

# Compiler Construction

## Lecture 25: Code Optimization II

### (The Dataflow Analysis Framework)

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- ① Repetition: Available Expression and Live Variables Analysis
- ② The Dataflow Analysis Framework
- ③ Solving Dataflow Equation Systems
- ④ Uniqueness of Solutions

- Given  $C \in Cmd$ ,  $\text{Lab}_C / \text{Blk}_C / AExp_C$  denote the sets of all labels/blocks/complex arithmetic expressions occurring in  $C$ , respectively
- An expression  $A$  is **killed** in a block  $\beta$  if any of the variables in  $A$  is modified in  $\beta$
- Formally:  $\text{kill}_{AE} : \text{Blk}_C \rightarrow 2^{AExp_C}$  is defined by  
$$\text{kill}_{AE}([I := A]^l) := \{A' \in AExp_C \mid I \in \text{FV}(A')\}$$
$$\text{kill}_{AE}([B]^l) := \emptyset$$
- An expression  $A$  is **generated** in a block  $\beta$  if it is evaluated in and none of its variables are modified by  $\beta$
- Formally:  $\text{gen}_{AE} : \text{Blk}_C \rightarrow 2^{AExp_C}$  is defined by  
$$\text{gen}_{AE}([I := A]^l) := \{A \mid I \notin \text{FV}(A)\}$$
$$\text{gen}_{AE}([B]^l) := AExp_B$$

- Analysis itself defined by setting up an **equation system**
- For each  $l \in Lab_C$ ,  $AE_l \subseteq AExp_C$  represents the **set of available expressions at the entry of block  $\beta^l$**
- 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  $\varphi_{l'} : 2^{AExp_C} \rightarrow 2^{AExp_C}$  denotes the **transfer function** of block  $\beta^{l'}$ , given by

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

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

- A variable on the left-hand side of an assignment is **killed** by the assignment; tests do not kill
- Formally:  $\text{kill}_{LV} : Blk_C \rightarrow 2^{Var_C}$  is defined by  
 $\text{kill}_{LV}([I := A]^l) := \{I\}$   
 $\text{kill}_{LV}([B]^l) := \emptyset$
- Every reading access **generates** a live variable
- Formally:  $\text{gen}_{LV} : Blk_C \rightarrow 2^{Var_C}$  is defined by  
 $\text{gen}_{LV}([I := A]^l) := \text{FV}(A)$   
 $\text{gen}_{LV}([B]^l) := \text{FV}(B)$

# The Live Variables Equation System

- For each  $l \in Lab_C$ ,  $LV_l \subseteq Var_c$  represents the set of **live variables at the exit of block  $\beta^l$**

- 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  $\beta^{l'}$ , given by

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

- Characterization of analysis:

**backward**: starts in  $\text{final}(C)$  and proceeds upwards

**may**:  $\bigcup$  in equation for  $LV_l$

**flow-sensitive**: results depending on order of assignments

- Later: solution **not necessarily unique**

$\implies$  choose **least one**

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- **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  $l \in Lab_C$ , the **analysis information** ( $AI$ ) is described by **equations** of the form

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

where

- $\iota$  specifies the initial analysis information
- $E$  is  $\{\text{init}(C)\}$  or  $\text{final}(C)$
- $\bigoplus$  is  $\bigcap$  or  $\bigcup$
- $\varphi_{l'}$  denotes the transfer function of block  $\beta^{l'}$
- $F$  is  $\text{flow}(C)$  or  $\text{flow}^R(C)$  ( $:= \{(l', l) \mid (l, l') \in \text{flow}(C)\}$ )

- **Direction of information flow:**

- **forward:**

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

- **backward:**

- $E = \text{final}(C)$
    - $c$  has isolated exits
    - $F = \text{flow}^R(C)$
    - $AI_l$  concerns exit of  $\beta^l$

- **Quantification over paths:**

- **may:**

- $\oplus = \bigcup$
    - property satisfied by some path
    - interested in least solution (later)

- **must:**

- $\oplus = \bigcap$
    - property satisfied by all paths
    - interested in greatest solution (later)

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**Idea:** use fixpoint iteration to solve dataflow equation system

- ① For  $C \in Cmd$  and  $l \in Lab_C$ , start with “initial” information  $AI_l$   
( $AE_l = AExp_C$ ,  $LV_l = \emptyset$ )
- ② Iteratively evaluate dataflow equations until fixpoint reached

**Theoretical foundations:**

- Analysis information  $D$  forms **complete lattice**  
( $D_{AE} = 2^{AExp_C}$ ,  $D_{LV} = 2^{Var_C}$ )
  - every subset of  $D$  has a least upper/greatest lower bound  
 $\implies$  well-definedness of  $\bigoplus$
- ... that satisfies the **ascending chain condition**
  - $d_1 \supseteq d_2 \supseteq \dots \implies \exists n : d_n = d_{n+1} = \dots$
- Combination operator and all transfer functions **monotonic**
  - $d_1 \supseteq d_2 \implies \varphi(d_1) \supseteq \varphi(d_2)$

$\implies$  Fixpoint effectively computable by iteration

# Fixpoint Iteration II

Example 25.1 (Available Expressions; cf. Example 24.9)

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 25.2 (Live Variables; cf. Example 24.12)

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:

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

Fixpoint iteration:

$i$	1	2	3	4	5	6	7
0	$\emptyset$						
1	$\emptyset$	$\emptyset$	$\{y\}$	$\{x, y\}$	$\{z\}$	$\{z\}$	$\{x, y, z\}$
2	$\emptyset$	$\{y\}$	$\{x, y\}$	$\{x, y\}$	$\{y, z\}$	$\{y, z\}$	$\{x, y, z\}$
3	$\emptyset$	$\{y\}$	$\{x, y\}$	$\{x, y\}$	$\{y, z\}$	$\{y, z\}$	$\{x, y, z\}$

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## Example 25.3 (Available Expressions)

Consider

$$\begin{array}{ll} [\mathbf{z} := \mathbf{x+y}]^1; & \Rightarrow AE_1 = \emptyset \\ \mathbf{while} [\mathbf{true}]^2 \mathbf{do} & AE_2 = (AE_1 \cup \{\mathbf{x+y}\}) \cap AE_3 \\ [\mathbf{z} := \mathbf{z}]^3; & AE_3 = AE_2 \\ & \Rightarrow AE_1 = \emptyset \\ & AE_2 = \{\mathbf{x+y}\} \cap AE_3 \\ & AE_3 = AE_2 \end{array}$$

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

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

$\Rightarrow$  start fixpoint iteration with **greatest element**  $AExp_C$

## Example 25.4 (Live Variables)

Consider

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

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

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

$\implies$  start fixpoint iteration with **least element**  $\emptyset$