

Static Program Analysis

Lecture 4: Dataflow Analysis III (The Framework)

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- 1 Repetition: Heading for a Dataflow Analysis Framework
- 2 Order-Theoretic Foundations: the Function
- 3 Application to Dataflow Analysis

Similarities between Analysis Problems

- **Observation:** the analyses presented so far have some **similarities**

⇒ Look for underlying **framework**

- **Advantage:** possibility for designing (efficient) **generic algorithms for solving dataflow equations**
- **Overall pattern:** for $c \in Cmd$ and $l \in L_c$, the **analysis information** (AI) is described by **equations** of the form

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

where

- the set of **extremal labels**, E , is $\{\text{init}(c)\}$ or $\{\text{final}(c)\}$
- ι specifies the **extremal analysis information**
- the **combination operator**, $\bigsqcup$, is $\cap$ or $\cup$
- $\varphi_{l'}$ denotes the **transfer function** of block $B_{l'}$
- the **flow relation** F is $\text{flow}(c)$ or $\text{flow}^R(c)$ ($:= \{(l', l) \mid (l, l') \in \text{flow}(c)\}$)

Goal: solve dataflow equation system by **fixpoint iteration**

- ① Characterize solution of equation system as **fixpoint** of a transformation
- ② Introduce **partial order** for comparing analysis results
- ③ Establish **least upper bound** as combination operator
- ④ Ensure **monotonicity** of transfer functions
- ⑤ Guarantee termination of fixpoint iteration by **ascending chain condition**
- ⑥ Optimize fixpoint iteration by **worklist algorithm**

- **Wanted:** solution of (dataflow) equation system
- **Problem:** recursive dependencies between dataflow variables
- **Idea:** characterize solution as fixpoint of transformation:

$$(AI_I = \tau_I)_{I \in L_c} \iff \Phi((AI_I)_{I \in L_c}) = (AI_I)_{I \in L_c}$$

where $\Phi((AI_I)_{I \in L_c}) := (\tau_I)_{I \in L_c}$

- **Approach:** approximate fixpoint by iteration

Partial Orders

The domain of analysis information usually forms a partial order where the ordering relation compares the “precision” of information.

Definition (Partial order)

A **partial order (PO)** $(D, \sqsubseteq)$ consists of a set D , called **domain**, and of a relation $\sqsubseteq \subseteq D \times D$ such that, for every $d_1, d_2, d_3 \in D$,

reflexivity: $d_1 \sqsubseteq d_1$

transitivity: $d_1 \sqsubseteq d_2$ and $d_2 \sqsubseteq d_3 \implies d_1 \sqsubseteq d_3$

antisymmetry: $d_1 \sqsubseteq d_2$ and $d_2 \sqsubseteq d_1 \implies d_1 = d_2$

It is called **total** if, in addition, always $d_1 \sqsubseteq d_2$ or $d_2 \sqsubseteq d_1$.

Example

- ① $(\mathbb{N}, \leq)$ is a total partial order
- ② $(\mathbb{N}, <)$ is not a partial order (since not reflexive)
- ③ (Live Variables) $(2^{Var_c}, \sqsubseteq)$ is a (non-total) partial order
- ④ (Available Expressions) $(2^{AExp_c}, \supseteq)$ is a (non-total) partial order

Upper Bounds

In the dataflow equation system, analysis information from several predecessors is combined by taking the least upper bound.

Definition ((Least) upper bound)

Let $(D, \sqsubseteq)$ be a partial order and $S \subseteq D$.

- 1 An element $d \in D$ is called an **upper bound** of S if $s \sqsubseteq d$ for every $s \in S$ (notation: $S \sqsubseteq d$).
- 2 An upper bound d of S is called **least upper bound (LUB)** or **supremum** of S if $d \sqsubseteq d'$ for every upper bound d' of S (notation: $d = \bigsqcup S$).

Example

- 1 $S \subseteq \mathbb{N}$ has a LUB in $(\mathbb{N}, \leq)$ iff it is finite
- 2 (Live Variables) $(D, \sqsubseteq) = (2^{Var_c}, \subseteq)$. Given $V_1, \dots, V_n \subseteq Var_c$,
$$\bigsqcup \{V_1, \dots, V_n\} = \bigcup \{V_1, \dots, V_n\}$$
- 3 (Avail. Expr.) $(D, \sqsubseteq) = (2^{AExp_c}, \supseteq)$. Given $A_1, \dots, A_n \subseteq AExp_c$,
$$\bigsqcup \{A_1, \dots, A_n\} = \bigcap \{A_1, \dots, A_n\}$$

Complete Lattices

Since $\{\varphi_{I'}(AI_{I'}) \mid (I', I) \in F\}$ can contain arbitrary elements, the existence of least upper bounds must be ensured for arbitrary subsets.

Definition (Complete lattice)

A **complete lattice** is a partial order $(D, \sqsubseteq)$ such that all subsets of D have least upper bounds. In this case,

$$\perp := \bigsqcup \emptyset$$

denotes the **least element** of D .

Example

- 1 $(\mathbb{N}, \leq)$ is not a complete lattice as, e.g., $\mathbb{N}$ does not have a LUB
- 2 (Live Variables)
 $(D, \sqsubseteq) = (2^{Var_c}, \sqsubseteq)$ is a complete lattice with $\perp = \emptyset$
- 3 (Available Expressions)
 $(D, \sqsubseteq) = (2^{AExp_c}, \supseteq)$ is a complete lattice with $\perp = AExp_c$

Chains are generated by the approximation of the analysis information in the fixpoint iteration.

Definition (Chain)

Let $(D, \sqsubseteq)$ be a partial order. A subset $S \subseteq D$ is called an **(ascending) chain** in D if, for every $s_1, s_2 \in S$,

$$s_1 \sqsubseteq s_2 \text{ or } s_2 \sqsubseteq s_1$$

(that is, S is a totally ordered subset of D).

Example

- 1 Every $S \subseteq \mathbb{N}$ is a chain in $(\mathbb{N}, \leq)$
- 2 $\{\emptyset, \{0\}, \{0, 1\}, \{0, 1, 2\}, \dots\}$ is a chain in $(2^{\mathbb{N}}, \subseteq)$
- 3 $\{\emptyset, \{0\}, \{1\}\}$ is not a chain in $(2^{\mathbb{N}}, \subseteq)$

The Ascending Chain Condition

Termination of fixpoint iteration is guaranteed by the Ascending Chain Condition.

Definition (Ascending Chain Condition)

A partial order $(D, \sqsubseteq)$ satisfies the **Ascending Chain Condition (ACC)** if each ascending chain $d_1 \sqsubseteq d_2 \sqsubseteq \dots$ eventually stabilizes, i.e., there exists $n \in \mathbb{N}$ such that $d_n = d_{n+1} = \dots$

Example

- ① $(\mathbb{N}, \leq)$ does not satisfy ACC
- ② (Live Variables) $(D, \sqsubseteq) = (2^{Var_c}, \subseteq)$ satisfies ACC since Var_c (unlike Var) is finite
- ③ (Available Expressions) $(D, \sqsubseteq) = (2^{AExp_c}, \supseteq)$ satisfies ACC since $AExp_c$ (unlike $AExp$) is finite

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Monotonicity of Functions

The monotonicity of transfer functions excludes “oscillating behavior” in fixpoint iteration.

Definition 4.1 (Monotonicity)

Let $(D, \sqsubseteq)$ and $(D', \sqsubseteq')$ be partial orders, and let $\Phi : D \rightarrow D'$. Φ is called **monotonic (w.r.t. $(D, \sqsubseteq)$ and $(D', \sqsubseteq')$)** if, for every $d_1, d_2 \in D$,

$$d_1 \sqsubseteq d_2 \implies \Phi(d_1) \sqsubseteq' \Phi(d_2).$$

Example 4.2

- 1 Let $T := \{S \subseteq \mathbb{N} \mid S \text{ finite}\}$. Then $\Phi_1 : T \rightarrow \mathbb{N} : S \mapsto \sum_{n \in S} n$ is monotonic w.r.t. $(2^{\mathbb{N}}, \subseteq)$ and $(\mathbb{N}, \leq)$.
- 2 $\Phi_2 : 2^{\mathbb{N}} \rightarrow 2^{\mathbb{N}} : S \mapsto \mathbb{N} \setminus S$ is not monotonic w.r.t. $(2^{\mathbb{N}}, \subseteq)$ (since, e.g., $\emptyset \subseteq \mathbb{N}$ but $\Phi_2(\emptyset) = \mathbb{N} \not\subseteq \Phi_2(\mathbb{N}) = \emptyset$).
- 3 (Live Variables) $(D, \sqsubseteq) = (D', \sqsubseteq') = (2^{\text{Var}_c}, \subseteq)$
Each transfer function $\varphi_{l'}(V) := (V \setminus \text{kill}_{\text{LV}}(B'')) \cup \text{gen}_{\text{LV}}(B'')$ is obviously monotonic
- 4 (Available Expressions) $(D, \sqsubseteq) = (D', \sqsubseteq') = (2^{\text{AExp}_c}, \supseteq)$ ditto

Definition 4.3 (Fixpoint)

Let D be some domain, $d \in D$, and $\Phi : D \rightarrow D$. If

$$\Phi(d) = d$$

then d is called a **fixpoint** of Φ .

Example 4.4

The (only) fixpoints of $\Phi : \mathbb{N} \rightarrow \mathbb{N} : n \mapsto n^2$ are 0 and 1

The Fixpoint Theorem I



Alfred Tarski (1901–1983)



Bronislaw Knaster (1893–1990)

Theorem 4.5 (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

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

Remark: $\text{ACC} \implies (\Phi^k(\perp) \mid k \in \mathbb{N})$ stabilizes at some $k_0 \in \mathbb{N}$ with $\text{fix}(\Phi) = \Phi^{k_0}(\perp)$ (where k_0 bounded by maximal chain length in $(D, \sqsubseteq)$)

The Fixpoint Theorem II

The proof of Theorem 4.5 requires the following lemma.

Lemma 4.6

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

$$\Phi(\bigsqcup S) = \bigsqcup \Phi(S)$$

Proof (Lemma 4.6).

on the board ☐

Proof (Theorem 4.5).

on the board ☐

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Definition 4.7 (Dataflow system)

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

- a finite set of (program) **labels** L (here: L_c),
- a set of **extremal labels** $E \subseteq L$ (here: $\{\text{init}(c)\}$ or $\text{final}(c)$),
- a **flow relation** $F \subseteq L \times L$ (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** $\{\varphi_I \mid I \in L\}$ of type $\varphi_I : D \rightarrow D$.

Example 4.8

Problem	Available Expressions	Live Variables
E	$\{\text{init}(c)\}$	$\text{final}(c)$
F	$\text{flow}(c)$	$\text{flow}^R(c)$
D	2^{AExp_c}	2^{Var_c}
$\sqsubseteq$	$\supseteq$	$\subseteq$
$\sqcup$	$\bigcap$	$\bigcup$
$\perp$	$AExp_c$	$\emptyset$
ι	$\emptyset$	Var_c
φ_I	$\varphi_I(d) = (d \setminus \text{kill}(B^I)) \cup \text{gen}(B^I)$	

Dataflow Systems and Fixpoints

Definition 4.9 (Dataflow equation system)

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

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

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

- $(d_1, \dots, d_n) \in D^n$ 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 4.10

$(d_1, \dots, d_n) \in D^n$ **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_l 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:

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

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

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

Example: Available Expressions

Example 4.11 (Available Expressions; cf. Example 2.9)

Program:

```
c = [x := a+b]1;  
    [y := a*b]2;  
    while [y > a+b]3 do  
        [a := a+1]4;  
        [x := a+b]5
```

Equation system:

$$\begin{aligned}AE_1 &= \emptyset \\AE_2 &= AE_1 \cup \{a+b\} \\AE_3 &= (AE_2 \cup \{a*b\}) \cap (AE_5 \cup \{a+b\}) \\AE_4 &= AE_3 \cup \{a+b\} \\AE_5 &= AE_4 \setminus \{a+b, a*b, a+1\}\end{aligned}$$

Fixpoint iteration:

i	1	2	3	4	5
0	$AExp_c$	$AExp_c$	$AExp_c$	$AExp_c$	$AExp_c$
1	$\emptyset$	$AExp_c$	$AExp_c$	$AExp_c$	$\emptyset$
2	$\emptyset$	$\{a+b\}$	$\{a+b\}$	$AExp_c$	$\emptyset$
3	$\emptyset$	$\{a+b\}$	$\{a+b\}$	$\{a+b\}$	$\emptyset$
4	$\emptyset$	$\{a+b\}$	$\{a+b\}$	$\{a+b\}$	$\emptyset$

Example: Live Variables

Example 4.12 (Live Variables; cf. Example 3.3)

Program:

```
[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
```

Equation system:

```
LV1 = LV2 \ {y}
LV2 = LV3 \ {x}
LV3 = LV4 ∪ {y}
LV4 = ((LV5 \ {z}) ∪ {x}) ∪ ((LV6 \ {z}) ∪ {y})
LV5 = (LV7 \ {x}) ∪ {z}
LV6 = (LV7 \ {x}) ∪ {z}
LV7 = {x, y, z}
```

Fixpoint iteration:

<i>i</i>	1	2	3	4	5	6	7
0	∅	∅	∅	∅	∅	∅	∅
1	∅	∅	{y}	{x, y}	{z}	{z}	{x, y, z}
2	∅	{y}	{x, y}	{x, y}	{y, z}	{y, z}	{x, y, z}
3	∅	{y}	{x, y}	{x, y}	{y, z}	{y, z}	{x, y, z}