



Static Program Analysis

Lecture 5: Dataflow Analysis IV (Worklist Algorithm & MOP Solution)

Winter Semester 2016/17

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<https://moves.rwth-aachen.de/teaching/ws-1617/spa/>

Re-Scheduling of First Exam

New date: Tue 21 Feb, 15:00 - 17:00, AH 2/3

Recap: The Fixpoint Approach

Outline of Lecture 5

Recap: The Fixpoint Approach

Efficient Fixpoint Computation

The MOP Solution

Another Analysis: Constant Propagation

Recap: The Fixpoint Approach

The Fixpoint Theorem



Alfred Tarski (1901–1983)



Bronislaw Knaster (1893–1990)

Theorem (Fixpoint Theorem by Tarski and Knaster)

Let (D, \sqsubseteq) be a complete lattice satisfying ACC and $\Phi : D \rightarrow D$ monotonic. Then

$$\text{fix}(\Phi) := \bigsqcup \{ \Phi^k(\perp) \mid k \in \mathbb{N} \}$$

is the **least fixpoint** of Φ where $\Phi^0(d) := d$ and $\Phi^{k+1}(d) := \Phi(\Phi^k(d))$.

Function requirements for dataflow analysis

All transfer functions must be a **monotonic**

Recap: The Fixpoint Approach

Dataflow Systems

Definition (Dataflow system)

A **dataflow system** $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ consists of

- a finite set of (program) **labels** Lab (here: Lab_c),
- a set of **extremal labels** $E \subseteq Lab$ (here: $\{\text{init}(c)\}$ or $\{\text{final}(c)\}$),
- a **flow relation** $F \subseteq Lab \times Lab$ (here: $\text{flow}(c)$ or $\text{flow}^R(c)$),
- a **complete lattice** (D, \sqsubseteq) satisfying ACC (with LUB operator \sqcup and least element \perp),
- an **extremal value** $\iota \in D$ (for the extremal labels), and
- a collection of **monotonic transfer functions** $\{\varphi_l \mid l \in Lab\}$ of type $\varphi_l : D \rightarrow D$.

Recap: The Fixpoint Approach

Dataflow Systems and Fixpoints

Definition (Dataflow equation system)

Given: dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$, $Lab = \{1, \dots, n\}$ (w.l.o.g.)

- S determines the **equation system** (where $l \in Lab$)

$$AI_l = \begin{cases} \iota & \text{if } l \in E \\ \bigsqcup \{\varphi_{l'}(AI_{l'}) \mid (l', l) \in F\} & \text{otherwise} \end{cases}$$

- $(d_1, \dots, d_n) \in D^n$ is called a **solution** if

$$d_l = \begin{cases} \iota & \text{if } l \in E \\ \bigsqcup \{\varphi_{l'}(d_{l'}) \mid (l', l) \in F\} & \text{otherwise} \end{cases}$$

- S determines the **transformation**

$$\Phi_S : D^n \rightarrow D^n : (d_1, \dots, d_n) \mapsto (d'_1, \dots, d'_n)$$

where

$$d'_l := \begin{cases} \iota & \text{if } l \in E \\ \bigsqcup \{\varphi_{l'}(d_{l'}) \mid (l', l) \in F\} & \text{otherwise} \end{cases}$$

Corollary

$(d_1, \dots, d_n) \in D^n$ **solves** the equation system iff it is a **fixpoint** of Φ_S

Recap: The Fixpoint Approach

Solving Dataflow Problems by Fixpoint Iteration

Remarks:

- (D, \sqsubseteq) being a **complete lattice** ensures that Φ_S is well defined
- Since (D, \sqsubseteq) is a **complete lattice satisfying ACC**, so is (D^n, \sqsubseteq^n) (where $(d_1, \dots, d_n) \sqsubseteq^n (d'_1, \dots, d'_n)$ iff $d_i \sqsubseteq d'_i$ for every $1 \leq i \leq n$)
- Monotonicity of transfer functions φ_i in (D, \sqsubseteq) implies **monotonicity of Φ_S** in (D^n, \sqsubseteq^n) (since \sqcup also monotonic)
- Thus the **(least) fixpoint is effectively computable** by iteration:

$$\text{fix}(\Phi_S) = \bigsqcup \{ \Phi_S^k(\perp_{D^n}) \mid k \in \mathbb{N} \}$$

where $\perp_{D^n} = \underbrace{(\perp_D, \dots, \perp_D)}_{n \text{ times}}$

- If height of (D, \sqsubseteq) is m
 - \implies height of (D^n, \sqsubseteq^n) is $m \cdot n$
 - \implies **fixpoint iteration requires at most $m \cdot n$ steps**

Efficient Fixpoint Computation

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A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_i in every step

Efficient Fixpoint Computation

A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_l in every step

\implies **redundant** if AI_l at no F -predecessor l' changed

Efficient Fixpoint Computation

A Worklist Algorithm I

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⇒ **redundant** if AI_l at no F -predecessor l' changed

⇒ optimisation by **worklist** over control-flow edges

Efficient Fixpoint Computation

A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_l in every step

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Algorithm 5.1 (Worklist algorithm)

Input: *dataflow system* $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$

Efficient Fixpoint Computation

A Worklist Algorithm I

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Algorithm 5.1 (Worklist algorithm)

Input: *dataflow system* $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$

Variables: $W \in (Lab \times Lab)^*$, $\{AI_l \in D \mid l \in Lab\}$

Efficient Fixpoint Computation

A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_l in every step

⇒ **redundant** if $AI_{l'}$ at no F -predecessor l' changed

⇒ optimisation by **worklist** over control-flow edges

Algorithm 5.1 (Worklist algorithm)

Input: *dataflow system* $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$

Variables: $W \in (Lab \times Lab)^*$, $\{AI_l \in D \mid l \in Lab\}$

Procedure: $W := \varepsilon$; **for** $(l, l') \in F$ **do** $W := W \cdot (l, l')$; % Initialise W

for $l \in Lab$ **do**

if $l \in E$ **then** $AI_l := \iota$ **else** $AI_l := \perp_D$; % Initialise AI

Efficient Fixpoint Computation

A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_I in every step

⇒ **redundant** if $AI_{I'}$ at no F -predecessor I' changed

⇒ optimisation by **worklist** over control-flow edges

Algorithm 5.1 (Worklist algorithm)

```
Input: dataflow system  $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ 
Variables:  $W \in (Lab \times Lab)^*$ ,  $\{AI_I \in D \mid I \in Lab\}$ 
Procedure:  $W := \varepsilon$ ; for  $(I, I') \in F$  do  $W := W \cdot (I, I')$ ;           % Initialise  $W$ 
           for  $I \in Lab$  do
               if  $I \in E$  then  $AI_I := \iota$  else  $AI_I := \perp_D$ ;           % Initialise  $AI$ 
           while  $W \neq \varepsilon$  do
                $(I, I') := \mathbf{head}(W)$ ;  $W := \mathbf{tail}(W)$ ;                       % Next control-flow edge
               if  $\varphi_I(AI_I) \not\sqsubseteq AI_{I'}$  then                               % Fixpoint not yet reached
                    $AI_{I'} := AI_{I'} \sqcup \varphi_I(AI_I)$ ;                       % Update analysis information
                   for  $(I', I'') \in F$  do
                       if  $(I', I'')$  not in  $W$  then  $W := (I', I'') \cdot W$ ; % Propagate modification
```

Efficient Fixpoint Computation

A Worklist Algorithm I

Observation: fixpoint iteration re-computes every AI_I in every step

⇒ **redundant** if $AI_{I'}$ at no F -predecessor I' changed

⇒ optimisation by **worklist** over control-flow edges

Algorithm 5.1 (Worklist algorithm)

```
Input: dataflow system  $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ 
Variables:  $W \in (Lab \times Lab)^*$ ,  $\{AI_I \in D \mid I \in Lab\}$ 
Procedure:  $W := \varepsilon$ ; for  $(I, I') \in F$  do  $W := W \cdot (I, I')$ ;           % Initialise  $W$ 
           for  $I \in Lab$  do
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           while  $W \neq \varepsilon$  do
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                   for  $(I', I'') \in F$  do
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Output:  $\{AI_I \mid I \in Lab\}$ 
```


Efficient Fixpoint Computation

A Worklist Algorithm II

Example 5.2 (Worklist algorithm)

Available Expression analysis for $c = [x := a+b]^1;$
(cf. Examples 2.9 and 4.11) $[y := a*b]^2;$
 $\text{while } [y > a+b]^3 \text{ do}$
 $[a := a+1]^4;$
 $[x := a+b]^5$
 end

Transfer functions: $\varphi_1(A) = A \cup \{a+b\}$
 $\varphi_2(A) = A \cup \{a*b\}$
 $\varphi_3(A) = A \cup \{a+b\}$
 $\varphi_4(A) = A \setminus \{a+b, a*b, a+1\}$
 $\varphi_5(A) = A \cup \{a+b\}$

Computation protocol: on the board

Efficient Fixpoint Computation

An “Optimisation”

Conjecture: it suffices to initialise worklist with **edges leaving extremal labels** (such that analysis information will propagate through CFG)

Efficient Fixpoint Computation

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But ...

Example 5.3 (Counterexample)

Live Variables analysis for $c = [x := 0]^1;$
 $[x := x + 1]^2;$
 $[x := 2]^3$

Solution: $LV_1 = \{x\}, LV_2 = \emptyset, LV_3 = \{x\}$

Efficient Fixpoint Computation

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Solution: $LV_1 = \{x\}, LV_2 = \emptyset, LV_3 = \{x\}$

“Optimised” worklist algorithm:

W	LV_1	LV_2	LV_3
$(3, 2)$	\emptyset	\emptyset	$\{x\}$

Efficient Fixpoint Computation

An “Optimisation”

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Solution: $LV_1 = \{x\}, LV_2 = \emptyset, LV_3 = \{x\}$

“Optimised” worklist algorithm:

W	LV_1	LV_2	LV_3
$(3, 2)$	\emptyset	\emptyset	$\{x\}$
ε	\emptyset	\emptyset	$\{x\}$

\Rightarrow **wrong result!**

Correctness of Worklist Algorithm

Theorem 5.4 (Correctness of worklist algorithm)

Given a dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$, Algorithm 5.1 always terminates and computes $\text{fix}(\Phi_S)$.

Efficient Fixpoint Computation

Correctness of Worklist Algorithm

Theorem 5.4 (Correctness of worklist algorithm)

Given a dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$, Algorithm 5.1 always terminates and computes $\text{fix}(\Phi_S)$.

Proof.

see [Nielson/Nielson/Hankin 2005, p. 75 ff] □

The MOP Solution

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The MOP Solution I

- Other **solution method** for dataflow systems

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- MOP = **Meet Over all Paths**

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- MOP = **Meet Over all Paths**
- Analysis information for block B^l
 - = **least upper bound over all paths leading to l**
 - = **most precise** information for l (“reference solution”)

The MOP Solution

The MOP Solution I

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Definition 5.5 (Paths)

Let $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ be a dataflow system. For every $l \in Lab$, the set of **paths up to l** is given by

$$Path(l) := \{[l_1, \dots, l_{k-1}] \mid k \geq 1, l_1 \in E, (l_i, l_{i+1}) \in F \text{ for every } 1 \leq i < k, l_k = l\}.$$

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For a path $\pi = [l_1, \dots, l_{k-1}] \in Path(l)$, we define the **transfer function** $\varphi_\pi : D \rightarrow D$ by

$$\varphi_\pi := \varphi_{l_{k-1}} \circ \dots \circ \varphi_{l_1} \circ \text{id}_D$$

(so that $\varphi_{[]} = \text{id}_D$).

The MOP Solution

The MOP Solution II

Definition 5.6 (MOP solution)

Let $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ be a dataflow system where $Lab = \{l_1, \dots, l_n\}$. The **MOP solution** for S is determined by

$$\text{mop}(S) := (\text{mop}(l_1), \dots, \text{mop}(l_n)) \in D^n$$

where, for every $l \in Lab$,

$$\text{mop}(l) := \bigsqcup \{ \varphi_\pi(\iota) \mid \pi \in Path(l) \}.$$

The MOP Solution

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Remark:

- $Path(l)$ is generally infinite
- ⇒ not clear how to compute $\text{mop}(l)$

The MOP Solution

The MOP Solution II

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Remark:

- $Path(l)$ is generally infinite
- ⇒ not clear how to compute $\text{mop}(l)$
- In fact: MOP solution generally undecidable (later)

The MOP Solution

The MOP Solution III

Example 5.7 (Live Variables; cf. Examples 2.12 and 4.12)

```
c = [x := 2]1;  
    [y := 4]2;  
    [x := 1]3;  
    if [y > 0]4 then  
        [z := x]5  
    else  
        [z := y*y]6;  
    [x := z]7
```

The MOP Solution

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$\implies Path(1) = \{[7, 5, 4, 3, 2],$
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$$\implies \text{mop}(1) = \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota)$$

$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2], [7, 6, 4, 3, 2]\}$$

The MOP Solution

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$$\begin{aligned} c = & [x := 2]^1; & \implies \text{mop}(1) = & \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota) \\ & [y := 4]^2; & & = \varphi_2(\varphi_3(\varphi_4(\varphi_5(\varphi_7(\{x, y, z\})))) \sqcup \\ & [x := 1]^3; & & \varphi_2(\varphi_3(\varphi_4(\varphi_6(\varphi_7(\{x, y, z\})))) \\ & \text{if } [y > 0]^4 \text{ then} \\ & \quad [z := x]^5 \\ & \text{else} \\ & \quad [z := y*y]^6; \\ & \quad [x := z]^7 \end{aligned}$$
$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2], [7, 6, 4, 3, 2]\}$$

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$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2], [7, 6, 4, 3, 2]\}$$

The MOP Solution

The MOP Solution III

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$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2], [7, 6, 4, 3, 2]\}$$

The MOP Solution

The MOP Solution III

Example 5.7 (Live Variables; cf. Examples 2.12 and 4.12)

$$\begin{aligned} c &= [x := 2]^1; & \implies \text{mop}(1) &= \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota) \\ & [y := 4]^2; & &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\varphi_7(\{x, y, z\})))) \sqcup \\ & [x := 1]^3; & & \varphi_2(\varphi_3(\varphi_4(\varphi_6(\varphi_7(\{x, y, z\})))) \\ & \text{if } [y > 0]^4 \text{ then} & &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\{y, z\}))) \sqcup \\ & \quad [z := x]^5 & & \varphi_2(\varphi_3(\varphi_4(\varphi_6(\{y, z\})))) \\ & \text{else} & &= \varphi_2(\varphi_3(\varphi_4(\{x, y\})) \sqcup \\ & \quad [z := y*y]^6; & & \varphi_2(\varphi_3(\varphi_4(\{y\}))) \\ & \quad [x := z]^7 & &= \varphi_2(\varphi_3(\{x, y\})) \sqcup \varphi_2(\varphi_3(\{y\})) \end{aligned}$$
$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2], [7, 6, 4, 3, 2]\}$$

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The MOP Solution

The MOP Solution III

Example 5.7 (Live Variables; cf. Examples 2.12 and 4.12)

$$\begin{aligned} c &= [x := 2]^1; \\ & [y := 4]^2; \\ & [x := 1]^3; \\ & \text{if } [y > 0]^4 \text{ then} \\ & \quad [z := x]^5 \\ & \text{else} \\ & \quad [z := y*y]^6; \\ & \quad [x := z]^7 \\ \implies \text{Path}(1) &= \{[7, 5, 4, 3, 2], \\ & \quad [7, 6, 4, 3, 2]\} \end{aligned} \quad \implies \text{mop}(1) = \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota)$$
$$\begin{aligned} &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\varphi_7(\{x, y, z\})))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\varphi_6(\varphi_7(\{x, y, z\})))) \\ &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\{y, z\}))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\varphi_6(\{y, z\})))) \\ &= \varphi_2(\varphi_3(\varphi_4(\{x, y\}))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\{y\}))) \\ &= \varphi_2(\varphi_3(\{x, y\})) \sqcup \varphi_2(\varphi_3(\{y\})) \\ &= \varphi_2(\{y\}) \sqcup \varphi_2(\{y\}) \\ &= \emptyset \sqcup \emptyset \end{aligned}$$

The MOP Solution

The MOP Solution III

Example 5.7 (Live Variables; cf. Examples 2.12 and 4.12)

$$\begin{aligned} c &= [x := 2]^1; \\ & \quad [y := 4]^2; \\ & \quad [x := 1]^3; \\ & \quad \text{if } [y > 0]^4 \text{ then} \\ & \quad \quad [z := x]^5 \\ & \quad \text{else} \\ & \quad \quad [z := y*y]^6; \\ & \quad \quad [x := z]^7 \\ \implies \text{Path}(1) &= \{[7, 5, 4, 3, 2], \\ & \quad [7, 6, 4, 3, 2]\} \\ \implies \text{mop}(1) &= \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota) \\ &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\varphi_7(\{x, y, z\})))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\varphi_6(\varphi_7(\{x, y, z\})))) \\ &= \varphi_2(\varphi_3(\varphi_4(\varphi_5(\{y, z\}))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\varphi_6(\{y, z\})))) \\ &= \varphi_2(\varphi_3(\varphi_4(\{x, y\}))) \sqcup \\ & \quad \varphi_2(\varphi_3(\varphi_4(\{y\}))) \\ &= \varphi_2(\varphi_3(\{x, y\})) \sqcup \varphi_2(\varphi_3(\{y\})) \\ &= \varphi_2(\{y\}) \sqcup \varphi_2(\{y\}) \\ &= \emptyset \sqcup \emptyset \\ &= \emptyset \quad (\text{same as } \text{fix}(\Phi_S)[1] \text{ – Ex. 4.12}) \end{aligned}$$

Another Analysis: Constant Propagation

Outline of Lecture 5

Recap: The Fixpoint Approach

Efficient Fixpoint Computation

The MOP Solution

Another Analysis: Constant Propagation

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Goal of Constant Propagation Analysis

Constant Propagation Analysis

The goal of **Constant Propagation Analysis** is to determine, for each program point, whether a variable has a constant value whenever execution reaches that point.

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Example 5.8 (Constant Propagation Analysis)

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[x := 1]1; [y := 1]2; [z := 1]3;  
while [z > 0]4 do  
  [w := x+y]5;  
  if [w = 2]6 then  
    [x := y+2]7  
  end  
end  
end
```

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- $y = z = 1$ at labels 4–7

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- w, x not constant at labels 4–7

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  end  
end
```

- $y = z = 1$ at labels 4–7
- w, x not constant at labels 4–7
- possible optimisations:

```
[true]4  
[w := x+1]5  
[x := 3]7
```

Another Analysis: Constant Propagation

Formalising Constant Propagation Analysis I

The **dataflow system** $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ is given by

- set of labels $Lab := Lab_c$,
- extremal labels $E := \{\text{init}(c)\}$ (forward problem)
- flow relation $F := \text{flow}(c)$ (forward problem)

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- extremal labels $E := \{\text{init}(c)\}$ (forward problem)
- flow relation $F := \text{flow}(c)$ (forward problem)
- complete lattice (D, \sqsubseteq) where
 - $D := \{\delta \mid \delta : Var_c \rightarrow \mathbb{Z} \cup \{\perp, \top\}\}$
 - $\delta(x) = z \in \mathbb{Z}$: x has **constant value** z (i.e., possible values in $\{z\}$)
 - $\delta(x) = \perp$: x **undefined** (i.e., possible values in \emptyset)
 - $\delta(x) = \top$: x **overdefined** (i.e., possible values in \mathbb{Z})
 - $\sqsubseteq \subseteq D \times D$ defined by pointwise extension of $\perp \sqsubseteq z \sqsubseteq \top$ (for every $z \in \mathbb{Z}$)

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Example 5.9

$$Var_c = \{w, x, y, z\}, \delta_1 = (\underbrace{\perp}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z), \delta_2 = (\underbrace{3}_w, \underbrace{1}_x, \underbrace{4}_y, \underbrace{\top}_z)$$
$$\implies \delta_1 \sqcup \delta_2 = (\underbrace{3}_w, \underbrace{1}_x, \underbrace{\top}_y, \underbrace{\top}_z)$$

Another Analysis: Constant Propagation

Formalising Constant Propagation Analysis II

Dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ (continued):

- extremal value $\iota := \delta_{\top} \in D$ where $\delta_{\top}(x) := \top$ for every $x \in Var_c$ (i.e., every x has (unknown) default value)

Another Analysis: Constant Propagation

Formalising Constant Propagation Analysis II

Dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$ (continued):

- extremal value $\iota := \delta_{\top} \in D$ where $\delta_{\top}(x) := \top$ for every $x \in Var_c$ (i.e., every x has (unknown) default value)
- transfer functions $\{\varphi_l \mid l \in Lab\}$ defined by

$$\varphi_l(\delta) := \begin{cases} \delta & \text{if } B' = \text{skip or } B' \in BExp \\ \delta[x \mapsto val_{\delta}(a)] & \text{if } B' = (x := a) \end{cases}$$

where

$$\begin{aligned} val_{\delta}(x) &:= \delta(x) \\ val_{\delta}(z) &:= z \end{aligned} \quad val_{\delta}(a_1 \text{ op } a_2) := \begin{cases} z_1 \text{ op } z_2 & \text{if } z_1, z_2 \in \mathbb{Z} \\ \perp & \text{if } z_1 = \perp \text{ or } z_2 = \perp \\ \top & \text{otherwise} \end{cases}$$

for $z_1 := val_{\delta}(a_1)$ and $z_2 := val_{\delta}(a_2)$

Another Analysis: Constant Propagation

Formalising Constant Propagation Analysis III

Example 5.10

If $\delta = (\underbrace{\perp}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z)$, then

$$\varphi_1(\delta) = \begin{cases} (\underbrace{0}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z) & \text{if } B^l = (w := 0) \\ (\underbrace{3}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z) & \text{if } B^l = (w := y+1) \\ (\underbrace{\perp}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z) & \text{if } B^l = (w := w+x) \\ (\underbrace{\top}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z) & \text{if } B^l = (w := z+2) \end{cases}$$