

Static Program Analysis

Lecture 3: Dataflow Analysis II (Order-Theoretic Foundations)

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

Winter Semester 2014/15

- 1 Recap: Dataflow Analysis
- 2 Heading for a Dataflow Analysis Framework
- 3 Order-Theoretic Foundations: The Domain

Labelled Programs

- Goal: **localisation** of analysis information
- Dataflow information will be associated with
 - **skip** statements
 - assignments
 - tests in conditionals (**if**) and loops (**while**)
- Assume set of **labels** Lab with meta variable $l \in Lab$ (usually $Lab = \mathbb{N}$)

Definition (Labelled WHILE programs)

The **syntax of labelled WHILE programs** is defined by the following context-free grammar:

$$\begin{aligned} a &::= z \mid x \mid a_1 + a_2 \mid a_1 - a_2 \mid a_1 * a_2 \in AExp \\ b &::= t \mid a_1 = a_2 \mid a_1 > a_2 \mid \neg b \mid b_1 \wedge b_2 \mid b_1 \vee b_2 \in BExp \\ c &::= [\text{skip}]^l \mid [x := a]^l \mid c_1 ; c_2 \mid \\ &\quad \text{if } [b]^l \text{ then } c_1 \text{ else } c_2 \mid \text{while } [b]^l \text{ do } c \in Cmd \end{aligned}$$

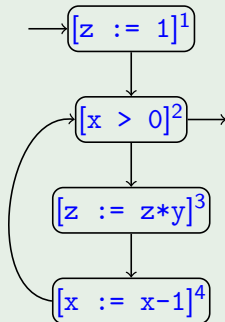
- All labels in $c \in Cmd$ assumed distinct, denoted by Lab_c
- Labelled fragments of c called **blocks**, denoted by Blk_c

Example

```
c = [z := 1]1;  
  while [x > 0]2 do  
    [z := z*y]3;  
    [x := x-1]4
```

```
init(c) = 1  
final(c) = {2}  
flow(c) = {(1, 2), (2, 3), (3, 4), (4, 2)}
```

Visualization by
(control) flow graph:



Goal of Available Expressions Analysis

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The goal of **Available Expressions Analysis** is to determine, for each program point, which (complex) expressions *must* have been computed, and not later modified, on all paths to the program point.

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replace subexpression by variable that contains up-to-date value
- Only interesting for non-trivial (i.e., complex) arithmetic expressions

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[x := a+b]1;  
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- **a+b** available at label 3

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- **a+b not available at label 5**

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```

- **a+b** available at label 3
- **a+b** not available at label 5
- possible optimization:
`while [y > x]3 do`

The Equation System I

- Analysis itself defined by setting up an **equation system**
- For each $l \in Lab_c$, $AE_l \subseteq CExp_c$ represents the **set of available expressions at the entry of block B^l**
- Formally, for $c \in Cmd$ with isolated entry:

$$AE_l = \begin{cases} \emptyset & \text{if } l = \text{init}(c) \\ \bigcap \{ \varphi_{l'}(AE_{l'}) \mid (l', l) \in \text{flow}(c) \} & \text{otherwise} \end{cases}$$

where $\varphi_{l'} : 2^{CExp_c} \rightarrow 2^{CExp_c}$ denotes the **transfer function** of block $B^{l'}$, given by

$$\varphi_{l'}(A) := (A \setminus \text{kill}_{AE}(B^{l'})) \cup \text{gen}_{AE}(B^{l'})$$

- Characterization of analysis:
 - flow-sensitive**: results depending on order of assignments
 - forward**: starts in $\text{init}(c)$ and proceeds downwards
 - must**: \bigcap in equation for AE_l
- Later: solution **not necessarily unique**
 \implies choose **greatest one**

The Equation System II

Reminder:

$$AE_l = \begin{cases} \emptyset & \text{if } l = \text{init}(c) \\ \bigcap \{\varphi_{l'}(AE_{l'}) \mid (l', l) \in \text{flow}(c)\} & \text{otherwise} \end{cases}$$
$$\varphi_{l'}(E) = (E \setminus \text{kill}_{AE}(B_{l'})) \cup \text{gen}_{AE}(B_{l'})$$

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$l \in Lab_c$	$\text{kill}_{AE}(B^l)$	$\text{gen}_{AE}(B^l)$
1	\emptyset	$\{a+b\}$
2	\emptyset	$\{a*b\}$
3	\emptyset	$\{a+b\}$
4	$\{a+b, a*b, a+1\}$	\emptyset
5	\emptyset	$\{a+b\}$

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Equations:

$$\begin{aligned} AE_1 &= \emptyset \\ AE_2 &= \varphi_1(AE_1) = AE_1 \cup \{a+b\} \\ AE_3 &= \varphi_2(AE_2) \cap \varphi_5(AE_5) \\ &= (AE_2 \cup \{a*b\}) \cap (AE_5 \cup \{a+b\}) \\ AE_4 &= \varphi_3(AE_3) = AE_3 \cup \{a+b\} \\ AE_5 &= \varphi_4(AE_4) = AE_4 \setminus \{a+b, a*b, a+1\} \end{aligned}$$

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Solution:

$$\begin{aligned} AE_1 &= \emptyset \\ AE_2 &= \{a+b\} \\ AE_3 &= \{a+b\} \\ AE_4 &= \{a+b\} \\ AE_5 &= \emptyset \end{aligned}$$

Live Variables Analysis

The goal of **Live Variables Analysis** is to determine, for each program point, which variables *may* be live at the exit from the point.

- A variable is called **live** at the exit from a block if there exists a path from the block to a use of the variable that does not re-define the variable
- All variables considered to be live at the **end** of the program (alternative: restriction to output variables)
- Can be used for **Dead Code Elimination**:
remove assignments to non-live variables

The Equation System I

- For each $l \in Lab_c$, $LV_l \subseteq Var_c$ represents the set of **live variables at the exit of block B^l**
- Formally, for a program $c \in Cmd$ with isolated exits:

$$LV_l = \begin{cases} Var_c & \text{if } l \in \text{final}(c) \\ \bigcup \{\varphi_{l'}(LV_{l'}) \mid (l, l') \in \text{flow}(c)\} & \text{otherwise} \end{cases}$$

where $\varphi_{l'} : 2^{Var_c} \rightarrow 2^{Var_c}$ denotes the **transfer function** of block $B^{l'}$, given by

$$\varphi_{l'}(V) := (V \setminus \text{kill}_{LV}(B^{l'})) \cup \text{gen}_{LV}(B^{l'})$$

- Characterization of analysis:
 - flow-sensitive: results depending on order of assignments
 - backward: starts in $\text{final}(c)$ and proceeds upwards
 - may: \bigcup in equation for LV_l
- Later: solution **not necessarily unique**
 \implies choose **least one**

The Equation System II

Reminder:

$$LV_I = \begin{cases} Var_c & \text{if } I \in \text{final}(c) \\ \bigcup \{ \varphi_{I'}(LV_{I'}) \mid (I, I') \in \text{flow}(c) \} & \text{otherwise} \end{cases}$$
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Example (LV equation system)

```
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
```

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$I \in Lab_c$ $\text{kill}_{LV}(B^I)$ $\text{gen}_{LV}(B^I)$

1	{x}	∅
2	{y}	∅
3	{x}	∅
4	∅	{y}
5	{z}	{x}
6	{z}	{y}
7	{x}	{z}

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```

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$l \in Lab_c$ $\text{kill}_{LV}(B^l)$ $\text{gen}_{LV}(B^l)$

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Solution:

$$\begin{aligned} LV_1 &= \emptyset \\ LV_2 &= \{y\} \\ LV_3 &= \{x, y\} \\ LV_4 &= \{x, y\} \\ LV_5 &= \{y, z\} \\ LV_6 &= \{y, z\} \\ LV_7 &= \{x, y, z\} \end{aligned}$$

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⇒ Look for underlying **framework**

- **Advantage:** possibility for designing (efficient) **generic algorithms for solving dataflow equations**
- **Overall pattern:** for $c \in \text{Cmd}$ and $l \in \text{Lab}_c$, 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 $\text{flow}^R(c) (:= \{(l', l) \mid (l, l') \in \text{flow}(c)\})$

- **Direction of information flow:**

- **forward:**

- $F = \text{flow}(c)$
 - AI_l concerns entry of B^l
 - c has isolated entry

- **backward:**

- $F = \text{flow}^R(c)$
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- **Quantification over paths:**
 - **may:**
 - $\sqcup = \cup$
 - property satisfied by some path
 - interested in least solution (later)
 - **must:**
 - $\sqcap = \cap$
 - property satisfied by all paths
 - interested in greatest solution (later)

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- ① Characterize solution of equation system as **fixpoint** of a transformation

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- ⑤ Guarantee termination of fixpoint iteration by **ascending chain condition**
- ⑥ Optimize fixpoint iteration by **worklist algorithm**

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- **Idea:** characterize solution as **fixpoint** of transformation:

$$(A|_l = \tau_l)_{l \in Lab_c} \iff \Phi((A|_l)_{l \in Lab_c}) = (A|_l)_{l \in Lab_c}$$

where $\Phi((A|_l)_{l \in Lab_c}) := (\tau_l)_{l \in Lab_c}$

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- **Problem:** recursive dependencies between dataflow variables
- **Idea:** characterize solution as fixpoint of transformation:

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where $\Phi((A_l)_{l \in Lab_c}) := (\tau_l)_{l \in Lab_c}$

- **Approach:** approximate fixpoint by iteration

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

Definition 3.1 (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 $d_2 \sqsubseteq d_1$.

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

Example 3.2

- 1 (\mathbb{N}, \leq) is a total partial order
- 2 $(\mathbb{N}, <)$ is not a partial order (since not reflexive)

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

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Complete Lattices

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

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

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Duality in Complete Lattices

- **Dual** concept of least upper bound: greatest lower bound
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The Ascending Chain Condition I

Termination of fixpoint iteration is guaranteed by the following condition.

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Notes:

- The finite height property implies ACC, but not vice versa (as there might be non-stabilizing descending chains)
- The complete lattice and ACC properties are orthogonal

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Domain requirements for dataflow analysis

(D, \sqsubseteq) must be a **complete lattice satisfying ACC**