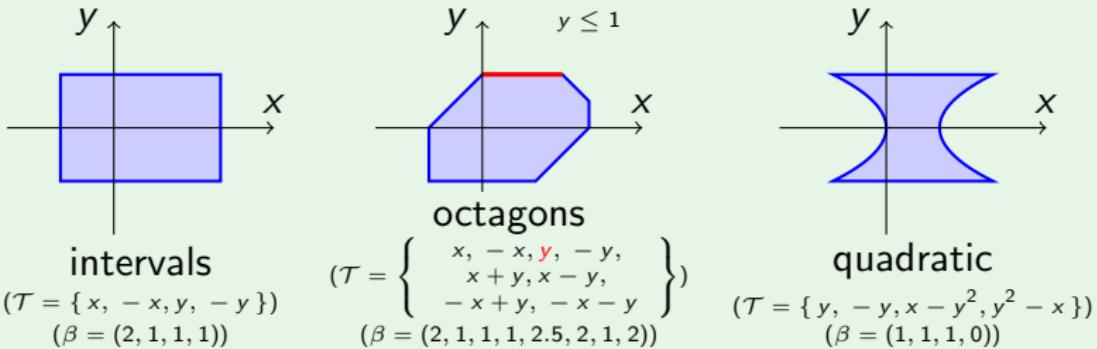


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



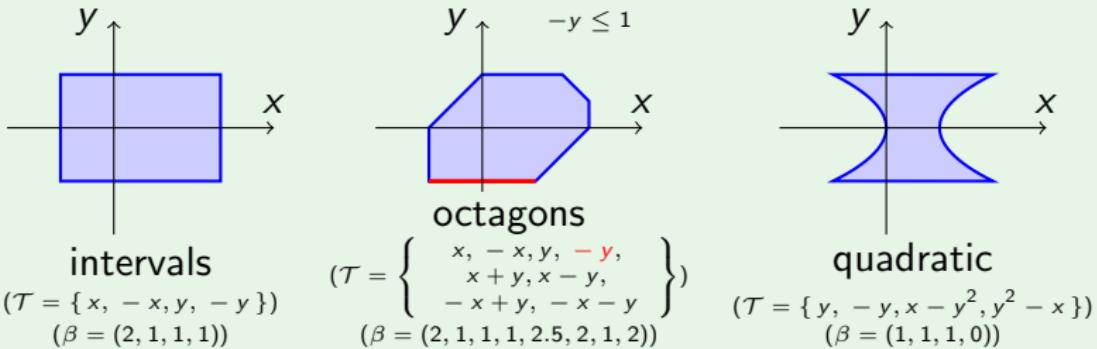


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid \llbracket t_1 \rrbracket(\rho) \leq b_1, \dots, \llbracket t_n \rrbracket(\rho) \leq b_n \}$ .

## Example



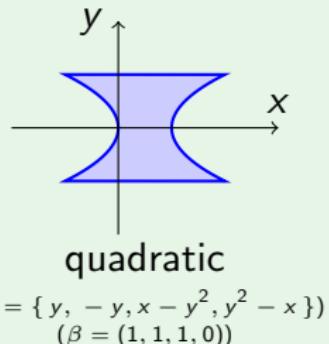
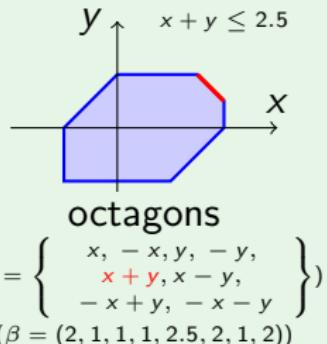
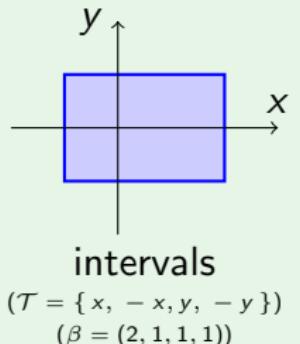


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



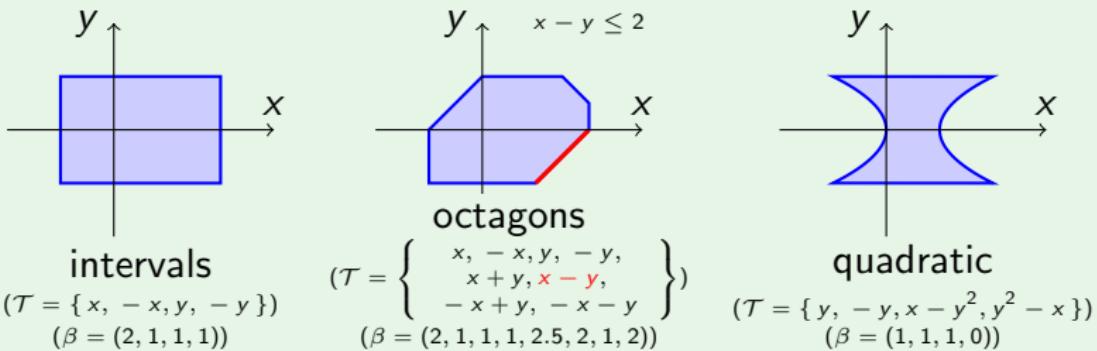


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid \llbracket t_1 \rrbracket(\rho) \leq b_1, \dots, \llbracket t_n \rrbracket(\rho) \leq b_n \}$ .

## Example



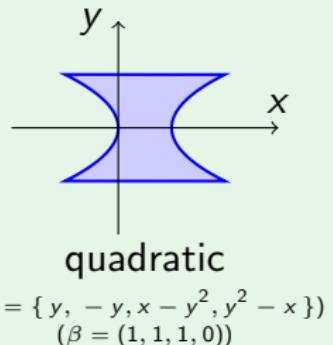
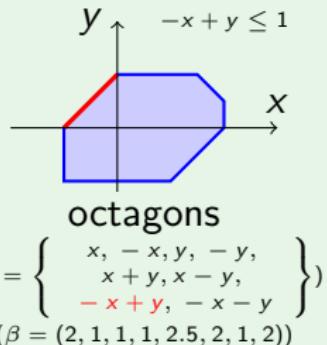
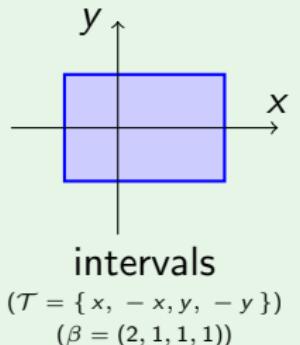


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



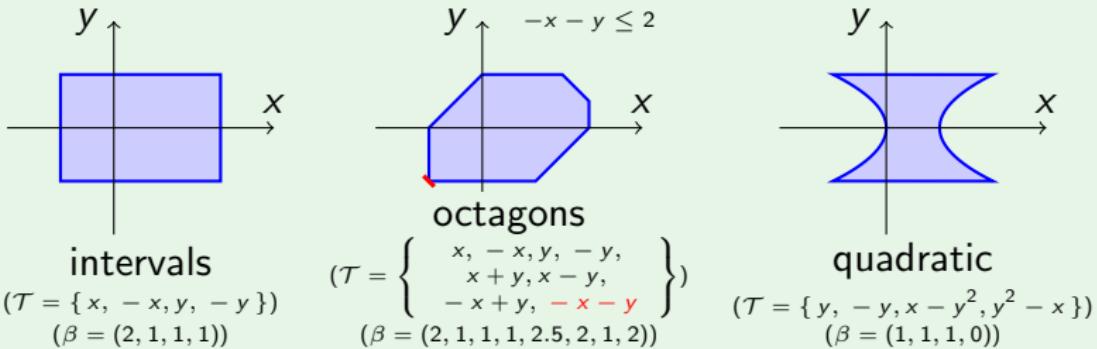


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



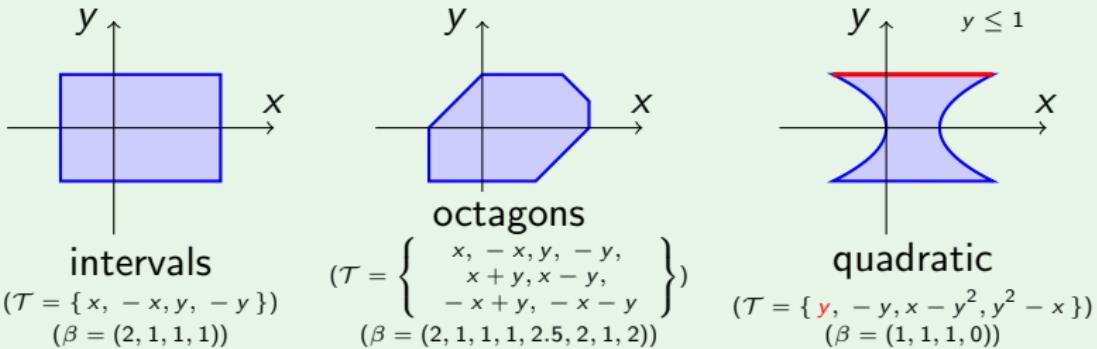


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



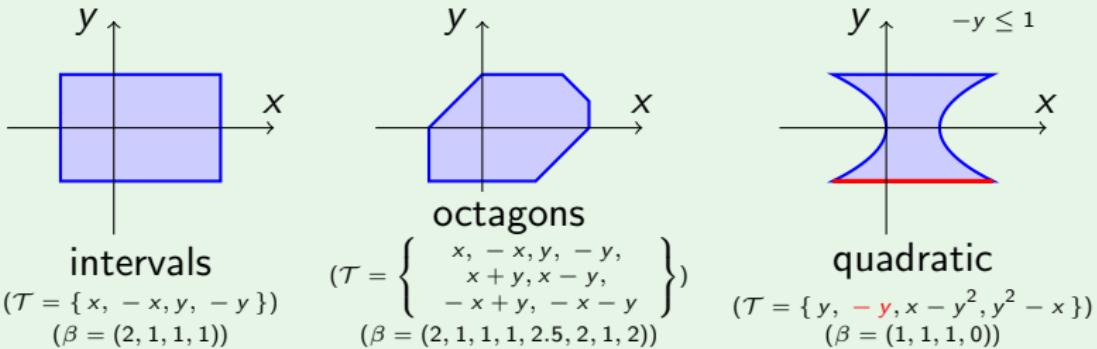


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid [\![t_1]\!](\rho) \leq b_1, \dots, [\![t_n]\!](\rho) \leq b_n \}$ .

## Example



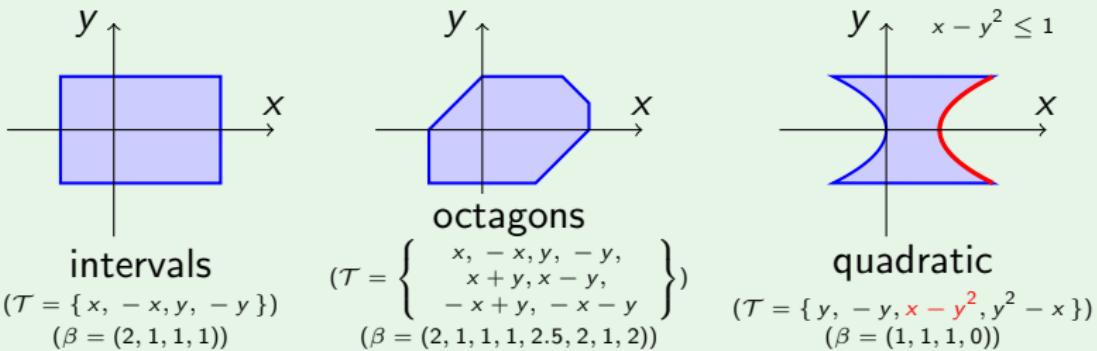


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid \llbracket t_1 \rrbracket(\rho) \leq b_1, \dots, \llbracket t_n \rrbracket(\rho) \leq b_n \}$ .

## Example



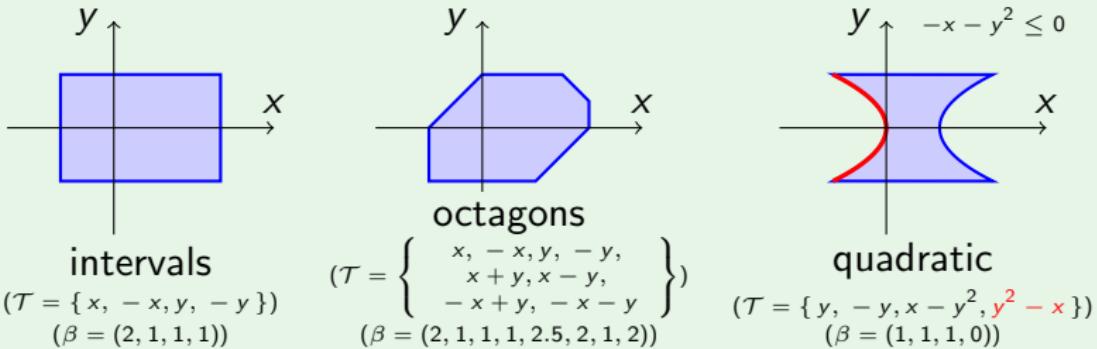


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{t_1, \dots, t_n\}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{\rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid \llbracket t_1 \rrbracket(\rho) \leq b_1, \dots, \llbracket t_n \rrbracket(\rho) \leq b_n\}$ .

## Example



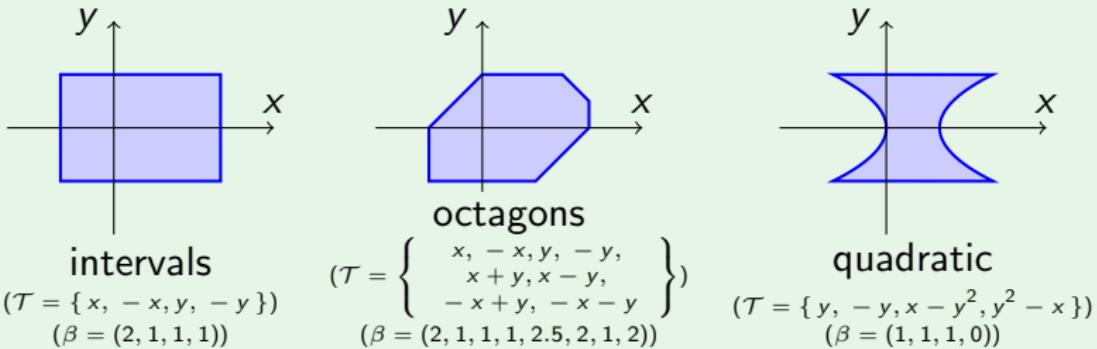


# Policy Iterations: Template Domains

## Definition (Template Domain)

Given a set of expressions  $\mathcal{T} = \{ t_1, \dots, t_n \}$ , abstract values are tuples  $\beta = (b_1, \dots, b_n) \in \bar{\mathbb{R}}^n$ , representing  $\gamma_{\mathcal{T}}(b_1, \dots, b_n) = \{ \rho \in (\mathbb{V} \rightarrow \mathbb{R}) \mid \llbracket t_1 \rrbracket(\rho) \leq b_1, \dots, \llbracket t_n \rrbracket(\rho) \leq b_n \}$ .

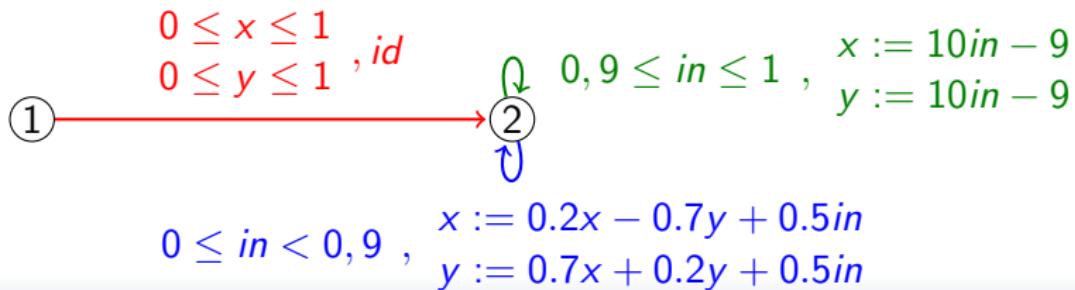
## Example





# Policy Iterations: System of Equations

First step: building the **control flow graph**.





## Policy Iterations: System of Equations

Second step: deriving a system of equations  
from the graph and the templates ( $\mathcal{T} = \{x, -x, y, -y\}$ ).

$$\left\{ \begin{array}{llll} b_{1,1} = +\infty & b_{1,2} = +\infty & b_{1,3} = +\infty & b_{1,4} = +\infty \\ b_{2,1} = \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10in - 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10in + 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10in - 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10in + 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



# Policy Iterations: Min or Max Strategy

Two main approaches to solve the equations:

- **Min-Strategy Iteration**

descending iterations (like Newton-Raphson)

- no guarantee to reach a fixpoint
- can be stopped at any time leaving a sound result
- convergence is usually fast

- **Max-Strategy Iteration**

starts from bottom and iterates greatest fixpoint computations on max-strategies until a global fixpoint is reached

- have to wait until the end for a sound result
- terminates with a precise fixpoint



# Policy Iterations: Min or Max Strategy

Two main approaches to solve the equations:

- **Min-Strategy Iteration**

descending iterations (like Newton-Raphson)

- no guarantee to reach a fixpoint
- can be stopped at any time leaving a sound result
- convergence is usually fast

- **Max-Strategy Iteration**

starts from bottom and iterates greatest fixpoint computations on max-strategies until a global fixpoint is reached

- have to wait until the end for a sound result
- terminates with a precise fixpoint

We detail the second approach.



## Policy Iterations: Max-Strategies

A **max-strategy** is the choice of one disjunct per equation:

$$\left\{ \begin{array}{l} b_{1,1} = +\infty \quad b_{1,2} = +\infty \quad b_{1,3} = +\infty \quad b_{1,4} = +\infty \\ b_{2,1} = \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10in - 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10in + 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10in - 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10in + 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Policy Iterations: Max-Strategies

A **max-strategy** is the choice of one disjunct per equation:

$$\left\{ \begin{array}{l} b_{1,1} = +\infty \quad b_{1,2} = +\infty \quad b_{1,3} = +\infty \quad b_{1,4} = +\infty \\ b_{2,1} = \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Policy Iterations: Max-Strategies

A **max-strategy** is the choice of one disjunct per equation:

$$\left\{ \begin{array}{l} b_{1,1} = +\infty \quad b_{1,2} = +\infty \quad b_{1,3} = +\infty \quad b_{1,4} = +\infty \\ b_{2,1} = \max\{10 \cdot in - 9 \mid 0.9 \leq in \leq 1 \wedge \text{be}(2)\} \\ \\ b_{2,2} = \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \\ b_{2,3} = \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \\ b_{2,4} = \max\{-0.7x - 0.2y - 0.5in \mid 0 \leq in \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



# Max-Strategies Iterations

## Algorithm

- add a disjunct  $-\infty$  to each equation
- initialize strategy  $\sigma_0$  to all the  $-\infty$  and  $\beta_0$  to the tuple  $(-\infty, \dots, -\infty)$
- iterates on strategies until there is no more improving strategy
  - find an improving strategy  $\sigma_{i+1}$   
(i.e.  $\sigma_{i+1}$  evaluated at  $\beta_i$  reaches values greater than  $\beta_i$ )
  - compute  $\beta_{i+1}$  the greatest fixpoint of  $\sigma_{i+1}$



## Max-Strategies Iterations: Example

We first add the  $-\infty$  disjunct:

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

Then initiate strategy  $\sigma_0$  and  $\beta_0 = (-\infty, \dots, -\infty)$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

We find an improving strategy  $\sigma_1$ :

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

And compute the greatest fixpoint

$$\beta_1 = (+\infty, +\infty, +\infty, +\infty, -\infty, -\infty, -\infty, -\infty)$$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

We find an improving strategy  $\sigma_2$ :

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

And compute the greatest fixpoint  $\beta_2 = (+\infty, +\infty, +\infty, +\infty, 1, 0, 1, 0)$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

We find an improving strategy  $\sigma_3$ :

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

And compute the greatest fixpoint

$$\beta_3 = (+\infty, +\infty, +\infty, +\infty, 1, 1.12578125, 1.14375, 0)$$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

We find an improving strategy  $\sigma_4$ :

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

And compute the greatest fixpoint

$$\beta_4 = (+\infty, +\infty, +\infty, +\infty, 1, 1.12578125, 1.14375, 1.1005859375)$$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

We find an improving strategy  $\sigma_5$ :

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

And compute the greatest fixpoint

$$\beta_5 \approx (+\infty, +\infty, +\infty, +\infty, 2.27, 2.23, 2.55, 1.95)$$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{ in } -9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{ in } +9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{ in } \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max-Strategies Iterations: Example

No more improving strategy, result:

$$-2.24 \leq x \leq 2.27 \wedge -1.96 \leq y \leq 2.55.$$

$$\left\{ \begin{array}{ll} b_{1,1} = -\infty \vee +\infty & b_{1,2} = -\infty \vee +\infty \\ b_{1,3} = -\infty \vee +\infty & b_{1,4} = -\infty \vee +\infty \\ b_{2,1} = -\infty \vee \max\{x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.2x - 0.7y + 0.5 \text{in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,2} = -\infty \vee \max\{-x \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.2x + 0.7y - 0.5 \text{in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,3} = -\infty \vee \max\{y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{10 \text{in} - 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{0.7x + 0.2y + 0.5 \text{in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \\ b_{2,4} = -\infty \vee \max\{-y \mid 0 \leq x \leq 1 \wedge 0 \leq y \leq 1 \wedge \text{be}(1)\} \\ \quad \vee \max\{-10 \text{in} + 9 \mid 0.9 \leq \text{in} \leq 1 \wedge \text{be}(2)\} \\ \quad \vee \max\{-0.7x - 0.2y - 0.5 \text{in} \mid 0 \leq \text{in} \leq 0.9 \wedge \text{be}(2)\} \end{array} \right.$$

with  $\text{be}(i)$  still a shortcut for  $x \leq b_{i,1} \wedge -x \leq b_{i,2} \wedge y \leq b_{i,3} \wedge -y \leq b_{i,4}$



## Max Strategies Iterations: Termination

- each strategy is considered at most once



## Max Strategies Iterations: Termination

- each strategy is considered at most once
- there is an exponential number of strategies



## Max Strategies Iterations: Termination

- each strategy is considered at most once
  - there is an exponential number of strategies
- ⇒ bounds the number of iterations



## Max Strategies Iterations: Termination

- each strategy is considered at most once
  - there is an exponential number of strategies
- ⇒ bounds the number of iterations
- in practice, only a few strategies are considered



# Policy Iterations vs Kleene Iterations

- policy iterations are an efficient alternative to widening



## Policy Iterations vs Kleene Iterations

- policy iterations are an efficient alternative to widening
- however their use seems orthogonal to traditional Kleene iterations



## Policy Iterations vs Kleene Iterations

- policy iterations are an efficient alternative to widening
- however their use seems orthogonal to traditional Kleene iterations
- this prevents an easy use in existing static analyzers and collaboration with existing domains through reduced products



## Policy Iterations vs Kleene Iterations

- policy iterations are an efficient alternative to widening
  - however their use seems orthogonal to traditional Kleene iterations
  - this prevents an easy use in existing static analyzers and collaboration with existing domains through reduced products
- ⇒ we want to integrate policy iterations into a traditional abstract domain

- 1 State of the Art – a Policy Iteration Primer
- 2 An Abstract Control Flow Graph Domain
- 3 Application to Quadratic Invariants on Linear Systems
- 4 Experimental Results
- 5 Conclusion and Future Work



## How to Integrate Policy Iterations into a Traditional Abstract Domain

- policy iterations require a system of equations
  - system of equations is immediately derived from control flow graph
- ⇒ we will develop an abstract domain to build the graph

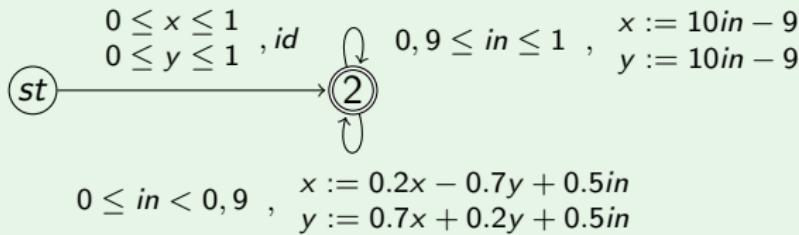


# Lattice Structure

We use graphs with

- edges of the form *constraint, assignment*
- special vertices *st* and *fi*
  - *st* : starting point
  - *fi* : temporary final point
- a doubly circled vertex, indicating the “current point”

## Example





## Lattice Structure: Order

Basically, we define  $g_1 \sqsubseteq_{\mathcal{G}}^{\#} g_2$  if

- for all edge from  $u$  to  $v$  in  $g_2$ 
  - there is no edge from  $u$  to  $v$  with the same assignment
  - or all edges with the same assignment have stronger constraints
- and for all edge from  $u$  to  $v$  in  $g_1$ 
  - there is at least one edge from  $u$  to  $v$  with the same assignment in  $g_2$  having weaker constraints
- or we can reach the previous case by redirecting all edges from  $f_i$  in  $g_2$  to the “current point” of  $g_1$  (see widening, later)

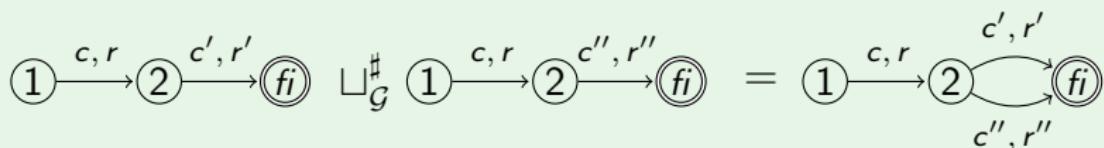


## Lattice Structure: Order

Basically, we define  $g_1 \sqsubseteq_{\mathcal{G}}^{\sharp} g_2$  if

- for all edge from  $u$  to  $v$  in  $g_2$ 
  - there is no edge from  $u$  to  $v$  with the same assignment
  - or all edges with the same assignment have stronger constraints
- and for all edge from  $u$  to  $v$  in  $g_1$ 
  - there is at least one edge from  $u$  to  $v$  with the same assignment in  $g_2$  having weaker constraints
- or we can reach the previous case by redirecting all edges from  $fi$  in  $g_2$  to the “current point” of  $g_1$  (see widening, later)

### Example (Join)





# Abstract Operators: Guards

## Example

$$\llbracket x \leq 0 \rrbracket^\sharp(\top_G) = \textcircled{st} \xrightarrow{x \leq 0, id} \textcircled{fi}$$

(a) case  $g = \top_G$

$$\llbracket x \leq 0 \rrbracket^\sharp\left(\textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{fi}\right) = \textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{fi}$$

(b) case  $g = (e, fi)$

$$\llbracket x \leq 0 \rrbracket^\sharp\left(\textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{t}\right) = \textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{t} \xrightarrow{x \leq 0, id} \textcircled{fi}$$

(c) case  $g = (e, t), t \neq fi$



# Abstract Operators: Assignments

## Example

$$[x := x + 1]^\sharp(\top_G) = \textcircled{st} \xrightarrow{\text{true}, x := x + 1} \textcircled{fi}$$

(a) case  $g = \top_G$

$$[x := x + 1]^\sharp \left( \textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{fi} \right) = \textcircled{st} \xrightarrow[r[x := r(x) + 1]]{} \textcircled{fi}$$

(b) case  $g = (e, fi)$

$$[x := x + 1]^\sharp \left( \textcircled{st} \xrightarrow{y \leq 0, r} \textcircled{t} \right) = \textcircled{st} \xrightarrow[\text{true}, x := x + 1]{} \textcircled{t} \xrightarrow{} \textcircled{fi}$$

(c) case  $g = (e, t), t \neq fi$



# Abstract Operators: Random Assignments

## Definition

$$\llbracket x := ?(r_1, r_2) \rrbracket^\sharp(g) := \llbracket r_1 - x_{\text{ex}} \leq 0 \rrbracket^\sharp \left( \llbracket x_{\text{ex}} - r_2 \leq 0 \rrbracket^\sharp(g') \right)$$

with  $g' := \llbracket x := x_{\text{ex}} \rrbracket^\sharp(g)$

with  $x_{\text{ex}}$  an extra variable not appearing anywhere in  $g$



## Widening

- the domain  $\mathcal{G}$  has infinite ascending chains  
⇒ a widening is required to enforce termination of the analyses



## Widening

- the domain  $\mathcal{G}$  has infinite ascending chains  
⇒ a widening is required to enforce termination of the analyses
- widening will
  - add nodes for loop heads
  - close the loops



## Widening: Examples

### Example

$$\perp_{\mathcal{G}} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(a) both code pointers equal to fi ( $\perp_{\mathcal{G}} = (\perp_c, fi)$ )



# Widening: Examples

## Example

$$\perp_{\mathcal{G}} \quad \nabla_{\mathcal{G}} \quad \textcircled{st} \xrightarrow{e, r} \textcircled{fi} \quad = \quad \textcircled{st} \xrightarrow{e, r} \textcircled{2}$$

(a) both code pointers equal to  $fi$  ( $\perp_{\mathcal{G}} = (\perp_c, fi)$ )

$$\textcircled{st} \xrightarrow{e, r} \textcircled{2} \quad \nabla_{\mathcal{G}} \quad \textcircled{st} \xrightarrow{e, r} \textcircled{2} \xrightarrow{e', r'} \textcircled{fi} \quad = \quad \textcircled{st} \xrightarrow{e, r} \textcircled{2}$$

(b) one code pointer not equal to  $fi$



# Widening: Examples

## Example

$$\perp_{\mathcal{G}} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(a) both code pointers equal to *fi* ( $\perp_{\mathcal{G}} = (\perp_c, fi)$ )

$$\text{(st)} \xrightarrow{e, r} \text{(2)} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e', r'} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(b) one code pointer not equal to *fi*

$$\text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e', r'} \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e'', r'} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(c) one code pointer not equal to *fi*



# Widening: Examples

## Example

$$\perp_{\mathcal{G}} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(a) both code pointers equal to fi ( $\perp_{\mathcal{G}} = (\perp_c, fi)$ )

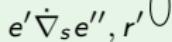
$$\text{(st)} \xrightarrow{e, r} \text{(2)} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e', r'} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(b) one code pointer not equal to fi



$$\text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e', r'} \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e, r} \text{(2)} \xrightarrow{e'', r'} \text{(fi)} = \text{(st)} \xrightarrow{e, r} \text{(2)}$$

(c) one code pointer not equal to fi



$$\text{(st)} \xrightarrow{e, r} \text{(2)} \quad \nabla_{\mathcal{G}} \quad \text{(st)} \xrightarrow{e', r'} \text{(3)} = \top_{\mathcal{G}}$$

(d) different code pointers



## Example

 $\top_{\mathcal{G}}$ 

(a)  $x := ?(0, 1); \quad y := ?(0, 1); \quad \top_{\mathcal{G}}$   
**while**  $-1 \leq 0$  **do**  
     $in := ?(0, 1);$   
    **if**  $0.9 - in \leq 0$  **then**  
         $x := 10 \times in - 9;$   
         $y := 10 \times in - 9$   
    **else**  
         $t := x;$   
         $x := 0.2 \times t - 0.7 \times y + 0.5 \times in;$   
         $y := 0.7 \times t + 0.2 \times y + 0.5 \times in$   
    **fi**  
**od**



## Example

```
(a) x := ?(0, 1);  
     y := ?(0, 1);  
while -1 ≤ 0 do  
    in := ?(0, 1);  
    if 0.9 - in ≤ 0 then  
        x := 10×in - 9;  
        y := 10×in - 9  
    else  
        t := x;  
        x := 0.2×t - 0.7×y + 0.5×in;  
        y := 0.7×t + 0.2×y + 0.5×in  
    fi  
od
```

 $\top_{\mathcal{G}}$ 

(a)

(b)  $[\![x := ?(0, 1)]\!]^\sharp(a)$

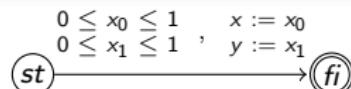


# Example

```
(a) x := ?(0, 1); (b) y := ?(0, 1); (c)
while -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

$\top_{\mathcal{G}}$

(a)

(b)  $\llbracket x := ?(0, 1) \rrbracket^{\sharp}(a)$ (c)  $\llbracket y := ?(0, 1) \rrbracket^{\sharp}(b)$

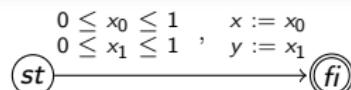
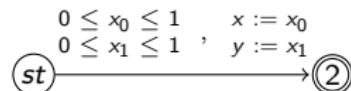


# Example

```
(a) x := ?(0, 1); (b) y := ?(0, 1); (c)
while(d) -1 ≤ 0 do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

$\top_{\mathcal{G}}$

(a)

(b)  $\llbracket x := ?(0, 1) \rrbracket^{\sharp}(a)$ (c)  $\llbracket y := ?(0, 1) \rrbracket^{\sharp}(b)$ (d)  $\perp \nabla_{\mathcal{G}}(c)$

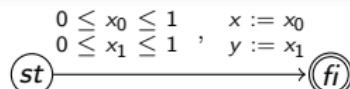
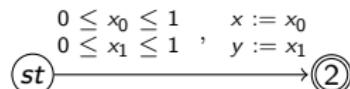
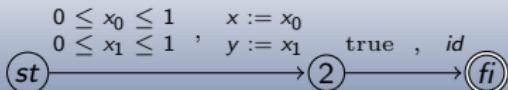


## Example

```
(a) x := ?(0, 1); (b) y := ?(0, 1); (c)
while(d) -1 ≤ 0(e) do
    in := ?(0, 1);
    if 0.9 - in ≤ 0 then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

$\top_{\mathcal{G}}$

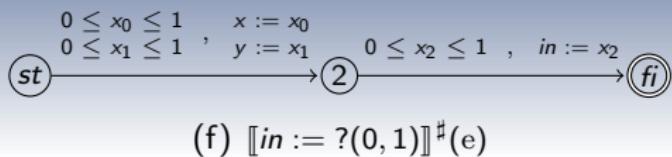
(a)

(b)  $\llbracket x := ?(0, 1) \rrbracket^{\sharp}(a)$ (c)  $\llbracket y := ?(0, 1) \rrbracket^{\sharp}(b)$ (d)  $\perp \nabla_{\mathcal{G}}(c)$ (e)  $\llbracket -1 ≤ 0 \rrbracket^{\sharp}(d)$



## Example

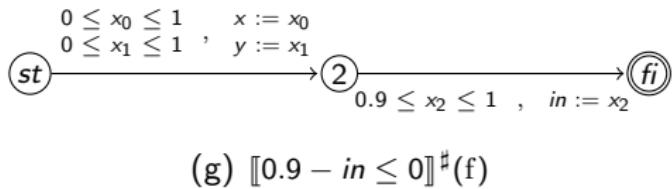
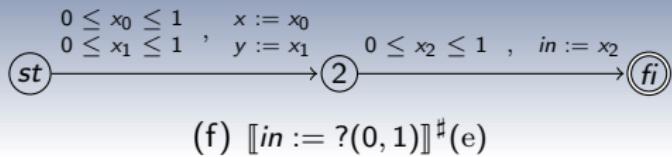
```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0   then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```





# Example

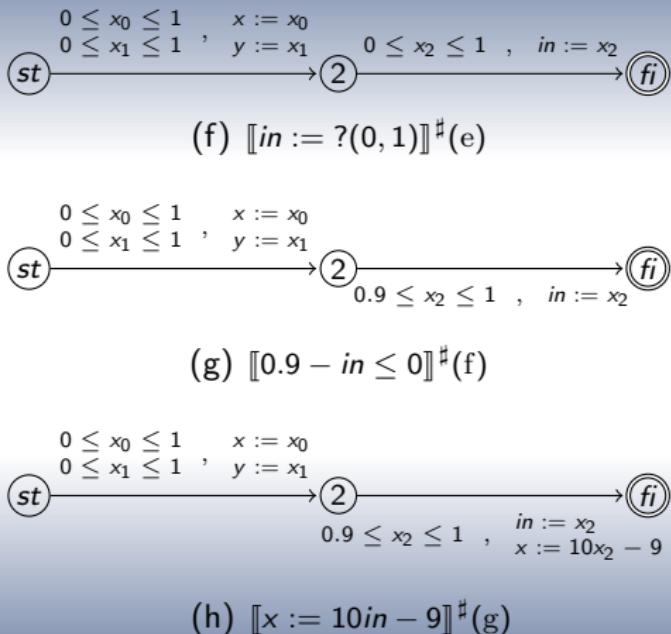
```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```





# Example

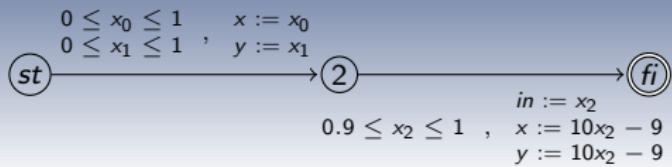
```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```





# Example

```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

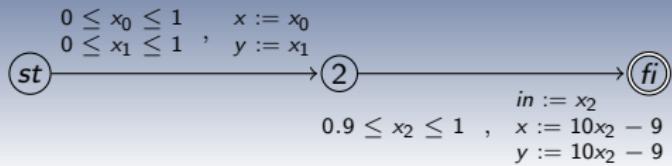


(i)  $\llbracket y := 10in - 9 \rrbracket^\sharp(h)$

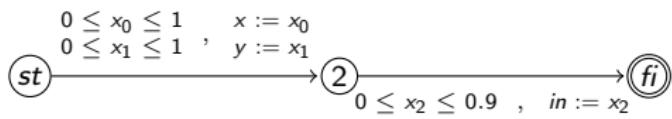


# Example

```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else(j)
        t := x;
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



$$(i) \llbracket y := 10in - 9 \rrbracket^{\sharp}(h)$$

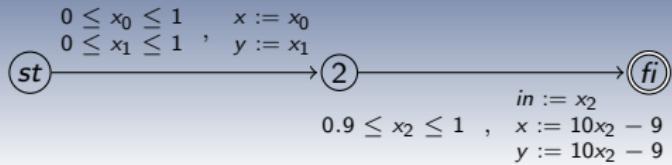


$$(j) \llbracket 0.9 - in ≥ 0 \rrbracket^{\sharp}(f)$$

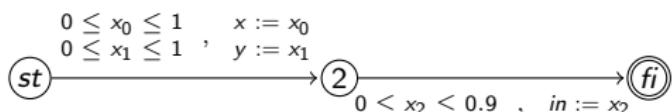


# Example

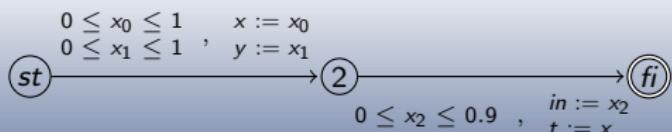
```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else(j)
        t := x;(k)
        x := 0.2×t - 0.7×y + 0.5×in;
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```



$$(i) \llbracket y := 10in - 9 \rrbracket^{\sharp}(h)$$



$$(j) \llbracket 0.9 - in ≥ 0 \rrbracket^{\sharp}(f)$$

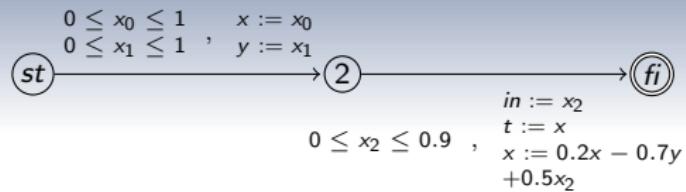


$$(k) \llbracket t := x \rrbracket^{\sharp}(j)$$



# Example

```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else(j)
        t := x;(k)
        x := 0.2×t - 0.7×y + 0.5×in;(l)
        y := 0.7×t + 0.2×y + 0.5×in
    fi
od
```

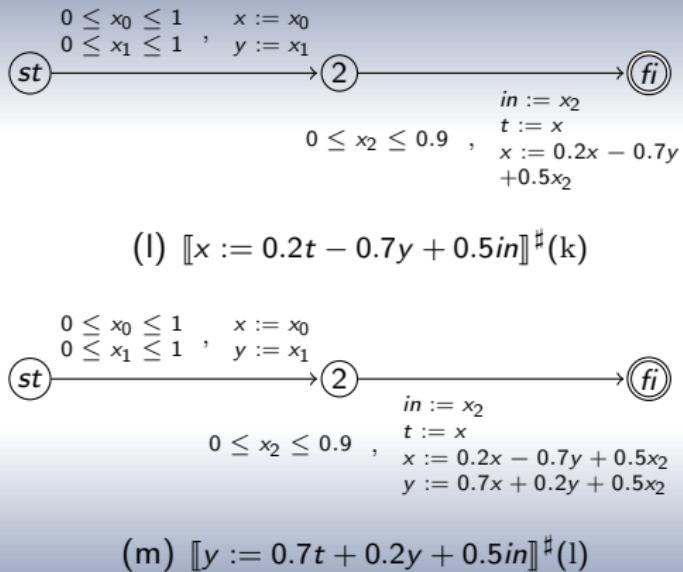


$$(l) \quad [x := 0.2t - 0.7y + 0.5in]^{\sharp}(k)$$



# Example

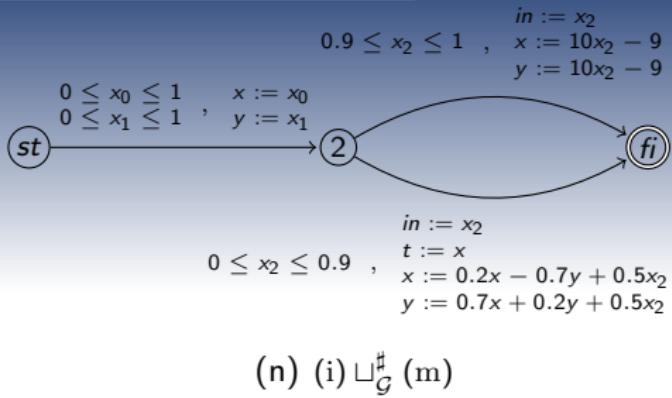
```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else(j)
        t := x;(k)
        x := 0.2×t - 0.7×y + 0.5×in;(l)
        y := 0.7×t + 0.2×y + 0.5×in(m)
    fi
od
```





## Example

```
(a) x := ?(0, 1);(b) y := ?(0, 1);(c)
while(d) - 1 ≤ 0(e) do
    in := ?(0, 1);(f)
    if 0.9 - in ≤ 0(g) then
        x := 10×in - 9;(h)
        y := 10×in - 9(i)
    else(j)
        t := x;(k)
        x := 0.2×t - 0.7×y + 0.5×in;(l)
        y := 0.7×t + 0.2×y + 0.5×in(m)
fi(n)
od
```





## Example

(a)  $x := ?(0, 1)$ ; (b)  $y := ?(0, 1)$ ; (c)

**while**  $(d), (o) - 1 \leq 0$  **do**

in :=  $?(0, 1)$ ; (f)

**if**  $0.9 - \text{in} \leq 0$  **then**

$x := 10 \times \text{in} - 9$ ; (h)

$y := 10 \times \text{in} - 9$ ; (i)

**else** **(j)**

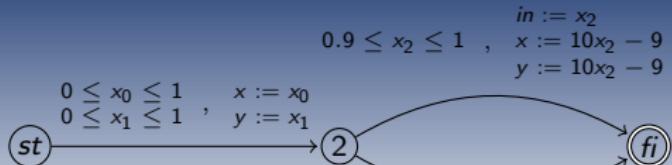
$t := x$ ; (k)

$x := 0.2 \times t - 0.7 \times y + 0.5 \times \text{in}$ ; (l)

$y := 0.7 \times t + 0.2 \times y + 0.5 \times \text{in}$ ; (m)

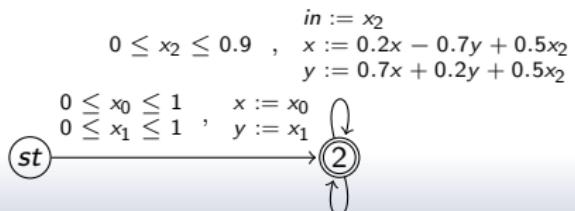
**fi** **(n)**

**od**



$0 \leq x_2 \leq 0.9$ ,  
 $\text{in} := x_2$ ,  
 $t := x$ ,  
 $x := 0.2x - 0.7y + 0.5x_2$ ,  
 $y := 0.7x + 0.2y + 0.5x_2$

(n) (i)  $\sqcup_G^\sharp$  (m)



$0.9 \leq x_2 \leq 1$ ,  
 $\text{in} := x_2$ ,  
 $t := x$ ,  
 $x := 10x_2 - 9$ ,  
 $y := 10x_2 - 9$

(o) (d)  $\nabla_G$  (n)



## Example

$(a) x := ?(0, 1); (b) y := ?(0, 1); (c)$

**while** $_{(d), (o)} - 1 \leq 0_{(e)}$  **do**

in :=  $?(0, 1); (f)$

**if**  $0.9 - \text{in} \leq 0_{(g)}$  **then**

$x := 10 \times \text{in} - 9; (h)$

$y := 10 \times \text{in} - 9; (i)$

**else** $_{(j)}$

$t := x; (k)$

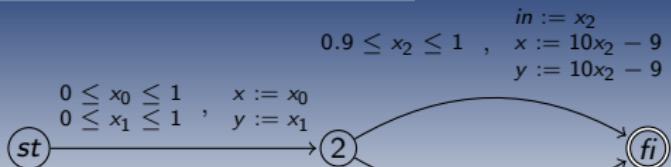
$x := 0.2 \times t - 0.7 \times y + 0.5 \times \text{in}; (l)$

$y := 0.7 \times t + 0.2 \times y + 0.5 \times \text{in}_{(m)}$

**fi** $_{(n)}$

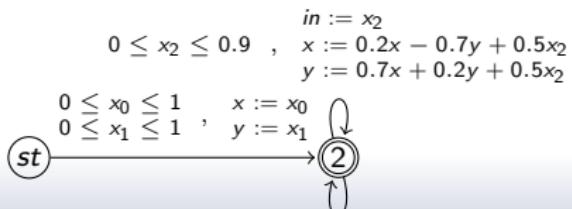
**od**

A second iteration  
proves that (o) is a **fixpoint**.



0  $\leq x_2 \leq 0.9$ ,  
 $\begin{aligned} in &:= x_2 \\ t &:= x \\ x &:= 0.2x - 0.7y + 0.5x_2 \\ y &:= 0.7x + 0.2y + 0.5x_2 \end{aligned}$

(n) (i)  $\sqcup_G^\sharp$  (m)



0.9  $\leq x_2 \leq 1$ ,  
 $\begin{aligned} in &:= x_2 \\ t &:= x \\ x &:= 10x_2 - 9 \\ y &:= 10x_2 - 9 \end{aligned}$

(o) (d)  $\nabla_G$  (n)

## ① State of the Art – a Policy Iteration Primer

## ② An Abstract Control Flow Graph Domain

## ③ Application to Quadratic Invariants on Linear Systems

## ④ Experimental Results

## ⑤ Conclusion and Future Work



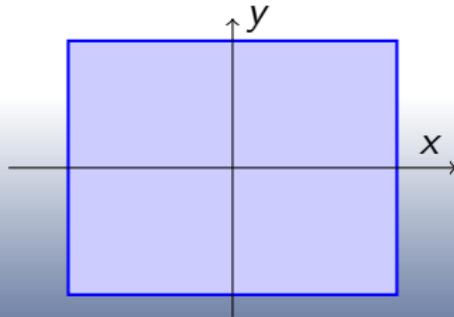
## Quadratic invariants

- **Linear invariants** commonly used in static analysis are not well suited for linear reactive systems:
  - at best costly;
  - at worst no result.



## Quadratic invariants

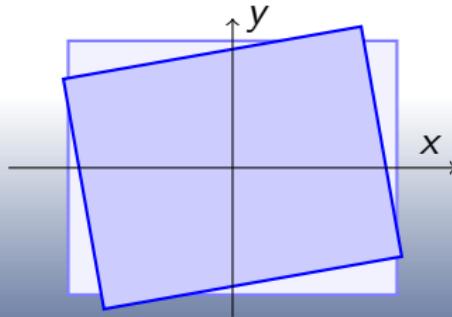
- **Linear invariants** commonly used in static analysis are not well suited for linear reactive systems:
  - at best costly;
  - at worst no result.





## Quadratic invariants

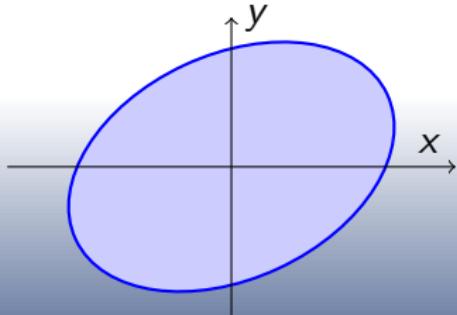
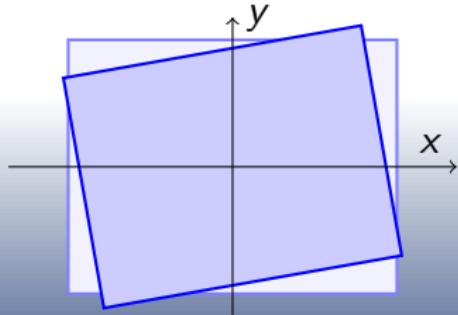
- **Linear invariants** commonly used in static analysis are not well suited for linear reactive systems:
  - at best costly;
  - at worst no result.





## Quadratic invariants

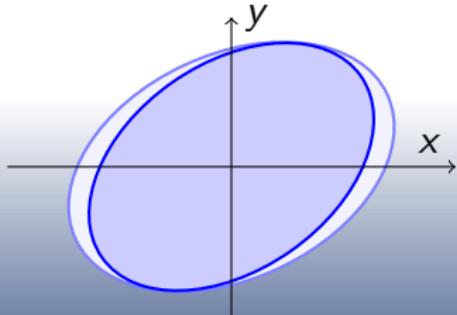
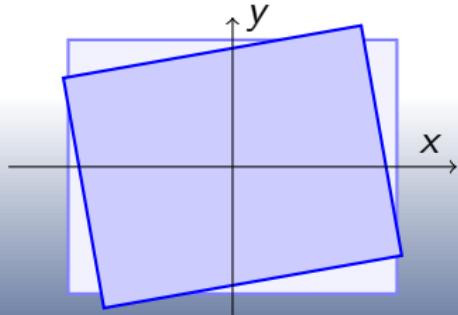
- **Linear invariants** commonly used in static analysis are not well suited for linear reactive systems:
  - at best costly;
  - at worst no result.
- Control theorists know for long that such systems are stable iff they admit **quadratic invariants**.





## Quadratic invariants

- **Linear invariants** commonly used in static analysis are not well suited for linear reactive systems:
  - at best costly;
  - at worst no result.
- Control theorists know for long that such systems are stable iff they admit **quadratic invariants**.





# Choosing Appropriate Templates

- from the control flow graph, we can extract subset of variables  $x$  and  $u$  and matrices  $A$  and  $B$  such that

$$x_{k+1} = Ax_k + Bu_k$$

- then we can extract a quadratic template by solving a Lyapunov equation
- we finally add templates  $y$  and  $-y$  for each variable  $y$  in the program

- 1 State of the Art – a Policy Iteration Primer
- 2 An Abstract Control Flow Graph Domain
- 3 Application to Quadratic Invariants on Linear Systems
- 4 Experimental Results

- 5 Conclusion and Future Work



## Implementation

- A new domain in our Lustre analyzer
- 4kloc of OCaml and 0.5kloc of C code
- use CSDP library for semidefinite programming
- soundness check of the result by Cholesky decompositions
- available under a GPL license at  
<http://cavale.enseeiht.fr/policy2013/>



## Benchmarks (1/2)

	$n$	#policies	total	templates	iterations	check
Ex. 1	3	5	0.84	0.34	0.44	0.02
	4	5	1.81	0.36	1.30	0.07
	3	5	0.92	0.34	0.52	0.03
Ex. 2	5	5	2.76	0.58	2.04	0.05
	6	6	13.36	0.51	12.45	0.18
	5	5	2.78	0.51	2.08	0.06
Ex. 3 Discretized lead-lag controller	3	5	1.08	0.32	0.71	$\perp$ (0.01)
	4	5	3.98	0.31	3.57	$\perp$ (0.03)
	3	5	1.12	0.31	0.75	$\perp$ (0.02)
Ex. 4 Linear quadratic gaussian regulator	4	4	1.65	0.33	1.21	0.06
	5	4	7.42	0.33	6.94	$\perp$ (0.02)
	4	5	1.49	0.34	1.00	0.06

All times are in second.



## Benchmarks (2/2)

	$n$	#policies	total	templates	iterations	check
Ex. 5 Observer based controller for a coupled mass system	6	4	3.16	0.59	2.31	0.11
	7	6	30.52	0.58	29.25	0.38
	6	5	4.50	0.60	3.60	0.13
Ex. 6 Butterworth low-pass filter	6	4	4.42	0.87	3.26	0.14
	7	5	24.80	0.88	23.10	0.47
	6	5	5.03	0.87	3.79	0.16
Ex. 7 Dampened oscillator	2	5	0.41	0.19	0.18	0.02
	3	5	1.07	0.20	0.72	0.07
	2	5	0.51	0.21	0.24	0.03
Ex. 8 Harmonic oscillator	2	5	0.41	0.19	0.17	0.02
	3	5	1.05	0.19	0.71	0.07
	2	5	0.48	0.19	0.23	0.03

All times are in second.

- 1 State of the Art – a Policy Iteration Primer
- 2 An Abstract Control Flow Graph Domain
- 3 Application to Quadratic Invariants on Linear Systems
- 4 Experimental Results
- 5 Conclusion and Future Work



## Conclusion and Future Work

- it works
- we should try min-strategy iteration  
which could bring performance improvements
- we should consider floating point semantic of programs



## Questions

Thank you for your attention!

?