

Static Program Analysis

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

Thomas Noll

Lehrstuhl für Informatik 2
(Software Modeling and Verification)



noll@cs.rwth-aachen.de

<http://moves.rwth-aachen.de/teaching/ws-1415/spa/>

Winter Semester 2014/15

Wanted: Software Engineering HiWis

- What we offer: work in
 - **EU project D-MILS**
 - Dependability and Security of Distributed Information and Communication Infrastructures
 - <http://www.d-mils.org/>
 - Goal: [design and] implementation of high-level specification language
 - **ESA project CATSY**
 - Catalogue of System and Software Properties
 - Successor of COMPASS project (<http://compass.informatik.rwth-aachen.de>)
 - goal: support early V & V activities in model-based system development
- What we expect: prospective candidates
 - like formal methods (model checking, program/model transformations)
 - program efficiently (Python)
 - work 9–19 hrs/week
- Contact: Thomas Noll (noll@cs.rwth-aachen.de)



- 1 Recap: The Fixpoint Approach
- 2 Uniqueness of Solutions
- 3 Efficient Fixpoint Computation
- 4 The MOP Solution
- 5 Another Analysis: Constant Propagation

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

$$\Phi^0(d) := d \text{ and } \Phi^{k+1}(d) := \Phi(\Phi^k(d)).$$

Function requirements for dataflow analysis

All transfer functions must be a **monotonic**

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 $\varphi_l : D \rightarrow D$.

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

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**

- 1 Recap: The Fixpoint Approach
- 2 Uniqueness of Solutions
- 3 Efficient Fixpoint Computation
- 4 The MOP Solution
- 5 Another Analysis: Constant Propagation

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Example 5.1 (Available Expressions)

```
[z := x+y]1;  
while [true]2 do  
  [skip]3;
```

Uniqueness of Solutions I

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Example 5.1 (Available Expressions)

```
[z := x+y]1;  
while [true]2 do  
  [skip]3;
```

$$\begin{aligned}\implies AE_1 &= \emptyset \\ AE_2 &= (AE_1 \cup \{x+y\}) \cap AE_3 \\ AE_3 &= AE_2\end{aligned}$$

Uniqueness of Solutions I

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Example 5.1 (Available Expressions)

```
[z := x+y]1;  
while [true]2 do  
  [skip]3;
```

$$\begin{aligned} \implies AE_1 &= \emptyset \\ AE_2 &= (AE_1 \cup \{x+y\}) \cap AE_3 \\ AE_3 &= AE_2 \\ \implies AE_1 &= \emptyset \\ AE_2 &= \{x+y\} \cap AE_3 \\ AE_3 &= AE_2 \end{aligned}$$

Uniqueness of Solutions I

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Example 5.1 (Available Expressions)

$[z := x+y]^1;$	$\implies AE_1 = \emptyset$
$\text{while } [\text{true}]^2 \text{ do}$	$AE_2 = (AE_1 \cup \{x+y\}) \cap AE_3$
$[\text{skip}]^3;$	$AE_3 = AE_2$
	$\implies AE_1 = \emptyset$
	$AE_2 = \{x+y\} \cap AE_3$
	$AE_3 = AE_2$
\implies Solutions:	$AE_1 = AE_2 = AE_3 = \emptyset$ or
	$AE_1 = \emptyset, AE_2 = AE_3 = \{x+y\}$

Uniqueness of Solutions I

Observation: (non-minimal) solutions of dataflow equation systems are **not always unique**.

Example 5.1 (Available Expressions)

$[z := x+y]^1;$ $\implies AE_1 = \emptyset$
 $\text{while } [true]^2 \text{ do}$ $AE_2 = (AE_1 \cup \{x+y\}) \cap AE_3$
 $\quad [skip]^3;$ $AE_3 = AE_2$

$\implies AE_1 = \emptyset$
 $AE_2 = \{x+y\} \cap AE_3$
 $AE_3 = AE_2$

\implies **Solutions:** $AE_1 = AE_2 = AE_3 = \emptyset$ or
 $AE_1 = \emptyset, AE_2 = AE_3 = \{x+y\}$

Here: **greatest** solution $\{x+y\}$ (maximal potential for optimisation)

Example 5.2 (Live Variables)

```
while [x>1]1 do  
  [skip]2;  
  [x := x+1]3;  
  [y := 0]4
```

Example 5.2 (Live Variables)

<code>while [x>1]¹ do</code>	\implies	$LV_1 = LV_2 \cup (LV_3 \cup \{x\})$
<code> [skip]²;</code>		$LV_2 = LV_1 \cup \{x\}$
<code> [x := x+1]³;</code>		$LV_3 = LV_4 \setminus \{y\}$
<code> [y := 0]⁴</code>		$LV_4 = \{x, y\}$

Example 5.2 (Live Variables)

```
while [x>1]1 do           ⇒ LV1 = LV2 ∪ (LV3 ∪ {x})
  [skip]2;                ⇒ LV2 = LV1 ∪ {x}
[x := x+1]3;            ⇒ LV3 = LV4 \ {y}
[y := 0]4                ⇒ LV4 = {x, y}
                               ⇒ LV3 = {x}
```

Example 5.2 (Live Variables)

<code>while [x>1]¹ do</code>	\implies	$LV_1 = LV_2 \cup (LV_3 \cup \{x\})$
<code> [skip]²;</code>		$LV_2 = LV_1 \cup \{x\}$
<code> [x := x+1]³;</code>		$LV_3 = LV_4 \setminus \{y\}$
<code> [y := 0]⁴</code>		$LV_4 = \{x, y\}$
	\implies	$LV_3 = \{x\}$
	\implies	$LV_1 = LV_2 \cup \{x\}$ $= LV_1 \cup \{x\}$

Example 5.2 (Live Variables)

while $[x > 1]^1$ do $\implies LV_1 = LV_2 \cup (LV_3 \cup \{x\})$
 $[\text{skip}]^2$; $LV_2 = LV_1 \cup \{x\}$
 $[x := x+1]^3$; $LV_3 = LV_4 \setminus \{y\}$
 $[y := 0]^4$ $LV_4 = \{x, y\}$

$\implies LV_3 = \{x\}$

$\implies LV_1 = LV_2 \cup \{x\}$
 $= LV_1 \cup \{x\}$

\implies **Solutions:** $LV_1 = LV_2 = (\{x\} \text{ or } \{x, y\})$,
 $LV_3 = \{x\}, LV_4 = \{x, y\}$

Example 5.2 (Live Variables)

`while [x>1]1 do` $\implies LV_1 = LV_2 \cup (LV_3 \cup \{x\})$
 `[skip]2;` $LV_2 = LV_1 \cup \{x\}$
 `[x := x+1]3;` $LV_3 = LV_4 \setminus \{y\}$
 `[y := 0]4` $LV_4 = \{x, y\}$

$$\implies LV_3 = \{x\}$$

$$\begin{aligned} \implies LV_1 &= LV_2 \cup \{x\} \\ &= LV_1 \cup \{x\} \end{aligned}$$

\implies **Solutions:** $LV_1 = LV_2 = (\{x\} \text{ or } \{x, y\})$,
 $LV_3 = \{x\}, LV_4 = \{x, y\}$

Here: **least** solution $\{x\}$ (maximal potential for optimisation)

- 1 Recap: The Fixpoint Approach
- 2 Uniqueness of Solutions
- 3 Efficient Fixpoint Computation**
- 4 The MOP Solution
- 5 Another Analysis: Constant Propagation

A Worklist Algorithm I

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

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

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**

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**

Algorithm 5.3 (Worklist algorithm)

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

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**

Algorithm 5.3 (Worklist algorithm)

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

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

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**

Algorithm 5.3 (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** % *Initialise AI*
 if $l \in E$ **then** $AI_l := \iota$ **else** $AI_l := \perp_D$;

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**

Algorithm 5.3 (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** % *Initialise AI*
 if $I \in E$ **then** $AI_I := \iota$ **else** $AI_I := \perp_D$;
while $W \neq \varepsilon$ **do**
 $(I, I') := \text{head}(W)$; $W := \text{tail}(W)$;
 if $\varphi_I(AI_I) \not\sqsubseteq AI_{I'}$ **then** % *Fixpoint not yet reached*
 $AI_{I'} := AI_{I'} \sqcup \varphi_I(AI_I)$;
 for $(I', I'') \in F$ **do**
 if (I', I'') *not in W* **then** $W := (I', I'') \cdot W$;

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**

Algorithm 5.3 (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** % *Initialise AI*
 if $I \in E$ **then** $AI_I := \iota$ **else** $AI_I := \perp_D$;
 while $W \neq \varepsilon$ **do**
 $(I, I') := \text{head}(W)$; $W := \text{tail}(W)$;
 if $\varphi_I(AI_I) \not\sqsubseteq AI_{I'}$ **then** % *Fixpoint not yet reached*
 $AI_{I'} := AI_{I'} \sqcup \varphi_I(AI_I)$;
 for $(I', I'') \in F$ **do**
 if (I', I'') *not in W* **then** $W := (I', I'') \cdot W$;

Output: $\{AI_I \mid I \in Lab\}$

An “Optimisation”

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

An “Optimisation”

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

But ...

Example 5.5 (Counterexample)

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

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

An “Optimisation”

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

But ...

Example 5.5 (Counterexample)

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

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

“Optimised” worklist algorithm:

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

An “Optimisation”

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

But ...

Example 5.5 (Counterexample)

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

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

“Optimised” worklist algorithm:

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

\Rightarrow **wrong result!**

Properties of the algorithm:

Theorem 5.6 (Correctness of worklist algorithm)

Given a dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$, Algorithm 5.3 always terminates and computes $\text{fix}(\Phi_S)$.

Correctness of Worklist Algorithm

Properties of the algorithm:

Theorem 5.6 (Correctness of worklist algorithm)

Given a dataflow system $S = (Lab, E, F, (D, \sqsubseteq), \iota, \varphi)$, Algorithm 5.3 always terminates and computes $\text{fix}(\Phi_S)$.

Proof.

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

- 1 Recap: The Fixpoint Approach
- 2 Uniqueness of Solutions
- 3 Efficient Fixpoint Computation
- 4 The MOP Solution**
- 5 Another Analysis: Constant Propagation

The MOP Solution I

- Other **solution method** for dataflow systems

The MOP Solution I

- Other **solution method** for dataflow systems
- MOP = **Meet Over all Paths**

The MOP Solution I

- Other **solution method** for dataflow systems
- 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 I

- Other **solution method** for dataflow systems
- 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”)

Definition 5.7 (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

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

The MOP Solution I

- Other **solution method** for dataflow systems
- 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”)

Definition 5.7 (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

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

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

Definition 5.8 (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 $l \in Lab$,

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

Definition 5.8 (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 $l \in Lab$,

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

Remark:

- $Path(l)$ is generally infinite

\Rightarrow not clear how to compute $\text{mop}(l)$

Definition 5.8 (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 $l \in Lab$,

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

Remark:

- $Path(l)$ is generally infinite

\Rightarrow not clear how to compute $\text{mop}(l)$

- In fact: MOP solution generally undecidable (later)

Example 5.9 (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
```

Example 5.9 (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
```

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

$$c = [x := 2]^1; \quad \implies \text{mop}(1) = \varphi_{[7,5,4,3,2]}(\iota) \sqcup \varphi_{[7,6,4,3,2]}(\iota)$$
$$[y := 4]^2;$$
$$[x := 1]^3;$$
$$\text{if } [y > 0]^4 \text{ then}$$
$$[z := x]^5$$
$$\text{else}$$
$$[z := y*y]^6;$$
$$[x := z]^7$$
$$\implies \text{Path}(1) = \{[7, 5, 4, 3, 2],$$
$$[7, 6, 4, 3, 2]\}$$

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

Example 5.9 (Live Variables; cf. Examples 2.12 and 4.12)

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

- 1 Recap: The Fixpoint Approach
- 2 Uniqueness of Solutions
- 3 Efficient Fixpoint Computation
- 4 The MOP Solution
- 5 Another Analysis: Constant Propagation

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.

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.

Used for **Constant Folding**: replace reference to variable by constant value and evaluate constant expressions

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.

Used for **Constant Folding**: replace reference to variable by constant value and evaluate constant expressions

Example 5.10 (Constant Propagation Analysis)

```
[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
```

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.

Used for **Constant Folding**: replace reference to variable by constant value and evaluate constant expressions

Example 5.10 (Constant Propagation Analysis)

```
[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
```

- $y = z = 1$ at labels 4–7

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.

Used for **Constant Folding**: replace reference to variable by constant value and evaluate constant expressions

Example 5.10 (Constant Propagation Analysis)

```
[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
```

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

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.

Used for **Constant Folding**: replace reference to variable by constant value and evaluate constant expressions

Example 5.10 (Constant Propagation Analysis)

```
[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
```

- $y = z = 1$ at labels 4–7
- w, x not constant at labels 4–7
- possible optimisations:
 $[\text{true}]^4 [w := x+1]^5 [x := 3]^7$

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 $E := \{init(c)\}$ (forward problem),
- flow relation $F := flow(c)$ (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
 - $\delta(x) = \perp$: x **undefined**
 - $\delta(x) = \top$: x **overdefined** (i.e., several possible values)
 - $\sqsubseteq \subseteq D \times D$ defined by pointwise extension of $\perp \sqsubseteq z \sqsubseteq \top$ (for every $z \in \mathbb{Z}$)

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 $E := \{init(c)\}$ (forward problem),
- flow relation $F := flow(c)$ (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
 - $\delta(x) = \perp$: x **undefined**
 - $\delta(x) = \top$: x **overdefined** (i.e., several possible values)
 - $\sqsubseteq \subseteq D \times D$ defined by pointwise extension of $\perp \sqsubseteq z \sqsubseteq \top$ (for every $z \in \mathbb{Z}$)

Example 5.11

$$Var_c = \{w, x, y, z\},$$

$$\delta_1 = (\underbrace{\perp}_w, \underbrace{1}_x, \underbrace{2}_y, \underbrace{\top}_z), \quad \delta_2 = (\underbrace{3}_w, \underbrace{1}_x, \underbrace{4}_y, \underbrace{\top}_z)$$

$$\implies \delta_1 \sqcup \delta_2 = (\underbrace{3}_w, \underbrace{1}_x, \underbrace{\top}_y, \underbrace{\top}_z)$$

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

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

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 $z_2 := val_{\delta}(a_2)$

Example 5.12

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

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