

Lecture 5

Foundations of Data Flow Analysis

- I. Meet operator
- II. Transfer functions
- III. Correctness, Precision, Convergence
- IV. Efficiency

•Reference: ALSU pp. 613-631
 •Background: Hecht and Ullman, Kildall, Allen and Cocke[76]
 •Marlowe & Ryder, Properties of data flow frameworks: a unified model. Rutgers tech report, Apr. 1988

A Unified Framework

- **Data flow problems are defined by**
 - Domain of values: V
 - Meet operator ($V \wedge V \rightarrow V$), initial value
 - A set of transfer functions ($V \rightarrow V$)
- **Usefulness of unified framework**
 - To answer questions such as **correctness, precision, convergence, speed of convergence** for a family of problems
 - If meet operators and transfer functions have properties X, then we know Y about the above.
 - Reuse code

I. Meet Operator

- **Properties of the meet operator**
 - **commutative:** $x \wedge y = y \wedge x$
-
- **idempotent:** $x \wedge x = x$
 - **associative:** $x \wedge (y \wedge z) = (x \wedge y) \wedge z$
 - there is a **Top** element T such that $x \wedge T = x$
- **Meet operator defines a partial ordering on values**
 - $x \leq y$ if and only if $x \wedge y = x$
 - **Transitivity:** if $x \leq y$ and $y \leq z$ then $x \leq z$
 - **Antisymmetry:** if $x \leq y$ and $y \leq x$ then $x = y$
 - **Reflexivity:** $x \leq x$

Partial Order

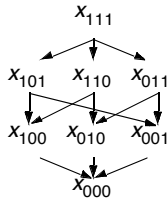
- Example: let $V = \{x \mid \text{such that } x \subseteq \{d_1, d_2, d_3\}\}$, $\wedge = \cap$
- **Top and Bottom elements**
 - **Top T** such that: $x \wedge T = x$
 - **Bottom \perp** such that: $x \wedge \perp = \perp$
- **Values and meet operator** in a data flow problem **define a semi-lattice:**
 - there exists a T , but not necessarily a \perp .
- x, y are **ordered:** $x \leq y$ then $x \wedge y = x$
- what if x and y are not ordered?
 - $x \wedge y \leq x$, $x \wedge y \leq y$, and if $w \leq x$, $w \leq y$, then $w \leq x \wedge y$

One vs. All Variables/Definitions

- Lattice for each variable: e.g. intersection



- Lattice for three variables:



Descending Chain

- Definition

- The **height** of a lattice is the largest number of **> relations** that will fit in a descending chain.

$$x_0 > x_1 > x_2 > \dots$$

- Height of values in reaching definitions?

- Important property: **finite descending chain**

- Can an infinite lattice have a finite descending chain?

- Example: **Constant Propagation/Folding**

- To determine if a variable is a constant

- Data values

- undef, ... -1, 0, 1, 2, ..., not-a-constant

II. Transfer Functions

- Basic Properties $f: V \rightarrow V$

- Has an identity function

- There exists an f such that $f(x) = x$, for all x .

- Closed under composition

- if $f_1, f_2 \in F$, then $f_1 \circ f_2 \in F$

Monotonicity

- A framework (F, V, \wedge) is **monotone** if and only if

- $x \leq y$ implies $f(x) \leq f(y)$

- i.e. a "smaller or equal" input to the same function will always give a "smaller or equal" output

- Equivalently, a framework (F, V, \wedge) is **monotone** if and only if

- $f(x \wedge y) \leq f(x) \wedge f(y)$

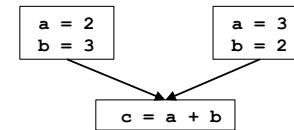
- i.e. merge input, then apply f is **small than or equal to** apply the transfer function individually and then merge the result

Example

- **Reaching definitions:** $f(x) = \text{Gen } U(x - \text{Kill})$, $\wedge = U$
 - Definition 1:
 - $x_1 \leq x_2$, $\text{Gen } U(x_1 - \text{Kill}) \leq \text{Gen } U(x_2 - \text{Kill})$
 - Definition 2:
 - $(\text{Gen } U(x_1 - \text{Kill})) \cup (\text{Gen } U(x_2 - \text{Kill}))$
 $= (\text{Gen } U((x_1 \cup x_2) - \text{Kill}))$
- **Note: Monotone framework does not mean that $f(x) \leq x$**
 - e.g., reaching definition for two definitions in program
 - suppose: $f_x: \text{Gen}_x = \{d_1, d_2\}$; $\text{Kill}_x = \{\}$
- **If input(second iteration) \leq input(first iteration)**
 - result(second iteration) \leq result(first iteration)

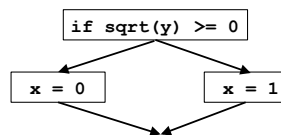
Distributivity

- A framework (F, V, \wedge) is **distributive** if and only if
 - $f(x \wedge y) = f(x) \wedge f(y)$
 - i.e. merge input, then apply f is **equal to** apply the transfer function individually then merge result
- Example: Constant Propagation



III. Data Flow Analysis

- **Definition**
 - Let $f_1, \dots, f_m \in F$, where f_i is the transfer function for node i
 - $f_p = f_{n_k} \dots f_{n_1}$, where p is a path through nodes n_1, \dots, n_k
 - $f_p = \text{identify function}$, if p is an empty path
- **Ideal data flow answer:**
 - For each node n :
 $\wedge f_{p_i}(T)$, for all **possibly executed** paths p_i reaching n .



- **Determining all possibly executed paths is undecidable**

Meet-Over-Paths (MOP)

- **Err in the conservative direction**
- **Meet-Over-Paths (MOP):**
 - For each node n :
 $\text{MOP}(n) = \wedge f_{p_i}(T)$, for all paths p_i reaching n
 - a path exists as long there is an edge in the code
 - consider more paths than necessary
 - $\text{MOP} = \text{Perfect-Solution} \wedge \text{Solution-to-Unexecuted-Paths}$
 - $\text{MOP} \leq \text{Perfect-Solution}$
 - Potentially more constrained, solution is small
 - hence *conservative*
 - It is not **safe** to be $>$ Perfect-Solution!
- **Desirable solution: as close to MOP as possible**

Solving Data Flow Equations

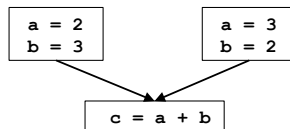
- **Example: Reaching definitions**
 - $out[entry] = \{\}$
 - **Values** = {subsets of definitions}
 - **Meet operator:** \cup
 - $in[b] = \cup out[p]$, for all predecessors p of b
 - **Transfer functions:** $out[b] = gen_b \cup (in[b] - kill_b)$
- **Any solution satisfying equations = Fixed Point Solution (FP)**
- **Iterative algorithm**
 - initializes $out[b]$ to $\{\}$
 - if converges, then it computes **Maximum Fixed Point (MFP)**:
 - **MFP** is the **largest of all solutions to equations**
- **Properties:**
 - $FP \leq MFP \leq MOP \leq \text{Perfect-solution}$
 - FP, MFP are safe
 - $in(b) \leq MOP(b)$

Partial Correctness of Algorithm

- **If data flow framework is monotone, then if the algorithm converges, $IN[b] \leq MOP[b]$**
- **Proof: Induction on path lengths**
 - Define $IN[entry] = OUT[entry]$ and transfer function of entry = Identity function
 - Base case: path of length 0
 - Proper initialization of $IN[entry]$
 - If true for path of length k , $p_k = (n_1, \dots, n_k)$, then true for path of length $k+1$: $p_{k+1} = (n_1, \dots, n_{k+1})$
 - Assume: $IN[n_k] \leq f_{n_{k-1}}(f_{n_{k-2}}(\dots f_{n_1}(IN[entry])))$
 - $IN[n_{k+1}] = OUT[n_k] \wedge \dots$
 - $\leq OUT[n_k]$
 - $\leq f_{n_k}(IN[n_k])$
 - $\leq f_{n_{k-1}}(f_{n_{k-2}}(\dots f_{n_1}(IN[entry])))$

Precision

- If data flow framework is **distributive**, then if the algorithm converges, $IN[b] = MOP[b]$
- Monotone but not distributive: behaves as if there are additional paths

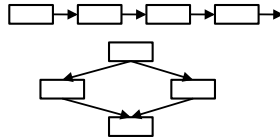


Additional Property to Guarantee Convergence

- **Data flow framework (monotone) converges if there is a finite descending chain**
- For each variable $IN[b]$, $OUT[b]$, consider the sequence of values set to each variable **across iterations**:
 - if sequence for $in[b]$ is **monotonically decreasing**
 - sequence for $out[b]$ is **monotonically decreasing**
 - ($out[b]$ initialized to T)
 - if sequence for $out[b]$ is **monotonically decreasing**
 - sequence of $in[b]$ is **monotonically decreasing**

IV. Speed of Convergence

- Speed of convergence depends on order of node visits



- Reverse "direction" for backward flow problems

Reverse Postorder

- Step 1: depth-first post order

```
main() {
    count = 1;
    Visit(root);
}
Visit(n) {
    for each successor s that has not been visited
        Visit(s);
    PostOrder(n) = count;
    count = count+1;
}
```

- Step 2: reverse order

```
For each node i
    rPostOrder = NumNodes - PostOrder(i)
```

Depth-First Iterative Algorithm (forward)

input: control flow graph CFG = (N, E, Entry, Exit)

```
/* Initialize */
out[entry] = init_value
For all nodes i
    out[i] = T
Change = True
/* iterate */
While Change {
    Change = False
    For each node i in rPostOrder {
        in[i] =  $\wedge$ (out[p]), for all predecessors p of i
        oldout = out[i]
        out[i] =  $f_i$ (in[i])
        if oldout  $\neq$  out[i]
            Change = True
    }
}
```

Speed of Convergence

- If cycles do not add information
 - information can flow in one pass down a series of nodes of increasing order number:
 - e.g., $1 \rightarrow 4 \rightarrow 5 \rightarrow 7 \rightarrow 2 \rightarrow 4 \dots$
 - passes determined by number of back edges in the path
 - essentially the nesting depth of the graph
 - Number of iterations = number of back edges in any acyclic path + 2
 - (2 are necessary even if there are no cycles)
- What is the depth?
 - corresponds to depth of intervals for "reducible" graphs
 - in real programs: average of 2.75

A Check List for Data Flow Problems

- **Semi-lattice**
 - set of values
 - meet operator
 - top, bottom
 - finite descending chain?
- **Transfer functions**
 - function of each basic block
 - monotone
 - distributive?
- **Algorithm**
 - initialization step (entry/exit, other nodes)
 - visit order: rPostOrder
 - depth of the graph