

# IA159 Formal Methods for Software Analysis

## Abstract Interpretation

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## focus

- lattices and fixpoints
- abstract interpretation with examples
- widening and narrowing

## source

- P. Cousot and R. Cousot: *Abstract Interpretation: A Unified Lattice Model for Static Analysis of Programs by Construction or Approximation of Fixpoints*, POPL 1977.

Special thanks to Marek Trtík for providing me with his slides.

## Floyd's conjecture

To prove static properties of program it is often sufficient to consider sets of states associated with each program point.

### examples

- to check safety properties (reachability of an error state), one only needs to know reachable states
- for many optimizations during compilation, static information is sufficient (e.g. detection of live variables, available expressions, etc.)

## operational semantics

- defines how a state changes along program execution
- it is concerned about **computational sequences**
- computes a function relating input and output states

## operational semantics

- defines how a state changes along program execution
- it is concerned about **computational sequences**
- computes a function relating input and output states

## static semantic

- observes which states pass which program location
- it is concerned with **observed sets of states at locations**
- computes a function assigning a set of states to each program location

# Motivation for abstract interpretation

- it is usually impossible to compute the sets of reachable states precisely
- we can compute them on some level of abstraction
- for example, instead with precise numbers we work only with abstract values  $\{+, 0, -\}$
- abstraction brings some level of imprecision, for example,  $15 - 17$  is seen as  $(+) - (+)$ , which can be  $+, 0, -$

## Preliminaries: lattices and fixpoints

Let  $(L, \leq)$  be a partially ordered set and  $M \subseteq L$ .

- $x \in L$  is an **upper bound** of  $M$  iff  $y \leq x$  holds for all  $y \in M$
- $x \in L$  is a **lower bound** of  $M$  iff  $x \leq y$  holds for all  $y \in M$
- **supremum** of  $M$  is the least upper bound of  $M$
- **infimum** of  $M$  is the greatest lower bound of  $M$
- $\text{sup}(M)$  and  $\text{inf}(M)$  denote supremum and infimum of  $M$ , respectively



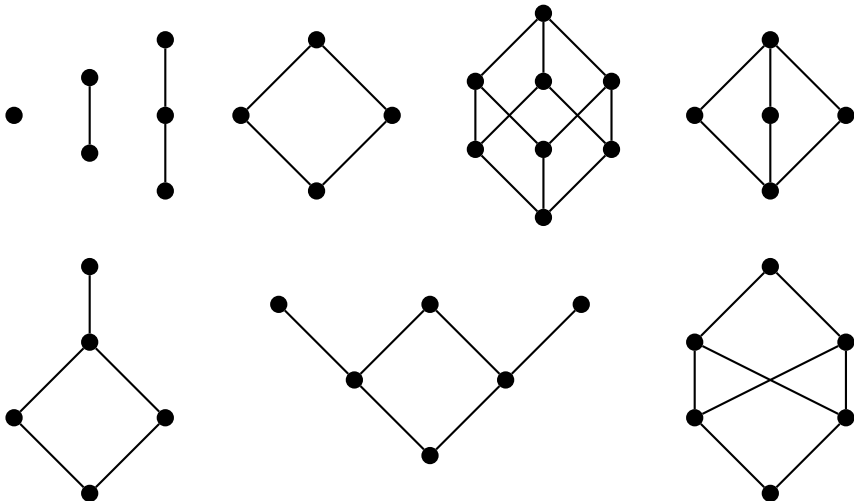
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## Definition (Complete lattice)

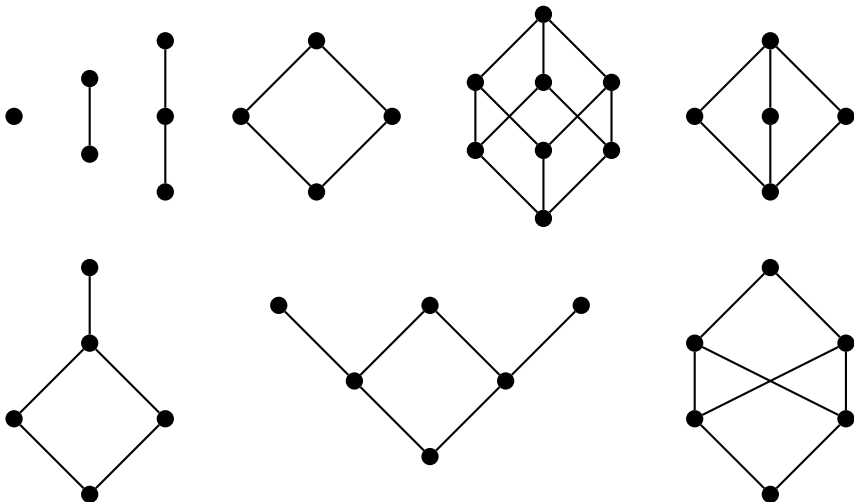
An ordered set  $(L, \leq)$  is called **complete lattice**, if for each  $M \subseteq L$  there exist both  $\text{sup}(M)$  and  $\text{inf}(M)$ .

# Introduction to lattices



Which of the partially ordered sets are complete lattices?

# Introduction to lattices

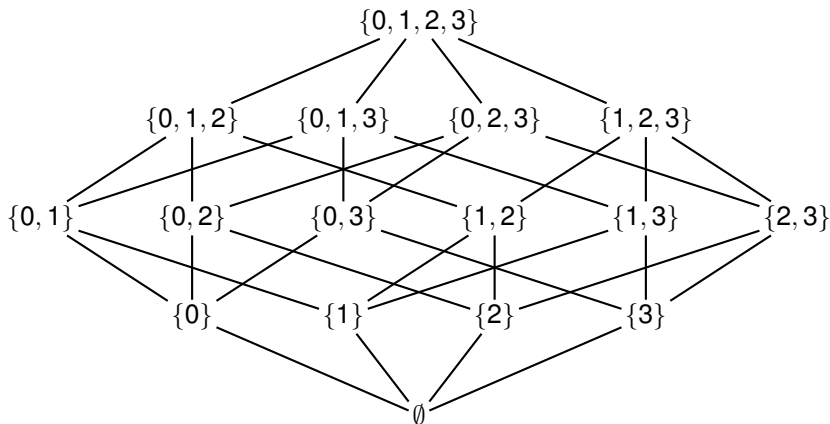


Which of the partially ordered sets are complete lattices?  
(All of the top row and the left of the bottom row.)

# Introduction to lattices

For every set  $S$ , the powerset  $\mathcal{P}(S)$  with the partial order  $\subseteq$  is a complete lattice.

For example,  $(\mathcal{P}(\{0, 1, 2, 3\}), \subseteq)$  looks like:



Let  $(L, \leq)$  be a complete lattice.

- the greatest element  $\top = \text{sup}(L)$  is called **one** of  $L$
- the least element  $\perp = \text{inf}(L)$  of  $L$  is called **zero** of  $L$
- the lattice is of **finite height** if there exists  $h \in \mathbb{N}$  such that the length of each strictly increasing chain of elements of  $L$  is less than or equal to  $h$
- minimal such  $h$  is called **lattice height**

# Fixpoint and Knaster-Tarski fixpoint theorem

Let  $(L, \leq)$  be a complete lattice.

- a function  $f : L \rightarrow L$  is **monotone** if for all  $x, y \in L$  it holds

$$x \leq y \implies f(x) \leq f(y)$$

- $x \in L$  is called a **fixpoint** of  $f$  if  $f(x) = x$

# Fixpoint and Knaster-Tarski fixpoint theorem

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- $x \in L$  is called a **fixpoint** of  $f$  if  $f(x) = x$

## Theorem (Knaster-Tarski)

*Let  $(L, \leq)$  be a complete lattice and  $f : L \rightarrow L$  be a monotone function. Then the set of fixpoints of  $f$  with partial order  $\leq$  is also a complete lattice.*

## Theorem (Kleene)

*Let  $(L, \leq)$  be a complete lattice of finite height and  $f : L \rightarrow L$  a monotone function. Then there exists  $n \in \mathbb{N}$  such that for all  $k \in \mathbb{N}$  it is  $f^n(\perp) = f^{n+k}(\perp)$  and  $f^n(\perp)$  is the least fixpoint of  $f$ .*



## Theorem (Kleene)

*Let  $(L, \leq)$  be a complete lattice of finite height and  $f : L \rightarrow L$  a monotone function. Then there exists  $n \in \mathbb{N}$  such that for all  $k \in \mathbb{N}$  it is  $f^n(\perp) = f^{n+k}(\perp)$  and  $f^n(\perp)$  is the least fixpoint of  $f$ .*

**Proof:** Since  $\perp$  is the least element of  $L$ , we have  $\perp \leq f(\perp)$ . Since  $f$  is monotone, then  $f(\perp) \leq f(f(\perp))$  and by induction  $f^i(\perp) \leq f^{i+1}(\perp)$ . Thus, we have a nondecreasing chain  $\perp \leq f(\perp) \leq f^2(\perp) \leq \dots$ . Since  $L$  is assumed to be of a finite height, there must exist  $n \in \mathbb{N}$  such that  $f^n(\perp) = f^{n+1}(\perp)$ . To show that  $f^n(\perp)$  is a least fixpoint of  $f$ , let us assume  $x$  is another fixpoint of  $f$ . Since  $\perp \leq x$  and  $f(\perp) \leq f(x) = x$  from monotonicity of  $f$ , we get by induction  $f^n(\perp) \leq x$ .  $\square$

# Fixpoint computation

algorithm for the least fixpoint computation

```
1  $x \leftarrow \perp$   
2 do  
3    $t \leftarrow x$   
4    $x \leftarrow f(x)$   
5 while  $x \neq t$ 
```

If we replace the first line with  $x \leftarrow \top$ , we get the **greatest fixpoint**.

## Lemma (Product lattice)

*Let  $(L_1, \leq_1), \dots, (L_n, \leq_n)$  be complete lattices and order  $\leq$  on  $L_1 \times \dots \times L_n$  is defined as  $(x_1, \dots, x_n) \leq (y_1, \dots, y_n)$  iff*

$$x_1 \leq_1 y_1 \wedge \dots \wedge x_n \leq_n y_n.$$

*Then  $(L_1 \times \dots \times L_n, \leq)$  is a complete lattice.*

# Fixpoints on product lattices

Let  $(L, \leq)$  be a complete lattice and  $(L^n, \sqsubseteq)$  be the corresponding product lattice. Further, let  $F_1, \dots, F_n : L^n \rightarrow L$  be monotone functions, i.e.  $(x_1, \dots, x_n) \sqsubseteq (y_1, \dots, y_n)$  implies  $F_i(x_1, \dots, x_n) \leq F_i(y_1, \dots, y_n)$  for each  $1 \leq i \leq n$ . Then the function  $F : L^n \rightarrow L^n$  defined as

$$F(x_1, \dots, x_n) = (F_1(x_1, \dots, x_n), \dots, F_n(x_1, \dots, x_n))$$

is a monotone function in  $(L^n, \sqsubseteq)$ . Further, the least fixpoint of  $F$  is the least solution of the system:

$$\begin{aligned}x_1 &= F_1(x_1, \dots, x_n) \\ &\vdots \\ x_n &= F_n(x_1, \dots, x_n)\end{aligned}$$

# Fixpoint computation of product lattices

naive algorithm for fixpoint computation

```
1  $\vec{x} \leftarrow \perp$   
2 do  
3    $\vec{t} \leftarrow \vec{x}$   
4    $\vec{x} \leftarrow F(\vec{x})$   
5 while  $\vec{x} \neq \vec{t}$ 
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# Fixpoint computation of product lattices

naive algorithm for fixpoint computation

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1  $\vec{x} \leftarrow \vec{\perp}$ 
2 do
3    $\vec{t} \leftarrow \vec{x}$ 
4    $\vec{x} \leftarrow F(\vec{x})$ 
5 while  $\vec{x} \neq \vec{t}$ 
```

better algorithm for fixpoint computation (faster convergence)

```
1  $\vec{x} \leftarrow \vec{\perp}$ 
2 do
3    $\vec{t} \leftarrow \vec{x}$ 
4    $x_1 \leftarrow F_1(x_1, \dots, x_n)$ 
5    $\vdots$ 
6    $x_n \leftarrow F_n(x_1, \dots, x_n)$ 
7 while  $\vec{x} \neq \vec{t}$ 
```

Abstract interpretation

- an abstract interpretation of a program is kind of a **static semantic**, where original data domains are replaced with **abstract** ones
- abstract data domain must constitute a **complete lattice**
- semantic of program instructions have to be changed as well: we define unique **monotone function** for each program instruction



## Definition (Abstract interpretation)

An **abstract interpretation**  $I$  of a program  $P$  with  $n$  program locations is a tuple

$$I = \langle L, \circ, \leq, \top, \perp, F \rangle$$

where  $(L, \leq)$  is complete lattice,  $\top$  and  $\perp$  are one and zero of  $(L, \leq)$ ,  $\circ$  is equal either to join or meet operation, and  $F$  is a monotone function on product lattice  $(L^n, \leq)$  defining the interpretation of basic instructions.

- meet operator is defined as  $a \circ b = \inf(\{a, b\})$
- join operator is defined as  $a \circ b = \sup(\{a, b\})$ .

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Typically,  $F(\vec{x}) = (F_1(\vec{x}), \dots, F_n(\vec{x}))$ , where each  $F_i : L^n \rightarrow L$  defines effect of  $i$ -th program instruction.

## Example: Available expressions

A nontrivial expression in a program is **available** at a program location if its current value has already been computed earlier in the execution.

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```
var x,y,z,a,b;  
z = a+b;  
y = a*b;  
while (y > a+b) {  
    a = a+1;  
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A.I.:  $I = \langle \mathcal{P}(AExprs), \cap, \subseteq, AExprs, \emptyset, \lambda \vec{x}. (F_1(\vec{x}), \dots, F_6(\vec{x})) \rangle$

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Product lattice:  $(\mathcal{P}^6(AExprs), \leq)$ .

```
var x, y, z, a, b;           X1
z = a+b;                     X2
y = a*b;                     X3
while (y > a+b) {           X4
    a = a+1;                 X5
    x = a+b;                 X6
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Direction: Forward

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**Analysis: Must**

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Are all functions  $F_i$  monotone?

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**Proof  $F_4$ :** Let  $\vec{x}, \vec{y} \in \mathcal{P}^6(AExprs)$  such that  $\vec{x} \leq \vec{y}$ . ...

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<code>}</code>	

Then  $x_3 \subseteq y_3$  and  $x_6 \subseteq y_6$ , which implies  $(x_3 \cap x_6) \subseteq (y_3 \cap y_6) \dots$

# Example: Available expressions

After fixpoint computation ...

```
var x, y, z, a, b;  
z = a+b;  
y = a*b;  
while (y > a+b) {  
    a = a+1;  
    x = a+b;  
}
```

```
x1 = ∅  
x2 = {a+b}  
x3 = {a+b, a*b}  
x4 = {a+b, y>a+b}  
x5 = ∅  
x6 = {a+b}
```

**Solution: Minimal**

# Example: Available expressions

After fixpoint computation ...

```
var x, y, z, a, b;  
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}
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```
x1 = ∅  
x2 = {a+b}  
x3 = {a+b, a*b}  
x4 = {a+b, y>a+b}  
x5 = ∅  
x6 = {a+b}
```

The expression `a+b` in the loop head is available (does not have to be computed).

## Example: Live variables

A variable is **live** at a program point if its current value may be read during the remaining execution of the program.

```
var x, y, z;  
x = input();  
while (x > 1) {  
    y = x/2;  
    if (y > 3)  
        x = x-y;  
    z = x-4;  
    if (z > 0)  
        x = x/2;  
    z = z-1; }  
output(x);
```



## Example: Live variables

A variable is **live** at a program point if its current value may be read during the remaining execution of the program.

$Vars = \{x, y, z\}$  and  $I = \langle \mathcal{P}(Vars), \cup, \subseteq, Vars, \emptyset, \lambda \vec{x}. (F_1(\vec{x}), \dots, F_{11}(\vec{x})) \rangle$

```
var x, y, z;  
x = input();  
while (x > 1) {  
    y = x/2;  
    if (y > 3)  
        x = x-y;  
    z = x-4;  
    if (z > 0)  
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Product lattice is  $(\mathcal{P}^{11}(\text{Vars}), \leq)$ .

$X_1 = X_2 \setminus \{x, y, z\}$	<code>var x, y, z;</code>
$X_2 = X_3 \setminus \{x\}$	<code>x = input();</code>
$X_3 = (X_4 \cup X_{11}) \cup \{x\}$	<code>while (x &gt; 1) {</code>
$X_4 = (X_5 \setminus \{y\}) \cup \{x\}$	<code>    y = x/2;</code>
$X_5 = (X_6 \cup X_7) \cup \{y\}$	<code>    if (y &gt; 3)</code>
$X_6 = (X_7 \setminus \{x\}) \cup \{x, y\}$	<code>        x = x-y;</code>
$X_7 = (X_8 \setminus \{z\}) \cup \{x\}$	<code>        z = x-4;</code>
$X_8 = (X_9 \cup X_{10}) \cup \{z\}$	<code>        if (z &gt; 0)</code>
$X_9 = (X_{10} \setminus \{x\}) \cup \{x\}$	<code>            x = x/2;</code>
$X_{10} = (X_3 \setminus \{z\}) \cup \{z\}$	<code>            z = z-1; } </code>
$X_{11} = \{x\}$	<code>output(x);</code>

## Example: Live variables

A variable is **live** at a program point if its current value may be read during the remaining execution of the program.

Direction: Backward

$$\begin{aligned} X_1 &= X_2 \setminus \{x, y, z\} \\ X_2 &= X_3 \setminus \{x\} \\ X_3 &= (X_4 \cup X_{11}) \cup \{x\} \\ X_4 &= (X_5 \setminus \{y\}) \cup \{x\} \\ X_5 &= (X_6 \cup X_7) \cup \{y\} \\ X_6 &= (X_7 \setminus \{x\}) \cup \{x, y\} \\ X_7 &= (X_8 \setminus \{z\}) \cup \{x\} \\ X_8 &= (X_9 \cup X_{10}) \cup \{z\} \\ X_9 &= (X_{10} \setminus \{x\}) \cup \{x\} \\ X_{10} &= (X_3 \setminus \{z\}) \cup \{z\} \\ X_{11} &= \{x\} \end{aligned}$$

```
var x, y, z;
x = input();
while (x > 1) {
    y = x/2;
    if (y > 3)
        x = x-y;
    z = x-4;
    if (z > 0)
        x = x/2;
    z = z-1; }
output(x);
```

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### Analysis: May

$$\begin{aligned}X_1 &= X_2 \setminus \{x, y, z\} \\X_2 &= X_3 \setminus \{x\} \\X_3 &= (X_4 \cup X_{11}) \cup \{x\} \\X_4 &= (X_5 \setminus \{y\}) \cup \{x\} \\X_5 &= (X_6 \cup X_7) \cup \{y\} \\X_6 &= (X_7 \setminus \{x\}) \cup \{x, y\} \\X_7 &= (X_8 \setminus \{z\}) \cup \{x\} \\X_8 &= (X_9 \cup X_{10}) \cup \{z\} \\X_9 &= (X_{10} \setminus \{x\}) \cup \{x\} \\X_{10} &= (X_3 \setminus \{z\}) \cup \{z\} \\X_{11} &= \{x\}\end{aligned}$$

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        x = x-y;
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        x = x/2;
    z = z-1; }
output(x);
```

## Example: Live variables

A variable is **live** at a program point if its current value may be read during the remaining execution of the program.

### Solution: Minimal

$$X_1 = \emptyset$$

$$X_2 = \emptyset$$

$$X_3 = \{x\}$$

$$X_4 = \{x\}$$

$$X_5 = \{x, y\}$$

$$X_6 = \{x, y\}$$

$$X_7 = \{x\}$$

$$X_8 = \{x, z\}$$

$$X_9 = \{x, z\}$$

$$X_{10} = \{x, z\}$$

$$X_{11} = \{x\}$$

$$X_1 = X_2 \setminus \{x, y, z\}$$

$$X_2 = X_3 \setminus \{x\}$$

$$X_3 = (X_4 \cup X_{11}) \cup \{x\}$$

$$X_4 = (X_5 \setminus \{y\}) \cup \{x\}$$

$$X_5 = (X_6 \cup X_7) \cup \{y\}$$

$$X_6 = (X_7 \setminus \{x\}) \cup \{x, y\}$$

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var x, y, z;  
x = input();  
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    z = z-1; }  
output(x);
```

## Example: Live variables

A variable is **live** at a program point if its current value may be read during the remaining execution of the program.

Variables  $y, z$  are never live together.

$$X_1 = \emptyset$$

$$X_2 = \emptyset$$

$$X_3 = \{x\}$$

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$$X_5 = \{x, y\}$$

$$X_6 = \{x, y\}$$

$$X_7 = \{x\}$$

$$X_8 = \{x, z\}$$

$$X_9 = \{x, z\}$$

$$X_{10} = \{x, z\}$$

$$X_{11} = \{x\}$$

$$X_1 = X_2 \setminus \{x, y, z\}$$

$$X_2 = X_3 \setminus \{x\}$$

$$X_3 = (X_4 \cup X_{11}) \cup \{x\}$$

$$X_4 = (X_5 \setminus \{y\}) \cup \{x\}$$

$$X_5 = (X_6 \cup X_7) \cup \{y\}$$

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    z = x-4;  
    if (z > 0)  
        x = x/2;  
    z = z-1; }  
output(x);
```

## Example: Reaching definitions

The **reaching definitions** for a given program point are those assignments that may have defined the current values of variables.

```
var x, y, z;  
x = input();  
while (x > 1) {  
    y = x/2;  
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```

Assignments:

*Asgns* = { $x = \text{input}()$ ,  $y = x/2$ ,  $x = x-y$ ,  
 $z = x-4$ ,  $x = x/2$ ,  $z = z-1$ }



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  z = x-4;  
  if (z > 0)  
    x = x/2;  
  z = z-1; }  
output(x);
```

Assignments:

$Asgns = \{x = \text{input}(), y = x/2, x = x-y,$   
 $z = x-4, x = x/2, z = z-1\}$

$I = \langle \mathcal{P}(Asgns), \cup, \subseteq, Asgns, \emptyset,$   
 $\lambda \vec{x}. (F_1(\vec{x}), \dots, F_{11}(\vec{x})) \rangle$

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  if (y > 3)  
    x = x-y;  
  z = x-4;  
  if (z > 0)  
    x = x/2;  
  z = z-1; }  
output(x);
```

Assignments:

$Asgns = \{x = input(), y = x/2, x = x-y, z = x-4, x = x/2, z = z-1\}$

$I = \langle \mathcal{P}(Asgns), \cup, \subseteq, \emptyset, \lambda \vec{x}. (F_1(\vec{x}), \dots, F_{11}(\vec{x})) \rangle$

Product lattice:  $(\mathcal{P}^{11}(Asgns), \subseteq)$

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    z = x-4;  
    if (z > 0)  
        x = x/2;  
    z = z-1;  
}  
output(x);
```

Assignments:

$Asgns = \{x = \text{input}(), y = x/2, x = x-y,$   
 $z = x-4, x = x/2, z = z-1\}$

$I = \langle \mathcal{P}(Asgns), \cup, \subseteq, Asgns, \emptyset,$   
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Product lattice:  $(\mathcal{P}^{11}(Asgns), \subseteq)$

Direction: Forward

Analysis: May

Solution: Minimal

## Example: Busy expressions

An expression is **busy** if it will definitely be evaluated again before its value changes.

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An expression is **busy** if it will definitely be evaluated again before its value changes.

Direction: Backward  
Analysis: Must  
Solution: Minimal

We may consider different abstraction levels of variable values:

- sets of integer values:  $\mathcal{P}(\mathbb{Z})$
- intervals:  $\{[l, u] \mid l, u \in \mathbb{Z} \cup \{-\infty, \infty\}, l \leq u\} \cup \{\perp\}$
- only signs with zero:  $\mathcal{P}(\{-, 0, +\})$
- initialized or not:  $\{\perp, \top\}$

We may consider different abstraction levels of variable values:

- sets of integer values:  $\mathcal{P}(\mathbb{Z})$
- intervals:  $\{[l, u] \mid l, u \in \mathbb{Z} \cup \{-\infty, \infty\}, l \leq u\} \cup \{\perp\}$
- only signs with zero:  $\mathcal{P}(\{-, 0, +\})$
- initialized or not:  $\{\perp, \top\}$

Which abstraction is more precise than other?

Fixpoint approximation techniques: widening and narrowing



When the extreme fixpoints of the system of equations cannot be computed in finitely many steps, they can be approximated.

Generally, we have these two approaches:

- 1 we can find more abstract interpretation
- 2 we can make approximations in the current interpretation to accelerate convergence of Kleene's sequence

Here we are concerned with the second approach – the technique called **widening**.

Widening makes Kleene's sequence to converge

- to a fixpoint possibly greater than the least one or
- to an element  $s$ , such that  $s > F(s)$ .

In the second case, since  $s$  is greater than the least fixpoint, we can use **narrowing** to make the solution more precise – i.e. to find some fixpoint smaller than  $s$  but possibly greater than the least fixpoint.

- If the Kleene's sequence does not converge, then there exists a location  $x_i$  on a program loop where the sequence does not converge.
- We need a **widening function**  $\nabla : L \times L \rightarrow L$ , which is applied every time the location  $x_i$  is updated:  $x_i = x_i \nabla F_i(\vec{x})$ .
- We must define  $\nabla$  such that
  - for each  $x, y \in L$ ,  $x \circ y \leq x \nabla y$ , i.e.  $\nabla$  overapproximates operation  $\circ$ ,
  - it ensures that every infinite sequence of elements occurring in  $x_i$  is not strictly increasing.

Example: Interval bounds of integer variable  $x$

```
{locations are after}  
1  x = 1;  
2  while (x <= 100) {  
3      x = x + 1;  
4  }
```

Example: Interval bounds of integer variable  $x$

{locations are after}

```
1  x = 1;
2  while (x <= 100) {
3      x = x + 1;
4  }
```

{functions}

```
 $x_1 = [1, 1]$ 
 $x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$ 
 $x_3 = x_2 + [1, 1]$ 
 $x_4 = (x_1 \cup x_3) \cap [101, \infty]$ 
```

Example: Interval bounds of integer variable  $x$

{locations are after}	{functions}
1 $x = 1;$	$x_1 = [1, 1]$
2 $\text{while } (x \leq 100) \{$	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 $x = x + 1;$	$x_3 = x_2 + [1, 1]$
4 $\}$	$x_4 = (x_1 \cup x_3) \cap [101, \infty]$

Widening operator  $\nabla$ :  $[i, j] \nabla [k, l] = [\text{ite}(k < i, -\infty, i), \text{ite}(l > j, \infty, j)]$

Example: Interval bounds of integer variable  $x$

{locations are after}	{functions}
1 $x = 1;$	$x_1 = [1, 1]$
2 while ( $x \leq 100$ ) {	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 $x = x + 1;$	$x_3 = x_2 + [1, 1]$
4 }	$x_4 = (x_1 \cup x_3) \cap [101, \infty]$

Widening operator  $\nabla$ :  $[i, j] \nabla [k, l] = [ite(k < i, -\infty, i), ite(l > j, \infty, j)]$

```
{no widening}  
 $x_1 = [1, 1]$   
 $x_2 = [1, 100]$   
 $x_3 = [2, 101]$   
 $x_4 = [101, 101]$   
100 iterations
```

Example: Interval bounds of integer variable  $x$

{locations are after}	{functions}
1 $x = 1;$	$x_1 = [1, 1]$
2 while ( $x \leq 100$ ) {	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 $x = x + 1;$	$x_3 = x_2 + [1, 1]$
4 }	$x_4 = (x_1 \cup x_3) \cap [101, \infty]$

Widening operator  $\nabla$ :  $[i, j] \nabla [k, l] = [ite(k < i, -\infty, i), ite(l > j, \infty, j)]$

{no widening}	{ $x_3 = x_3 \nabla (x_2 + [1, 1])$ }
$x_1 = [1, 1]$	$x_1 = [1, 1]$
$x_2 = [1, 100]$	$x_2 = [1, 100]$
$x_3 = [2, 101]$	$x_3 = [2, \infty]$
$x_4 = [101, 101]$	$x_4 = [101, \infty]$
100 iterations	2 iterations



Example: Interval bounds of integer variable  $x$

{locations are after}	{functions}
1 $x = 1;$	$x_1 = [1, 1]$
2 while ( $x \leq 100$ ) {	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 $x = x + 1;$	$x_3 = x_2 + [1, 1]$
4 }	$x_4 = (x_1 \cup x_3) \cap [101, \infty]$

Widening operator  $\nabla$ :  $[i, j] \nabla [k, l] = [ite(k < i, -\infty, i), ite(l > j, \infty, j)]$

{no widening}	{ $x_3 = x_3 \nabla (x_2 + [1, 1])$ }
$x_1 = [1, 1]$	$x_1 = [1, 1]$
$x_2 = [1, 100]$	$x_2 = [1, 100]$
$x_3 = [2, 101]$	$x_3 = [2, \infty]$
$x_4 = [101, 101]$	$x_4 = [101, \infty]$
<b>100 iterations</b>	<b>2 iterations</b>

- When widening ends with  $s > F(s)$ , we improve solution  $s$  as follows:  $s \geq F(s) \geq \dots \geq F^n(s) \geq \dots \geq s_0$ , where  $s_0$  is the least fixpoint.
- When the sequence is finite, its limit is better approximation of  $s_0$ .
- If the sequence is infinite, we apply **narrowing function**  $\Delta: L \times L \rightarrow L$  at not stabilizing location  $x_i$  such that  $x_i = x_i \Delta F_i(\vec{x})$ .
- Operator  $\Delta$  must satisfy:
  - for each  $x, y \in L$ ,  $x > y \rightarrow (x \geq x \Delta y \geq y)$ , i.e.  $\Delta$  tries to slow down the decreasing of the sequence,
  - it ensures, that every infinite sequence of elements starting from any  $s$  is not strictly decreasing.

Example: Interval bounds of integer variable  $x$

```
{locations are after}
1  x = 1;
2  while (x <= 100) {
3      x = x + 1;
4  }
```

Example: Interval bounds of integer variable  $x$

```
{locations are after}  
1  x = 1;  
2  while (x <= 100) {  
3      x = x + 1;  
4  }
```

```
{functions}  
x1 = [1, 1]  
x2 = (x1 ∪ x3) ∩ [-∞, 100]  
x3 = x2 + [1, 1]  
x4 = (x1 ∪ x3) ∩ [101, ∞]
```

Example: Interval bounds of integer variable  $x$

{locations are after}	{functions}
1 $x = 1;$	$x_1 = [1, 1]$
2 $\text{while } (x \leq 100) \{$	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 $x = x + 1;$	$x_3 = x_2 + [1, 1]$
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Narrowing operator  $\Delta$ :

$[i, j] \Delta [k, l] = [\text{ite}(i = -\infty, k, \min(i, k)), \text{ite}(j = \infty, l, \max(j, l))]$

Example: Interval bounds of integer variable  $x$

<code>{locations are after}</code>	<code>{functions}</code>
1 <code>x = 1;</code>	$x_1 = [1, 1]$
2 <code>while (x &lt;= 100) {</code>	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 <code>    x = x + 1;</code>	$x_3 = x_2 + [1, 1]$
4 <code>}</code>	$x_4 = (x_1 \cup x_3) \cap [101, \infty]$

Narrowing operator  $\Delta$ :

$[i, j] \Delta [k, l] = [ite(i = -\infty, k, \min(i, k)), ite(j = \infty, l, \max(j, l))]$

<code>{no widening}</code>	<code>{widening}</code>
$x_1 = [1, 1]$	$x_1 = [1, 1]$
$x_2 = [1, 100]$	$x_2 = [1, 100]$
$x_3 = [2, 101]$	$x_3 = [2, \infty]$
$x_4 = [101, 101]$	$x_4 = [101, \infty]$
100 iterations	2 iteration

Example: Interval bounds of integer variable  $x$

<code>{locations are after}</code>	<code>{functions}</code>
1 <code>x = 1;</code>	$x_1 = [1, 1]$
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<code>{no widening}</code>	<code>{widening}</code>	<code>{<math>x_3 = x_3 \Delta (x_2 + [1, 1])</math>}</code>
$x_1 = [1, 1]$	$x_1 = [1, 1]$	$x_1 = [1, 1]$
$x_2 = [1, 100]$	$x_2 = [1, 100]$	$x_2 = [1, 100]$
$x_3 = [2, 101]$	$x_3 = [2, \infty]$	$x_3 = [2, 101]$
$x_4 = [101, 101]$	$x_4 = [101, \infty]$	$x_4 = [101, 101]$
100 iterations	2 iteration	+1 iteration

# Narrowing

Example: Interval bounds of integer variable  $x$

<code>{locations are after}</code>	<code>{functions}</code>
1 <code>x = 1;</code>	$x_1 = [1, 1]$
2 <code>while (x &lt;= 100) {</code>	$x_2 = (x_1 \cup x_3) \cap [-\infty, 100]$
3 <code>    x = x + 1;</code>	$x_3 = x_2 + [1, 1]$
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<code>{no widening}</code>	<code>{widening}</code>	<code>{<math>x_3 = x_3 \Delta (x_2 + [1, 1])</math>}</code>
$x_1 = [1, 1]$	$x_1 = [1, 1]$	$x_1 = [1, 1]$
$x_2 = [1, 100]$	$x_2 = [1, 100]$	$x_2 = [1, 100]$
$x_3 = [2, 101]$	$x_3 = [2, \infty]$	$x_3 = [2, 101]$
$x_4 = [101, 101]$	$x_4 = [101, \infty]$	$x_4 = [101, 101]$
100 iterations	2 iteration	+1 iteration