

# Abstract Interpretation

## Semantics and applications to verification

Xavier RIVAL

École Normale Supérieure

April 18th, 2024

# Program of this lecture

## Towards a realistic abstract interpreter

Last class ended with a brief overview of a **simplistic static analyzer**

**Today:**

- **more general soundness proof:**  
using  $\gamma$ , and requiring no monotonicity in the abstract level
- **more general abstract domain:**  
signs is good for introduction only, we want to see constants, intervals...
- **extended language** with **expressions** and **conditions**  
i.e., not only three address arithmetic
- **more general abstract iteration technique:**  
convergence guaranteed even with **infinite height domain**

# Outline

- 1 Another Soundness Relation
- 2 Revisiting Abstract Iteration
- 3 Conclusion

# About soundness relations

## Several formalisms available:

- **abstraction function**  $\alpha : C \rightarrow A$ , returns the **best** approximation
- **concretization function**  $\gamma : A \rightarrow C$ , returns the meaning of an abstract element
- **Galois connection**  $(C, \subseteq) \xleftrightarrow[\alpha]{\gamma} (A, \sqsubseteq)$

## Limitations of our previous abstract interpreter:

- uses the best abstraction function  $\alpha$  all the time
- tries to establish equality  $\llbracket P \rrbracket^\# \circ \alpha = \alpha \circ \llbracket P \rrbracket$  but fails...  
indeed, some operators may only compute an over-approximation
- proves  $\alpha \circ \llbracket P \rrbracket \sqsubseteq \llbracket P \rrbracket^\# \circ \alpha$   
**at the cost of proving monotonicity of  $\llbracket P \rrbracket^\#$**

## Alternate approach

**Use  $\gamma$  only and prove  $\llbracket P \rrbracket \circ \gamma \subseteq \gamma \circ \llbracket P \rrbracket^\#$**

# Comparing soundness frameworks

We have seen **several ways to express soundness**:

- ❶  $\alpha(\llbracket P \rrbracket) = \llbracket P \rrbracket^\sharp$
- ❷  $\alpha(\llbracket P \rrbracket) \sqsubseteq \llbracket P \rrbracket^\sharp$
- ❸  $\llbracket P \rrbracket \subseteq \gamma(\llbracket P \rrbracket^\sharp)$
- ❹  $\llbracket P \rrbracket \vdash \llbracket P \rrbracket^\sharp$

**Some are stronger than others:**

- the first is the strongest (it implies the others when applicable)
- the fourth is the weakest
- the second and third are equivalent in a Galois connection setup

**Some are more general:**

- the first two require an  $\alpha$ , the third a  $\gamma$
- the fourth requires very little: it does not even require  $\alpha$  or  $\gamma$  to exist!

**The choice of the framework to use is a balance in general...**

# A language with expressions

We now consider the denotational semantics of our **imperative language**:

- **variables**  $\mathbb{X}$ : finite, predefined set of variables
- **values**  $\mathbb{V}$ :  $\mathbb{V}_{\text{int}} \cup \mathbb{V}_{\text{float}} \cup \dots$
- **expressions** are allowed (not just three address instructions)
- **conditions** are **simplified** compared to initial language

## Syntax

$e ::=$	$v \ (v \in \mathbb{V}) \mid x \ (x \in \mathbb{X}) \mid e + e \mid e * e \mid \dots$	expressions
$c ::=$	$x < v \mid x = v \mid \dots$	basic conditions
$P ::=$	$x := e$	assignment
	$\mid \text{input}(x)$	non deterministic value input
	$\mid \text{if}(c) \ P \ \text{else} \ P$	condition
	$\mid \text{while}(c) \ P$	loop
	$\mid P; P$	block, program( $\mathbb{P}$ )

# Semantics of expressions and conditions (refresher)

We have defined a few lectures ago:

- a **semantics for expressions**, defined **by induction over the syntax**:

$$\begin{aligned}
 \llbracket e \rrbracket : \mathbb{M} &\longrightarrow \mathbb{V} \uplus \{\Omega\} \\
 \llbracket v \rrbracket(m) &= v \\
 \llbracket x \rrbracket(m) &= m(x) \\
 \llbracket e_0 + e_1 \rrbracket(m) &= \llbracket e_0 \rrbracket(m) \pm \llbracket e_1 \rrbracket(m) \\
 \llbracket e_0 / e_1 \rrbracket(m) &= \begin{cases} \Omega & \text{if } \llbracket e_1 \rrbracket(m) = 0 \\ \llbracket e_0 \rrbracket(m) \_ \llbracket e_1 \rrbracket(m) & \text{otherwise} \end{cases}
 \end{aligned}$$

- a **semantics for conditions**, following the same principle:

$$\llbracket c \rrbracket : \mathbb{M} \longrightarrow \mathbb{V}_{\text{bool}} \uplus \{\Omega\}$$

# Semantics of statements

We have also defined:

## Denotational semantics of programs

We use the **denotational semantics**  $\llbracket P \rrbracket_{\mathcal{D}} : \mathcal{P}(\mathbb{M}) \longrightarrow \mathcal{P}(\mathbb{M})$  by:

$$\llbracket x := e \rrbracket_{\mathcal{D}}(\mathcal{M}) = \{m[x \leftarrow \llbracket e \rrbracket(m)] \mid m \in \mathcal{M} \wedge \llbracket e \rrbracket(m) \neq \Omega\}$$

$$\llbracket \text{input}(x) \rrbracket_{\mathcal{D}}(\mathcal{M}) = \{m[x \leftarrow v] \mid v \in \mathbb{V} \wedge m \in \mathcal{M}\}$$

$$\begin{aligned} \llbracket \text{if}(c) P_0 \text{ else } P_1 \rrbracket_{\mathcal{D}}(\mathcal{M}) &= \llbracket P_0 \rrbracket_{\mathcal{D}}(\{m \in \mathcal{M} \mid \llbracket c \rrbracket(m) = \text{TRUE}\}) \\ &\quad \cup \llbracket P_1 \rrbracket_{\mathcal{D}}(\{m \in \mathcal{M} \mid \llbracket c \rrbracket(m) = \text{FALSE}\}) \end{aligned}$$

$$\begin{aligned} \llbracket \text{while}(c) P \rrbracket_{\mathcal{D}}(\mathcal{M}) &= \{m \in \text{lfp } F_{\mathcal{D}} \mid \llbracket c \rrbracket(m) = \text{FALSE}\} \\ \text{where } F_{\mathcal{D}} : \mathcal{M}' &\longmapsto \mathcal{M} \cup \llbracket P \rrbracket_{\mathcal{D}}(\{m \in \mathcal{M}' \mid \llbracket c \rrbracket(m) = \text{TRUE}\}) \end{aligned}$$

$$\llbracket P_0; P_1 \rrbracket_{\mathcal{D}}(\mathcal{M}) = \llbracket P_1 \rrbracket_{\mathcal{D}} \circ \llbracket P_0 \rrbracket_{\mathcal{D}}(\mathcal{M})$$

- As before, we seek for **an abstract interpretation of  $\llbracket P \rrbracket_{\mathcal{D}}$**
- We first need to set up **the abstraction relation**



# Example

## Example program:

```

input(x);
x = 3 - x;
if(x ≥ 1){
    y = 8 - 2 * x;
}else{
    y = 1;
}

```

**Analysis with the lattice of signs:**  $x \mapsto \top; y \mapsto \top$

Can we use another abstract domain instead, such as **intervals** ?

- intuitively,  $x \leq 3$
- thus, either  $1 \leq x \leq 3$  and  $2 \leq y \leq 6$  or  $x < 1$  and  $y = 1$
- in any case,  $1 \leq y \leq 6$

# Galois-connection based non-relational abstraction

We compose two abstractions:

- **non relational abstraction:** the values a variable may take is abstracted separately from the other variables
- **parameter value abstraction:** an **abstract value** describes a set of concrete values (not necessarily the lattice of sign anymore) defined by

$$(\mathcal{P}(\mathbb{V}), \subseteq) \xleftrightarrow[\alpha_V]{\gamma_V} (D_V^\#, \sqsubseteq)$$

Definitions are quite similar:

## Abstraction

- **concrete domain:**  $(\mathcal{P}(\mathbb{X} \rightarrow \mathbb{V}), \subseteq) = (\mathcal{P}(\mathbb{M}), \subseteq)$
- **abstract domain:**  $(D^\#, \sqsubseteq)$  ( $D^\# = \mathbb{X} \rightarrow D_V^\#$  and  $\sqsubseteq$  is pointwise)
- **Galois connection**  $(\mathcal{P}(\mathbb{M}), \subseteq) \xleftrightarrow[\alpha]{\gamma} (D^\#, \sqsubseteq)$ , defined by

$$\alpha : \mathcal{M} \mapsto \lambda(\mathbf{x} \in \mathbb{X}) \cdot \alpha_V(\{m(\mathbf{x}) \mid m \in \mathcal{M}\})$$

$$\gamma : M^\# \mapsto \{m : \mathbb{X} \rightarrow \mathbb{V} \mid \forall \mathbf{x}, m(\mathbf{x}) \in \gamma_V(M^\#(\mathbf{x}))\}$$

## Or a more general abstraction, using only $\gamma$

As before, we compose two abstractions:

- **non relational abstraction:** the values a variable may take is abstracted separately from the other variables, as before, but we consider only concretization
- **parameter value abstraction:** an **abstract value** describes a set of concrete values defined by a monotone concretization function  
 $\gamma_V : (D_V^\sharp, \sqsubseteq) \rightarrow (\mathcal{P}(\mathbb{V}), \subseteq)$

### Abstraction relation based on concretization only

- **concrete domain:**  $(\mathcal{P}(\mathbb{X} \rightarrow \mathbb{V}), \subseteq)$
- **abstract domain:**  $(D^\sharp, \sqsubseteq)$  ( $D^\sharp = \mathbb{X} \rightarrow D_V^\sharp$  and  $\sqsubseteq$  is pointwise)
- **concretization function**  $\gamma : (D^\sharp, \sqsubseteq) \rightarrow (\mathcal{P}(\mathbb{M}), \subseteq)$ , defined by  

$$\gamma : M^\sharp \mapsto \{m : \mathbb{X} \rightarrow \mathbb{V} \mid \forall \mathbf{x}, m(\mathbf{x}) \in \gamma_V(M^\sharp(\mathbf{x}))\}$$

$\Rightarrow$  likewise, our proof will use only  $\gamma$  though it works even when  $\alpha$  is defined

# Abstract semantics of sequences (revised)

We search **for an abstract semantics**  $\llbracket P \rrbracket^\# : D^\# \rightarrow D^\#$  such that:

$$\llbracket P \rrbracket \circ \gamma \subseteq \gamma \circ \llbracket P \rrbracket^\#$$

We still aim for a **proof by induction over the syntax of programs**

**Sequences / composition** forced us to require **monotonicity** last time:

- we assume  $\llbracket P_0 \rrbracket \circ \gamma \subseteq \gamma \circ \llbracket P_0 \rrbracket^\#$
- we assume  $\llbracket P_1 \rrbracket \circ \gamma \subseteq \gamma \circ \llbracket P_1 \rrbracket^\#$
- since  $\llbracket P_0; P_1 \rrbracket = \llbracket P_1 \rrbracket \circ \llbracket P_0 \rrbracket$ , we search for something similar in the abstract level

$$\begin{aligned} \llbracket P_1 \rrbracket \circ \llbracket P_0 \rrbracket \circ \gamma &\subseteq \llbracket P_1 \rrbracket \circ \gamma \circ \llbracket P_0 \rrbracket^\# && \text{(by induction)} \\ &\subseteq \gamma \circ \llbracket P_1 \rrbracket^\# \circ \llbracket P_0 \rrbracket^\# && \text{(by induction)} \end{aligned}$$

**No more requirement that  $\llbracket P \rrbracket^\#$  be monotone (much better!)**

# Abstract semantics of expressions

## Analysis of an expression

- semantics of expressions  $\llbracket e \rrbracket : \mathbb{M} \longrightarrow \mathbb{V} \uplus \{\Omega\}$
- thus, the abstract semantics **should evaluate it into an abstract value**:

$$\llbracket e \rrbracket^\# : D^\# \longrightarrow D_V^\#$$

Since we use the concrete semantics as a guide, we need:

- **abstraction for constants:**

i.e., a function  $\phi_V : \mathbb{V} \rightarrow D_V^\#$  such that  $\forall v \in \mathbb{V}, v \in \gamma_V(\phi_V(v))$

note: if  $\alpha_V$  exists, then we may take  $v \mapsto \alpha_V(\{v\})$  note: if it is too hard to compute, we may take something coarser

- **abstract operators:**

i.e., for each binary operator  $\oplus$ , an abstract operator  $\oplus^\#$  such that:

$$\forall v_0^\#, v_1^\# \in D_V^\#, \{v_0 \oplus v_1 \mid \forall i, v_i \in \gamma_V(v_i^\#)\} \subseteq \gamma_V(v_0^\# \oplus^\# v_1^\#)$$

# Abstract semantics of expressions

## Analysis of expressions: definition

We define  $\llbracket e \rrbracket^\# : D^\# \longrightarrow D_V^\#$  by:

$$\begin{aligned}\llbracket v \rrbracket^\#(M^\#) &= \phi_V(v) \\ \llbracket x \rrbracket^\#(M^\#) &= M^\#(x) \\ \llbracket e_0 \oplus e_1 \rrbracket^\#(M^\#) &= \llbracket e_0 \rrbracket^\#(M^\#) \oplus^\# \llbracket e_1 \rrbracket^\#(M^\#)\end{aligned}$$

## Analysis of expressions: soundness

For all expression  $e$  and for all abstract memory state  $M^\# \in D^\#$ , we have:

$$\forall m \in \gamma(M^\#), \llbracket e \rrbracket(m) \text{ returns no error} \implies \llbracket e \rrbracket(m) \in \gamma_V(\llbracket e \rrbracket^\#(M^\#))$$

### Proof:

- basic **induction over the syntax**
- relies on the soundness of each operation

# Analysis of an assignment

We now rely on the abstract semantics of expressions:

$$\llbracket x = e \rrbracket^\#(M^\#) = M^\#[x \leftarrow \llbracket e \rrbracket^\#(M^\#)]$$

- soundness proof is very similar
- but now, is given in terms of  $\gamma$

**Example**, based on the **interval abstract domain**; analysis of

$x = 3 * y + z * x$ .

If  $M^\# : x \mapsto [-2, 3], y \mapsto [0, 1]$ , and  $z \mapsto [-1, 4]$  then:

$$\begin{aligned} \llbracket x = 3 * y + z * x \rrbracket^\#(M^\#)(x) &= 3 * [0, 1] + [-1, 4] * [-2, 3] \\ &= [0, 3] + [-8, 12] \\ &= [-8, 15] \end{aligned}$$

# Abstract semantics of conditions

## Analysis of a condition

- the semantics  $\llbracket c \rrbracket : \mathbb{M} \longrightarrow \mathbb{V}_{\text{bool}}$  of a condition evaluates it into a boolean value (or an error)
- but** the semantics relies on its functional inverse:  
e.g.,  $\{m \in \mathbb{M} \mid \llbracket c \rrbracket(m) = \text{TRUE}\}$  or  $\{m \in \mathbb{M} \mid \llbracket c \rrbracket(m) = \text{FALSE}\}$
- thus, the abstract semantics **should tell which memories satisfy a condition**:

$$\begin{aligned} \llbracket c \rrbracket^\# : \mathbb{V}_{\text{bool}} \times D^\# &\longrightarrow D^\# \\ \forall b \in \mathbb{V}_{\text{bool}}, \forall m \in \gamma(M^\#), \llbracket c \rrbracket(m) = b &\implies m \in \gamma(\llbracket c \rrbracket^\#(b, M^\#)) \end{aligned}$$

- we assume that the abstract domain provides such a function  
 $\llbracket c \rrbracket^\# : \mathbb{V}_{\text{bool}} \times D^\# \longrightarrow D^\#$
- this is also called a **backward abstract semantics**  
intuitively: it goes from outputs to arguments
- we will implement some when considering specific abstract domains



# Analysis of a condition statement

## Abstraction of concrete union:

- we assume a **sound abstract union operation**  $\text{join}_V^\sharp$  over the value abstract domain:

$$\forall v_0^\sharp, v_1^\sharp, \gamma_V(v_0^\sharp) \cup \gamma_V(v_1^\sharp) \subseteq \gamma_V(\text{join}_V^\sharp(v_0^\sharp, v_1^\sharp))$$

it may be  $\sqcup_V$  if it exists, but  $\text{join}_V^\sharp$  may also over-approximate it

- we let  $\text{join}^\sharp$  be the pointwise extension of  $\text{join}_V^\sharp$
- it is also sound:  $\forall M_0^\sharp, M_1^\sharp, \gamma(M_0^\sharp) \cup \gamma(M_1^\sharp) \subseteq \gamma(\text{join}^\sharp(M_0^\sharp, M_1^\sharp))$

We derive:

$$\llbracket \text{if}(c) P_0 \text{ else } P_1 \rrbracket^\sharp(M^\sharp) = \text{join}^\sharp(\llbracket P_0 \rrbracket^\sharp(\llbracket c \rrbracket^\sharp(\text{TRUE}, M^\sharp)), \llbracket P_1 \rrbracket^\sharp(\llbracket c \rrbracket^\sharp(\text{FALSE}, M^\sharp)))$$

## Proof of soundness:

- similar as in the previous course
- relies on the soundness of  $\llbracket c \rrbracket^\sharp$ ,  $\llbracket P_0 \rrbracket^\sharp$ ,  $\llbracket P_1 \rrbracket^\sharp$  and  $\text{join}^\sharp$

# Example condition statement analysis

We can now show the interval analysis of the example program:

```

input(x);
     $x \mapsto [0, +\infty[, y \mapsto ] - \infty, +\infty[$ 
 $x = 3 - x;$ 
     $x \mapsto ] - \infty, 3], y \mapsto ] - \infty, +\infty[$ 
if( $x \geq 1$ ){
     $x \mapsto [1, 3], y \mapsto ] - \infty, +\infty[$ 
     $y = 8 - 2 * x;$ 
     $x \mapsto [1, 3], y \mapsto [2, 6]$ 
}else{
     $x \mapsto ] - \infty, 1], y \mapsto ] - \infty, +\infty[$ 
     $y = 1;$ 
     $x \mapsto ] - \infty, 1], y \mapsto [1, 1]$ 
}
     $x \mapsto ] - \infty, 3], y \mapsto [1, 6]$ 

```

# Another example

```

input(x);
       $x \mapsto ] - \infty, +\infty[, y \mapsto ] - \infty, +\infty[$ 
x = y * y + 1;
       $x \mapsto [1, +\infty[, y \mapsto ] - \infty, +\infty[$ 
if(x ≤ 0){
    ?
    x = -1;
    ?
} else {
    ?
}
    ?

```

## Questions:

- fill the “?”...
- what is happening ? how to make the analysis precise ?

# Reduction in the non relational abstraction

Consider the following two abstract elements:

$$x \in \perp, y \in [1, 10]$$

and

$$x \in \perp, y \in \perp$$

- their concretisations are equal to  $\emptyset$
- in fact, applying  $\gamma \circ \alpha$  to the former returns the latter

## Reduction

The **optimal reduction function** is defined by  $\gamma \circ \alpha$  and returns an optimal abstract element, with the same concretisation.

While optimal reduction is not computable in general, it is in this case.

# Fixpoint approximation

Again, quite similar to the previous course:

- statement  $\text{while}(c) P$ , with abstract pre-condition  $M^\sharp$
- we assume sound  $\llbracket c \rrbracket^\sharp$  and  $\llbracket P \rrbracket^\sharp$  are defined

## Fixpoint approximation (instead of fixpoint transfer)

We assume  $(C, \subseteq)$  and  $(A, \sqsubseteq)$  are complete lattices, with a concretization function  $\gamma : (A, \sqsubseteq) \rightarrow (C, \subseteq)$ , two functions  $f : C \rightarrow C$  and  $f^\sharp : A \rightarrow A$ , and two elements  $c_0 \in C, a_0 \in A$  such that:

- $f$  is continuous
- $f \circ \gamma \subseteq \gamma \circ f^\sharp$
- $c_0 \subseteq \gamma(a_0)$

We let  $a_\infty = \sqcup \{(f^\sharp)^n(a_0) \mid n \in \mathbb{N}\}$ . Then:

- $f$  has a **least-fixpoint** (by Kleene's fixpoint theorem)
- $\text{lfp}_{c_0} f \subseteq \gamma(a_\infty)$

# Fixpoint approximation: proof

## Existence of the concrete fixpoint:

First, we remark that  $\text{lfp}_{c_0} f$  exists, following Kleene's fixpoint theorem.

Moreover:

$$\text{lfp}_{c_0} f = \bigcup_{n \in \mathbb{N}} f^n(c_0)$$

## Approximation of the fixpoint:

First,  $a_\infty = \sqcup \{(f^\sharp)^n(a_0) \mid n \in \mathbb{N}\}$  exists since we assume  $A$  is a complete lattice (note that we have not addressed how to compute it quite yet!). We prove by induction over  $n$  that  $f^n(c_0) \subseteq \gamma((f^\sharp)^n(a_0))$ :

- since we assumed  $c_0 \subseteq \gamma(a_0)$ , the property holds at rank 0;
- if we assume  $f^n(c_0) \subseteq \gamma((f^\sharp)^n(a_0))$ , then  $f(f^n(c_0)) \subseteq f(\gamma((f^\sharp)^n(a_0)))$  since  $f$  is monotone, which implies  $f^{n+1}(c_0) \subseteq \gamma((f^\sharp)^{n+1}(a_0))$  since  $f \circ \gamma \subseteq \gamma \circ f^\sharp$ .

The fixpoint approximation property follows from property of least upper-bounds and from the monotonicity of  $\gamma$ .

# Analysis of a loop

Again, quite similar to the previous course:

- statement  $\text{while}(c) P$ , with abstract pre-condition  $M^\sharp$
- we assume  $\llbracket c \rrbracket^\sharp$  and  $\llbracket P \rrbracket^\sharp$  sound abstract semantics for the condition and the loop body
- we assume the abstract domain is a finite height lattice  
this ensures that we can compute  $a_\infty = \sqcup \{(f^\sharp)^n(a_0) \mid n \in \mathbb{N}\}$   
(exercise), but **intervals do not satisfy this condition**

## Computation of abstract iterates:

$$\llbracket \text{while}(c) P \rrbracket^\sharp(M^\sharp) = \llbracket c \rrbracket^\sharp(\text{FALSE}, M_n^\sharp)$$

$$\text{where } \begin{cases} I_0^\sharp &= M^\sharp \\ I_{k+1}^\sharp &= \llbracket P \rrbracket^\sharp(\llbracket c \rrbracket^\sharp(\text{TRUE}, I_k^\sharp)) \end{cases}$$

$$\text{and } M_{n+1}^\sharp = M_n^\sharp$$

$$\begin{aligned} M_0^\sharp &= M^\sharp \\ M_{k+1}^\sharp &= \text{join}^\sharp(M_k^\sharp, I_{k+1}^\sharp) \end{aligned}$$

# Static analysis

We can now summarize the definition of our static analysis:

## Definition

$$\begin{aligned}
 \llbracket P_0; P_1 \rrbracket^\#(M^\#) &= \llbracket P_1 \rrbracket^\# \circ \llbracket P_0 \rrbracket^\#(M^\#) \\
 \llbracket x = e \rrbracket^\#(M^\#) &= M^\#[x \leftarrow \llbracket e \rrbracket^\#(M^\#)] \\
 \llbracket \text{input}() \rrbracket^\#(M^\#) &= M^\#[x \leftarrow \top] \\
 \llbracket \text{if}(c) P_0 \text{ else } P_1 \rrbracket^\#(M^\#) &= \text{join}^\#(\llbracket P_0 \rrbracket^\#(\llbracket c \rrbracket^\#(\text{TRUE}, M^\#)), \\
 &\quad \llbracket P_1 \rrbracket^\#(\llbracket c \rrbracket^\#(\text{FALSE}, M^\#))) \\
 \llbracket \text{while}(c) P \rrbracket^\#(M^\#) &= \lim_n M_n^\# \\
 \text{where } I_0^\# &= M_0^\# = M^\#, I_{k+1}^\# = \llbracket P \rrbracket^\#(\llbracket c \rrbracket^\#(\text{TRUE}, I_k^\#)) \\
 \text{and } M_{k+1}^\# &= \text{join}^\#(M_k^\#, I_{k+1}^\#)
 \end{aligned}$$

And, by induction over the syntax, we can prove:

## Soundness

For all program  $P$ ,  $\forall M^\# \in D^\#$ ,  $\llbracket P \rrbracket \circ \gamma(M^\#) \subseteq \gamma \circ \llbracket P \rrbracket^\#(M^\#)$



# Outline

- 1 Another Soundness Relation
- 2 Revisiting Abstract Iteration
- 3 Conclusion

# Limitations related to abstract iteration

## We need a finite height lattice:

- otherwise the computation of  $\text{lfp } F^\#$  **may not converge** as was the case when we discussed **WLP calculus**
- **consequence 1**: so far, the **abstract domain of intervals** is out...
- **consequence 2**: if the number of variables **is not fixed** or **bounded**, we cannot prove convergence at this point

Even when the abstract domain  $D_V^\#$  is of finite height, this height **may be huge**: then abstract computations are very costly!

**We now need a more general abstract iteration technique**

**Intuition** from **search for an unknown inductive property**:

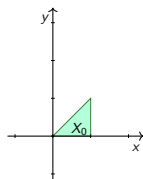
- 1 look at the base case and following cases
- 2 try to **generalize them**

# Widening iteration: search for inductive abstract properties

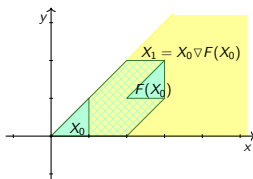
## Computing invariants about infinite executions with widening $\nabla$

- **Widening**  $\nabla$  over-approximates  $\cup$ : **soundness guarantee**
- **Widening**  $\nabla$  guarantees the **termination of the analyses**
- Typical choice of  $\nabla$ : **remove unstable constraints**

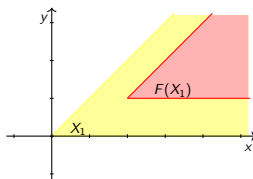
**Example:** iteration of the translation  $(2, 1)$ , with **octagonal polyhedra** (i.e., convex polyhedra the axes of which are either at a  $0^\circ$  or  $45^\circ$  angle)



initial



iteration 1

iteration 2: **stable !**

- Initially: **3 constraints**
- After one iteration: **2 constraints**, then stable

# Widening operator

## Widening operator: Definition

A **widening operator** over an abstract domain  $D^\sharp$  is a binary operator  $\nabla$  such that:

- $\forall M_0^\sharp, M_1^\sharp, \gamma(M_0^\sharp) \cup \gamma(M_1^\sharp) \subseteq \gamma(M_0^\sharp \nabla M_1^\sharp)$
- if  $(N_k^\sharp)_{k \in \mathbb{N}}$  is a sequence of elements of  $D^\sharp$  the **sequence**  $(M_k^\sharp)_{k \in \mathbb{N}}$  defined below is **stationary**:

$$\begin{aligned} M_0^\sharp &= N_0^\sharp \\ M_{k+1}^\sharp &= M_k^\sharp \nabla N_{k+1}^\sharp \end{aligned}$$

- **Intuition:**
  - point 1 expresses **over-approximation** of concrete union
  - point 2 enforces **termination**
- **Alternate definitions** exist:  
e.g., using  $\sqsubseteq$  instead of  $\subseteq$  over concretizations

# Widening operator in a finite height domain

## Theorem

We assume that  $(D^\sharp, \sqsubseteq)$  is a **finite height domain** and that  $\sqcup$  is the **least upper bound over  $D^\sharp$** .

Then  $\sqcup$  **defines a widening over  $D^\sharp$** .

## Proof:

- ① since  $M_0^\sharp \sqsubseteq M_0^\sharp \sqcup M_1^\sharp$ , we have  $\gamma(M_0^\sharp) \sqsubseteq \gamma(M_0^\sharp \sqcup M_1^\sharp)$
- ② a sequence of iterates  $(M_k^\sharp)_{k \in \mathbb{N}}$  is an **increasing chain**, so if every increasing chain is finite, it will eventually stabilize

## Applications:

- obvious widening operators for the lattices of constants, signs...
- abstract iteration algorithms are also the same

# A widening operator in an infinite height domain

We consider the **value lattice of semi intervals with left bound 0**:

- $D_V^\# = \{\perp\} \uplus \mathbb{Z}_+ \uplus \{+\infty\}$ ;  $\gamma_V(v) = \{0, 1, \dots, v\}$
- $\forall v^\#, \perp \sqsubseteq v^\#$  and if  $v_0^\# \leq v_1^\#$ , then  $v_0^\# \sqsubseteq v_1^\#$

We define **the widening operator** below:

## Widening operator

$$\begin{aligned} \perp \nabla v^\# &= v^\# \\ v^\# \nabla \perp &= v^\# \\ v_0^\# \nabla v_1^\# &= \begin{cases} v_0^\# & \text{if } v_0^\# \geq v_1^\# \\ +\infty & \text{if } v_0^\# < v_1^\# \end{cases} \end{aligned}$$

**Examples:**  $[0, 8] \nabla [0, 6] = [0, 8]$        $[0, 8] \nabla [0, 9] = [0, +\infty[$

## Widening for intervals

Exercise: generalize this definition for both bounds

# Fixpoint approximation using a widening operator

## Theorem: widening based fixpoint approximation

We assume  $(C, \subseteq)$  is a complete lattice and that  $(A, \sqsubseteq)$  is an abstract domain with a concretization function  $\gamma : A \rightarrow C$  and a widening operator  $\nabla$ . Moreover, we assume that:

- $f$  is continuous (so it has a least fixpoint  $\text{lfp } f = \bigcup_{n \in \mathbb{N}} f^n(\emptyset)$ )
- $f \circ \gamma \subseteq \gamma \circ f^\sharp$

We let the sequence  $(M_k^\sharp)_{k \in \mathbb{N}}$  be defined by:

$$\begin{aligned} M_0^\sharp &= \perp \\ M_{k+1}^\sharp &= M_k^\sharp \nabla f^\sharp(M_k^\sharp) \end{aligned}$$

Then:

- 1  $(M_k^\sharp)_{k \in \mathbb{N}}$  **is stationary** and we write  $M_{\text{lim}}^\sharp$  for its limit
- 2  $\text{lfp } f \subseteq \gamma(M_{\text{lim}}^\sharp)$

# Fixpoint approximation using a widening operator, proof

We assume all the assumptions of the theorem, and prove the two points:

- ① **Sequence convergence:** We let 
$$\begin{cases} N_0^\# &= \perp \\ N_{k+1}^\# &= f^\#(M_k^\#) \end{cases}$$

Then, convergence follows directly from the definition of widening.

There exists a rank  $K$  from which all iterates are stable.

- ② **Soundness of the limit:**

We prove by induction over  $k$  that  $\forall l \geq k, f^k(\emptyset) \subseteq \gamma(M_l^\#)$ :

- ▶ the result clearly holds for  $k = 0$ ;
- ▶ if the result holds at rank  $k$  and  $l \geq k$  then:

$$\begin{aligned} f^{k+1}(\emptyset) &= f(f^k(\emptyset)) \\ &\subseteq f(\gamma(M_l^\#)) && \text{by induction} \\ &\subseteq \gamma(f^\#(M_l^\#)) && \text{since } f \circ \gamma \subseteq \gamma \circ f^\# \\ &\subseteq \gamma(M_l^\# \nabla f^\#(M_l^\#)) && \text{by definition of } \nabla \\ &= \gamma(M_{l+1}^\#) \end{aligned}$$

When  $(M_k^\#)_{k \in \mathbb{N}}$  converges,  $\forall l \geq K, M_l^\# = M_K^\# = M_\infty^\#$ , thus

$\forall k, f^k(\emptyset) \subseteq \gamma(M_\infty^\#)$  thus  $\text{lfp } f \subseteq \gamma(M_\infty^\#)$



# Example widening iteration

```
int x = 0;

while(TRUE){

    if(x < 10 000){

        x = x + 1;

    } else {

        x = -x;

    }

}
```

# Example widening iteration

```
int x = 0;  
       $x \in [0, 0]$   
while(TRUE){  
    if(x < 10 000){  
        x = x + 1;  
    } else {  
        x = -x;  
    }  
}
```

# Example widening iteration

```

int x = 0;
       $x \in [0, 0]$ 
while(TRUE){
       $x \in [0, 0]$ 
      if(x < 10 000){

          x = x + 1;

      } else {

          x = -x;

      }

  }

```

Entry into the loop

# Example widening iteration

```

int x = 0;
       $x \in [0, 0]$ 
  while(TRUE){
     $x \in [0, 0]$ 
    if(x < 10 000){
       $x \in [0, 0]$ 
      x = x + 1;

    } else {
       $x \in \emptyset$ 
      x = -x;

    }

  }

```

Only the “true” branch may be taken

# Example widening iteration

```

int x = 0;
       $x \in [0, 0]$ 
  while(TRUE){
     $x \in [0, 0]$ 
    if(x < 10 000){
       $x \in [0, 0]$ 
      x = x + 1;
       $x \in [1, 1]$ 
    } else {
       $x \in \emptyset$ 
      x = -x;
       $x \in \emptyset$ 
    }
  }

```

Incrementation

# Example widening iteration

```

int x = 0;
      x ∈ [0, 0]
while(TRUE){
      x ∈ [0, 0]
      if(x < 10 000){
      x ∈ [0, 0]
      x = x + 1;
      x ∈ [1, 1]
      } else {
      x ∈ ∅
      x = -x;
      x ∈ ∅
      }
      x ∈ [1, 1]
}

```

Abstract union at the end of the condition

# Example widening iteration

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, +∞[
    if(x < 10 000){
        x ∈ [0, 0]
        x = x + 1;
        x ∈ [1, 1]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 1]
}

```

**Widening at loop head**

# Example widening iteration

```

int x = 0;
      x  $\in [0, 0]$ 
while(TRUE){
      x  $\in [0, +\infty[$ 
    if(x < 10 000){
      x  $\in [0, 9999]$ 
      x = x + 1;
      x  $\in [1, 1]$ 
    } else {
      x  $\in [10000, +\infty[$ 
      x = -x;
      x  $\in \emptyset$ 
    }
      x  $\in [1, 1]$ 
}

```

Now both branches may be taken



# Example widening iteration

```

int x = 0;
      x  $\in [0, 0]$ 
while(TRUE){
      x  $\in [0, +\infty[$ 
    if(x < 10 000){
      x  $\in [0, 9999]$ 
      x = x + 1;
      x  $\in [1, 10000]$ 
    } else {
      x  $\in [10000, +\infty[$ 
      x = -x;
      x  $\in ]-\infty, -10000]$ 
    }
      x  $\in [1, 1]$ 
}

```

Numerical assignments

# Example widening iteration

```

int x = 0;
      x  $\in [0, 0]$ 
while(TRUE){
      x  $\in [0, +\infty[$ 
    if(x < 10 000){
      x  $\in [0, 9999]$ 
      x = x + 1;
      x  $\in [1, 10000]$ 
    } else {
      x  $\in [10000, +\infty[$ 
      x = -x;
      x  $\in ]-\infty, -10000]$ 
    }
      x  $\in ]-\infty, 10000]$ 
}

```

Abstract union at the end of the condition

# Example widening iteration

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  ] -  $\infty$ , + $\infty$ [
    if(x < 10 000){
        x  $\in$  [0, 9999]
        x = x + 1;
        x  $\in$  [1, 10000]
    } else {
        x  $\in$  [10000, + $\infty$ [
        x = -x;
        x  $\in$  ] -  $\infty$ , -10000]
    }
    x  $\in$  ] -  $\infty$ , 10000]
}

```

**Widening at loop head**

# Example widening iteration

```

int x = 0;
    x  $\in [0, 0]$ 
  while(TRUE){
    x  $\in ] - \infty, +\infty[$ 
    if(x < 10 000){
      x  $\in ] - \infty, 9999]$ 
      x = x + 1;
      x  $\in [1, 10000]$ 
    } else {
      x  $\in [10000, +\infty[$ 
      x = -x;
      x  $\in ] - \infty, -10000]$ 
    }
    x  $\in ] - \infty, 10000]$ 
  }

```

Both branches may be taken

# Example widening iteration

```

int x = 0;
       $x \in [0, 0]$ 
  while(TRUE){
     $x \in ] - \infty, +\infty[$ 
    if(x < 10 000){
       $x \in ] - \infty, 9999]$ 
      x = x + 1;
       $x \in ] - \infty, 10000]$ 
    } else {
       $x \in [10000, +\infty[$ 
      x = -x;
       $x \in ] - \infty, -10000]$ 
    }
     $x \in ] - \infty, 10000]$ 
  }

```

Numerical assignments

# Example widening iteration

```

int x = 0;
    x  $\in [0, 0]$ 
while(TRUE){
    x  $\in ] - \infty, +\infty[$ 
    if(x < 10 000){
        x  $\in ] - \infty, 9999]$ 
        x = x + 1;
        x  $\in ] - \infty, 10000]$ 
    } else {
        x  $\in [10000, +\infty[$ 
        x = -x;
        x  $\in ] - \infty, -10000]$ 
    }
    x  $\in ] - \infty, 10000]$ 
}

```

**Stable! No information at loop head,  
but still, some interesting information inside the loop**

# Loop unrolling

From the example, we observe that **intervals widening is imprecise**:

- quickly **goes to  $-\infty$  or  $+\infty$**
- **ignores possible stable bounds**

**Can we do better ?**

**Yes, we can...** many techniques improve standard widening

## Loop unrolling: postpone widening

We fix an index  $l$ , and postpone widening until after  $l$

$$\begin{aligned}
 M_0^\# &= \perp \\
 M_{k+1}^\# &= \text{join}^\#(M_k^\#, f^\#(M_k^\#)) && \text{if } k < l \\
 M_{k+1}^\# &= M_k^\# \nabla f^\#(M_k^\#) && \text{otherwise}
 \end{aligned}$$

- Typically,  $k$  is set to 1 or 2...
- **Proof** of a new fixpoint approximation theorem: very similar

# Widening with threshold

Now, let us improve the widening itself:

- the standard  $\nabla$  operator of intervals goes straight to  $\infty$
- we can **slow down the process**

## Threshold widening

Let  $\mathcal{T}$  be a **finite set of integers**, called **thresholds**. We let the **threshold widening** be defined by:

$$\begin{aligned} \perp \nabla v^\# &= v^\# \\ v^\# \nabla \perp &= v^\# \\ v_0^\# \nabla v_1^\# &= \begin{cases} v_0^\# & \text{if } v_0^\# \geq v_1^\# \\ \min\{v^\# \in \mathcal{T} \mid \forall i, v_i^\# \leq v^\#\} & \text{if } \{v^\# \in \mathcal{T} \mid \forall i, v_i^\# \leq v^\#\} \neq \emptyset \\ +\infty & \text{otherwise} \end{cases} \end{aligned}$$

- Proof** of the widening property: exercise
- Example** with  $\mathcal{L} = \{10\}$ :

$$[0, 8] \nabla [0, 9] = [0, 10] \quad [0, 8] \nabla [0, 15] = [0, +\infty[$$



# Techniques related to iterations

## No widening after visiting a branch for the first time:

- loop unrolling **postpones** widening for a **finite number of times**
- there are **finitely many branches** in any block of code  
branch: condition block entry or inner loop entry

## Principle

Mark program branches and **apply widening** only **when no new branch was visited during the previous iteration**

## Iteration from a fixpoint approximant:

- **observation:** if  $f \circ \gamma \subseteq \gamma \circ f^\sharp$  and  $\text{lfp } f \subseteq \gamma(M^\sharp)$ , then:  
 $\text{lfp } f = f(\text{lfp } f) \subseteq f \circ \gamma(M^\sharp) \subseteq \gamma \circ f^\sharp(M^\sharp)$
- so  $f^\sharp(M^\sharp)$  **also approximates**  $\text{lfp } f$ , and may be better

## Principle

**After an abstract invariant is found, perform additional iterations**

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```
int x = 0;
```

```
while(TRUE){
```

```
    if(x < 10 000){      9999 will be a threshold value at loop head
```

```
        x = x + 1;
```

```
    } else {
```

```
        x = -x;
```

```
    }
```

```
}
```

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
     $x \in [0, 0]$ 
while(TRUE){

    if(x < 10 000){      9999 will be a threshold value at loop head

        x = x + 1;

    } else {

        x = -x;

    }

}
  
```

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 0]
    if(x < 10 000){      9999 will be a threshold value at loop head

        x = x + 1;

    } else {

        x = -x;

    }

}

```

Entering the loop

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 0]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 0]
        x = x + 1;

    } else {
        x ∈ ∅
        x = -x;

    }

}

```

Only true branch possible

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 0]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 0]
        x = x + 1;
        x ∈ [1, 1]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
}

```

## Incrementation of interval

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 0]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 0]
        x = x + 1;
        x ∈ [1, 1]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 1]
}

```

## Propagation

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 1]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 0]
        x = x + 1;
        x ∈ [1, 1]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 1]
}

```

Join at loop head



# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 1]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 1]
        x = x + 1;
        x ∈ [1, 1]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 1]
}
  
```

Still only the true branch may be taken

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 1]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 1]
        x = x + 1;
        x ∈ [1, 2]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 1]
}

```

## Incrementation of interval

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 1]
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 1]
        x = x + 1;
        x ∈ [1, 2]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 2]
}

```

## Propagation

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  [0, 9999]   instead of [0,  $+\infty$ [
    if(x < 10 000){   9999 will be a threshold value at loop head
        x  $\in$  [0, 1]
        x = x + 1;
        x  $\in$  [1, 2]
    } else {
        x  $\in$   $\emptyset$ 
        x = -x;
        x  $\in$   $\emptyset$ 
    }
    x  $\in$  [1, 2]
}
  
```

## Widening at the loop head, + threshold

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  [0, 9999]    instead of [0,  $+\infty$ [
    if(x < 10 000){    9999 will be a threshold value at loop head
        x  $\in$  [0, 9999]
        x = x + 1;
        x  $\in$  [1, 2]
    } else {
        x  $\in$   $\emptyset$ 
        x = -x;
        x  $\in$   $\emptyset$ 
    }
    x  $\in$  [1, 2]
}
  
```

Still only the true branch may be taken

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 9999]    instead of [0, +∞[
    if(x < 10 000){    9999 will be a threshold value at loop head
        x ∈ [0, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 2]
}

```

## Numerical assignments

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  [0, 9999]   instead of [0,  $+\infty$ [
    if(x < 10 000){   9999 will be a threshold value at loop head
        x  $\in$  [0, 9999]
        x = x + 1;
        x  $\in$  [1, 10000]
    } else {
        x  $\in$   $\emptyset$ 
        x = -x;
        x  $\in$   $\emptyset$ 
    }
    x  $\in$  [1, 10000]
}
  
```

Join at the end of the loop

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 10000]  instead of ] - ∞, +∞[
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ ∅
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 10000]
}
  
```

## Join after widening



# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 10000]  instead of ] - ∞, +∞[
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ [10000, 10000]  instead of [10000, +∞[
        x = -x;
        x ∈ ∅
    }
    x ∈ [1, 10000]
}
  
```

True branch stable, false branch visited for the first time

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [0, 10000]  instead of ] - ∞, +∞[
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ [10000, 10000]  instead of [10000, +∞[
        x = -x;
        x ∈ [-10000, -10000]
    }
    x ∈ [1, 10000]
}
  
```

True branch stable, false branch visited for the first time

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  [0, 10000]  instead of ] -  $\infty$ , + $\infty$ [
    if(x < 10 000){      9999 will be a threshold value at loop head
        x  $\in$  [0, 9999]
        x = x + 1;
        x  $\in$  [1, 10000]
    } else {
        x  $\in$  [10000, 10000]  instead of [10000, + $\infty$ [
        x = -x;
        x  $\in$  [-10000, -10000]
    }
    x  $\in$  [-10000, 10000]
}
  
```

Join at the end of the loop

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [-10000, 10000]  instead of ] - ∞, +∞[
    if(x < 10 000){      9999 will be a threshold value at loop head
        x ∈ [0, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ [10000, 10000]  instead of [10000, +∞[
        x = -x;
        x ∈ [-10000, -10000]
    }
    x ∈ [-10000, 10000]
}
  
```

**Join again: no widening after visiting a new branch**

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [-10000, 10000]  instead of ] - ∞, +∞[
    if(x < 10000){      9999 will be a threshold value at loop head
        x ∈ [-10000, 9999]
        x = x + 1;
        x ∈ [1, 10000]
    } else {
        x ∈ [10000, 10000]  instead of [10000, +∞[
        x = -x;
        x ∈ [-10000, -10000]
    }
    x ∈ [-10000, 10000]
}

```

## Branches

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x ∈ [0, 0]
while(TRUE){
    x ∈ [-10000, 10000]  instead of ] - ∞, +∞[
    if(x < 10000){      9999 will be a threshold value at loop head
        x ∈ [-10000, 9999]
        x = x + 1;
        x ∈ [-9999, 10000]
    } else {
        x ∈ [10000, 10000]  instead of [10000, +∞[
        x = -x;
        x ∈ [-10000, -10000]
    }
    x ∈ [-10000, 10000]
}
  
```

Incrementation of interval in true branch; false branch stable

# Example widening iteration, more precise

## Classical techniques:

- add test values and neighbors as thresholds
- alternate join and widening
- no widening after visiting a new branch

```

int x = 0;
    x  $\in$  [0, 0]
while(TRUE){
    x  $\in$  [-10000, 10000]  instead of ] -  $\infty$ , + $\infty$ [
    if(x < 10000){      9999 will be a threshold value at loop head
        x  $\in$  [-10000, 9999]
        x = x + 1;
        x  $\in$  [-9999, 10000]
    } else {
        x  $\in$  [10000, 10000]  instead of [10000, + $\infty$ [
        x = -x;
        x  $\in$  [-10000, -10000]
    }
    x  $\in$  [-10000, 10000]
}
  
```

Everything is stable; exact ranges inferred

# Widening and monotonicity

**Remarks** about the widening over intervals:

- it is **monotone** in its second argument,
- but it is **not monotone in its first argument!**

In fact, interesting widenings **are not monotone in their first argument:**

Let  $(D^\#, \sqsubseteq)$  be an infinite height domain, with a widening  $\nabla$  that is stable  $(\forall v^\#, v^\# \nabla v^\# = v^\#)$  and such that  $\forall v_0^\#, v_1^\#, \forall i, v_i^\# \sqsubseteq v_0^\# \nabla v_1^\#$ . Then,  $\nabla$  is **not monotone in its first argument** (proof: Patrick Cousot).

**Proof:** we assume it is, let  $w_0^\# \sqsubset w_1^\# \sqsubset \dots$  be an infinite chain over  $D^\#$  and define  $v_0^\# = w_0^\#, v_{k+1}^\# = v_k^\# \nabla w_{k+1}^\#$ ; we prove by induction that  $v_k^\# = w_k^\#$ :

- clear at rank 0
- we assume that  $v_k^\# = w_k^\#$ : then  $v_{k+1}^\# = v_k^\# \nabla w_{k+1}^\#$ , so  $w_{k+1}^\# \sqsubseteq v_{k+1}^\#$ ;  
moreover,  $v_{k+1}^\# = v_k^\# \nabla w_{k+1}^\# = w_k^\# \nabla w_{k+1}^\# \sqsubseteq w_{k+1}^\# \nabla w_{k+1}^\# = w_{k+1}^\#$

This contradicts the widening definition: the sequence should be stationary.



# Outline

- 1 Another Soundness Relation
- 2 Revisiting Abstract Iteration
- 3 Conclusion

# Summary

## This lecture:

- **abstraction** and its formalization
- **computation of an abstract semantics** in a very simplified case

## Next lectures:

- **construction** of a few **non trivial abstractions**
- **more general** ways to **compute sound abstract properties**