CSC D70: Compiler Optimization Dataflow Analysis

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The content of this lecture is adapted from the lectures of Todd Mowry and Phillip Gibbons

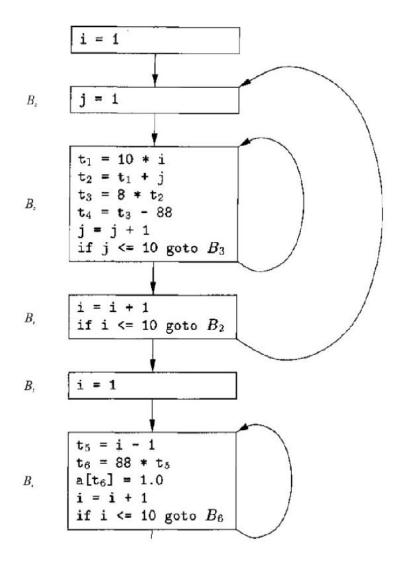
Refreshing from Last Lecture

• Basic Block Formation

• Value Numbering

Partitioning into Basic Blocks

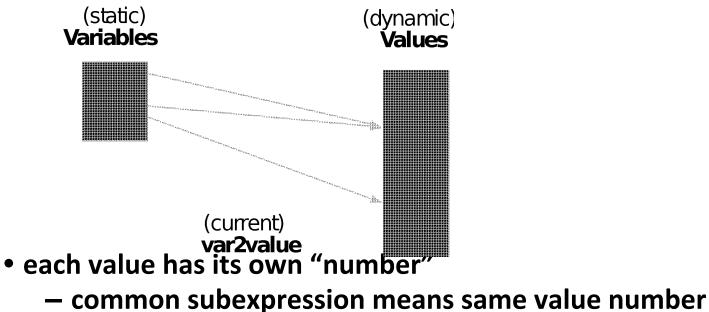
- Identify the leader of each basic block
 - First instruction
 - Any target of a jump
 - Any instruction immediately following a jump
- Basic block starts at leader & ends at instruction immediately before a leader (or the last instruction)



ALSU pp. 529-531

Value Numbering (VN)

• More explicit with respect to VALUES, and TIME



- var2value: current map of variable to value
 - used to determine the value number of current of
 - used to determine the value number of current expression

r1 + r2 => var2value(r1)+var2value(r2)

Algorithm

```
Data structure:
    VALUES = Table of
                       //[OP, valnum1, valnum2}
        expression
                       //name of variable currently holding expression
        var
For each instruction (dst = src1 OP src2) in execution order
 valnum1 = var2value(src1); valnum2 = var2value(src2);
  IF [OP, valnum1, valnum2] is in VALUES
    v = the index of expression
    Replace instruction with CPY dst = VALUES[v].var
  ELSE
     Add
        expression = [OP, valnum1, valnum2]
                   = dst
        var
     to VALUES
     v = index of new entry; tv is new temporary for v
     Replace instruction with: tv = VALUES[valnum1].var OP VALUES[valnum2].var
                               CPY dst = tv;
```

```
set_var2value (dst, v)
```

VN Example

Assign: $a \rightarrow r1, b \rightarrow r2, c \rightarrow r3, d \rightarrow r4$

- a = b+c; ADD t1 = r2,r3
- CPY r1 = t1 //(a = t1)
- b = a-d; SUB t2 = r1, r4
- CPY $r^2 = t^2 //(b = t^2)$ c = b+c; ADD t3 = r^2, r^3
 - CPY $r_3 = t_3$ //(c = t_3)
- d = a-d; CPY $r^2 = t^2$

Questions about Assignment #1

• Tutorial #1

- Tutorial #2 next week
 - More in-depth LLVM coverage

Outline

- 1. Structure of data flow analysis
- 2. Example 1: Reaching definition analysis
- 3. Example 2: Liveness analysis
- 4. Generalization

What is Data Flow Analysis?

- Local analysis (e.g., value numbering)
 - analyze effect of each instruction
 - compose effects of instructions to derive information from beginning of basic block to each instruction

• Data flow analysis

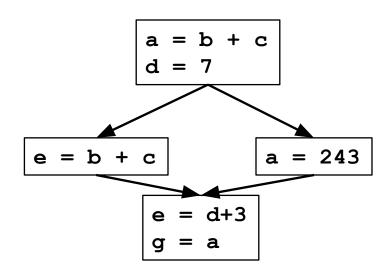
- analyze effect of each basic block
- compose effects of basic blocks to derive information at basic block boundaries
- from basic block boundaries, apply local technique to generate information on instructions

What is Data Flow Analysis? (2)

• Data flow analysis:

- Flow-sensitive: sensitive to the control flow in a function
- intraprocedural analysis
- Examples of optimizations:
 - Constant propagation
 - Common subexpression elimination
 - Dead code elimination

What is Data Flow Analysis? (3)



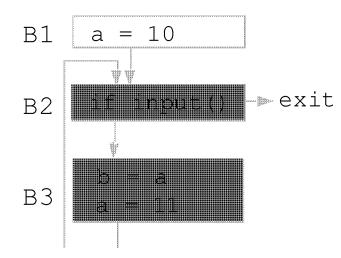
For each variable x determine:

Value of x?

Which "definition" defines x?

Is the definition still meaningful (live)?

Static Program vs. Dynamic Execution



- Statically: Finite program
- **Dynamically**: Can have infinitely many possible execution paths
- Data flow analysis abstraction:
 - For each point in the program: combines information of all the instances of the same program point.
- Example of a data flow question:
 - Which definition defines the value used in statement "b = a"?

Effects of a Basic Block

- Effect of a statement: **a** = **b**+**c**
 - Uses variables (b, c)
 - Kills an old definition (old definition of a)
 - new **definition** (a)
- Compose effects of statements -> Effect of a basic block
 - A locally exposed use in a b.b. is a use of a data item which is not preceded in the b.b. by a definition of the data item
 - any definition of a data item in the basic block kills all definitions of the same data item reaching the basic block.
 - A locally available definition = last definition of data item in b.b.

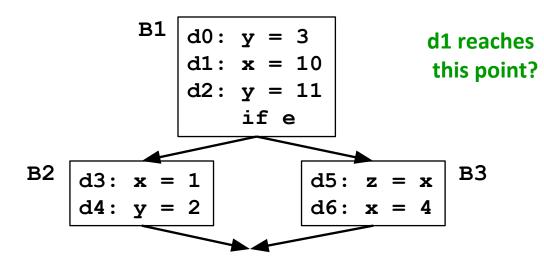
Effects of a Basic Block

A **locally available definition** = last definition of data item in b.b.

t1 = r1+r2	Locally exposed uses? r1	
r2 = t1		
t2 = r2 + r1	Kills any definitions?	Any other
r1 = t2		definition
t3 = r1*r1		of t2
r2 = t3		
if r2>100 goto	L1	

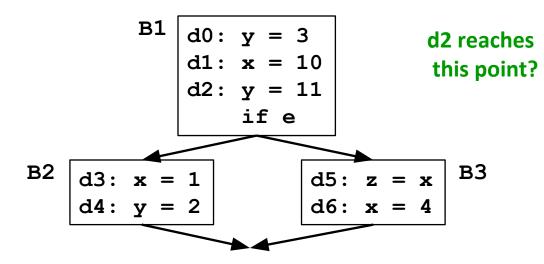
Locally avail. definition? t2

Reaching Definitions



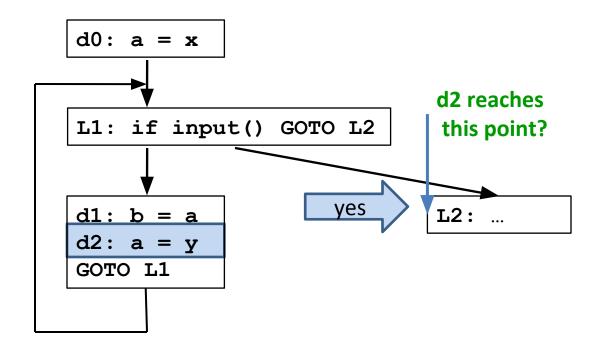
- Every assignment is a **definition**
- A definition d reaches a point p if there exists path from the point immediately following d to p such that d is not killed (overwritten) along that path.
- Problem statement
 - For each point in the program, determine if each definition in the program reaches the point
 - A bit vector per program point, vector-length = #defs

Reaching Definitions (2)

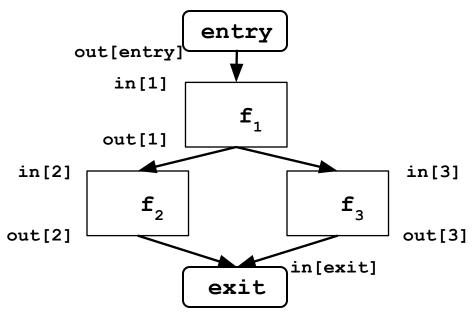


- Every assignment is a **definition**
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Reaching Definitions (3)

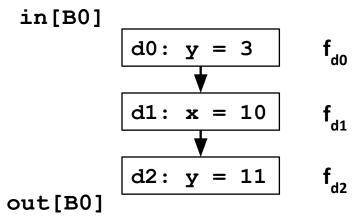


Data Flow Analysis Schema



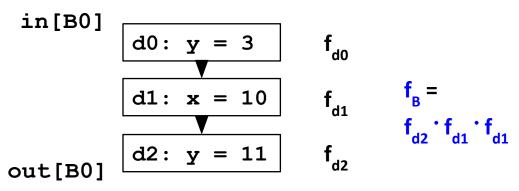
- Build a flow graph (nodes = basic blocks, edges = control flow)
- Set up a set of equations between in[b] and out[b] for all basic blocks b
 - Effect of code in basic block:
 - Transfer function f_b relates in[b] and out[b], for same b
 - Effect of flow of control:
 - relates out[b₁], in[b₂] if b₁ and b₂ are adjacent
- Find a solution to the equations

Effects of a Statement



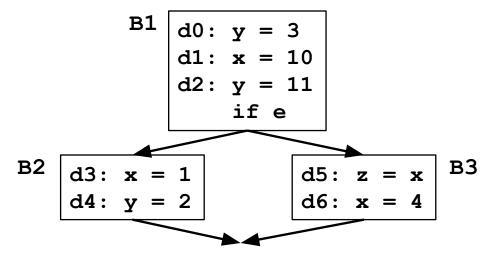
- f_c: A transfer function of a statement
 - abstracts the execution with respect to the problem of interest
- For a statement s (d: x = y + z) out[s] = f_s(in[s]) = Gen[s] U (in[s]-Kill[s])
 - Gen[s]: definitions generated: Gen[s] = {d}
 - Propagated definitions: in[s] Kill[s], where Kill[s]=set of all other defs to x in the rest of program

Effects of a Basic Block



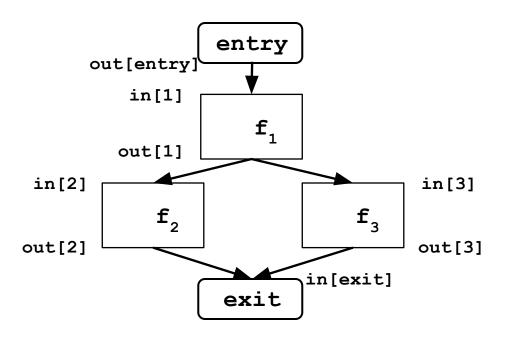
- Transfer function of a statement s:
 - out[s] = f_s(in[s]) = Gen[s] U (in[s]-Kill[s])
- Transfer function of a basic block B:
 - Composition of transfer functions of statements in B
- $out[B] = f_B(in[B]) = f_{d2}f_{d1}f_{d0}(in[B])$ = $Gen[d_2] \cup (Gen[d_1] \cup (Gen[d_0] \cup (in[B]-Kill[d_0]))-Kill[d_1])) -Kill[d_2]$ = $Gen[d_2] \cup (Gen[d_1] \cup (Gen[d_0] - Kill[d_1]) - Kill[d_2]) \cup in[B] - (Kill[d_0] \cup Kill[d_1] \cup Kill[d_2])$
 - = Gen[B] U (in[B] Kill[B])
 - Gen[B]: locally exposed definitions (available at end of bb)
 - Kill[B]: set of definitions killed by B

Example



- a transfer function f, of a basic block b: OUT[b] = f, (IN[b]) incoming reaching definitions -> outgoing reaching definitions
- A basic block b
 - generates definitions: Gen[b],
 - set of locally available definitions in b
 - kills definitions: in[b] Kill[b], where Kill[b]=set of defs (in rest of program) killed by defs in b
- out[b] = Gen[b] U (in(b)-Kill[b])

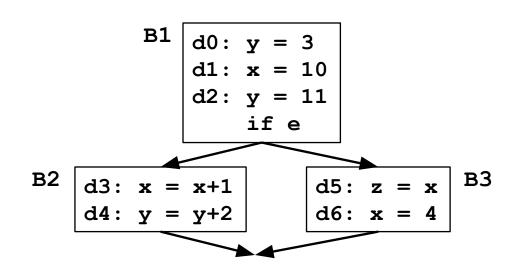
Effects of the Edges (acyclic)



- $out[b] = f_b(in[b])$
- Join node: a node with multiple predecessors
- meet operator:

in[b] = out[p_1] U out[p_2] U ... U out[p_n], where p_1 , ..., p_n are all predecessors of b

Example

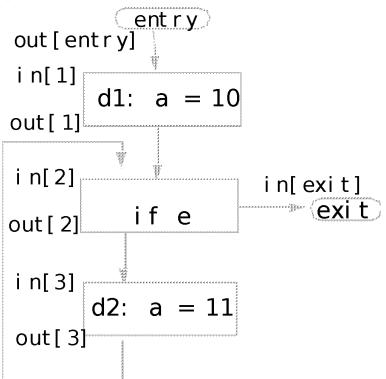


f Gen Kill
1 {1,2} {0,2,3,4,6}
2 {3,4} {0,1,2,6}
3 {5,6} {1,3}

- $out[b] = f_b(in[b])$
- Join node: a node with multiple predecessors
- meet operator:

in[b] = out[p_1] U out[p_2] U ... U out[p_n], where p_1 , ..., p_n are all predecessors of b

Cyclic Graphs



- Equations still hold
 - out[b] = f_b(in[b])
 - in[b] = out[p₁] U out[p₂] U ... U out[p_n], p₁, ..., p_n pred.
- Find: fixed point solution

Reaching Definitions: Iterative Algorithm

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

```
// Boundary condition
   out[Entry] = Ø
```

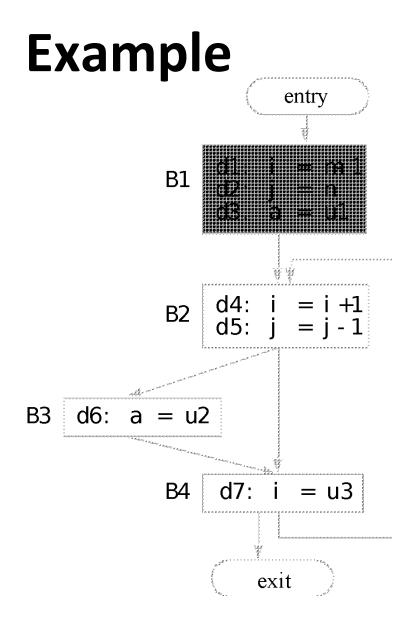
// Initialization for iterative algorithm
For each basic block B other than Entry
out[B] = Ø

```
// iterate
While (Changes to any out[] occur) {
    For each basic block B other than Entry {
        in[B] = U (out[p]), for all predecessors p of B
        out[B] = f<sub>B</sub>(in[B]) // out[B]=gen[B]U(in[B]-kill[B])
    }
```

Reaching Definitions: Worklist Algorithm

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

```
// iterate
While ChangedNodes ≠ Ø {
    Remove i from ChangedNodes
    in[i] = U (out[p]), for all predecessors p of i
    oldout = out[i]
    out[i] = f<sub>i</sub>(in[i]) // out[i]=gen[i]U(in[i]-kill[i])
    if (oldout ≠ out[i]) {
        for all successors s of i
            add s to ChangedNodes
    }
}
```



	First Pass	Second Pass
IN[B1]	000 00 0 0	000 00 0 0
OUT[B1]	111 00 0 0	111 00 0 0
IN[B2]	111 00 0 0	111 01 1 1
OUT[B2]	001 11 0 0	001 11 1 0
IN[B3]	001 11 0 0	001 11 1 0
OUT[B3]	000 11 1 0	000 11 1 0
IN[B4]	001 11 1 0	001 11 1 0
OUT[B4]	<mark>001</mark> 01 1 1	001 01 1 1
IN[exit]	001 01 1 1	001 01 1 1

Live Variable Analysis

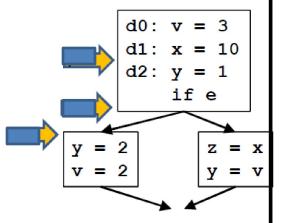
• Definition

- A variable \mathbf{v} is **live** at point p if
 - the value of \mathbf{v} is used along some path in the flow graph starting at p.
- Otherwise, the variable is **dead**.
- Motivation
 - e.g. register allocation

```
for i = 0 to n
... i ...
```

for i = 0 to n ... i ...

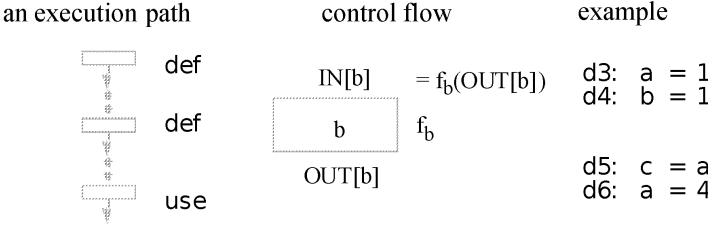
v live at this point?



- Problem statement
 - For each basic block
 - determine if each variable is live in each basic block
 - Size of bit vector: one bit for each variable

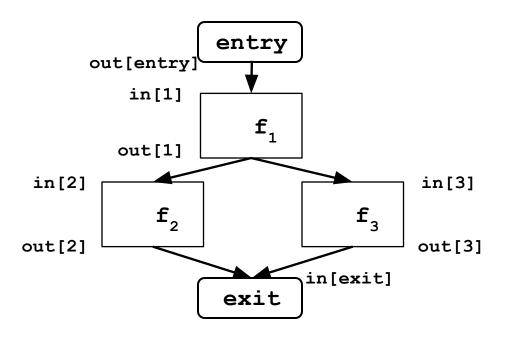
Transfer Function

Insight: Trace uses backwards to the definitions



- A basic block b can
 - generate live variables: Use[b]
 - set of locally exposed uses in b
 - propagate incoming live variables: OUT[b] Def[b],
 - where Def[b] = set of variables defined in b.b.
- transfer function for block b: in[b] = Use[b] U (out(b)-Def[b])

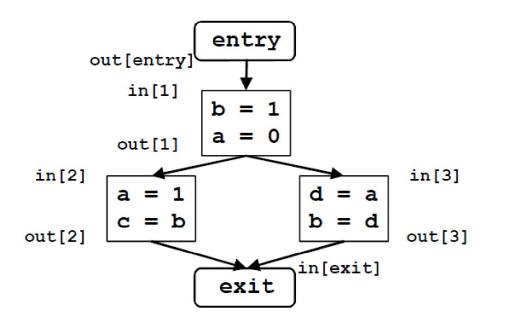
Flow Graph



- in[b] = f_b(out[b])
- Join node: a node with multiple successors
- meet operator:

out[b] = in[s₁] U in[s₂] U ... U in[s_n], where s₁, ..., s_n are all successors of b

Flow Graph (2)



f Use Def 1 {} {a,b} 2 {b} {a,c} 3 {a} {b,d}

- in[b] = f_b(out[b])
- Join node: a node with multiple successors
- meet operator:

out[b] = in[s₁] U in[s₂] U ... U in[s_n], where s₁, ..., s_n are all successors of b

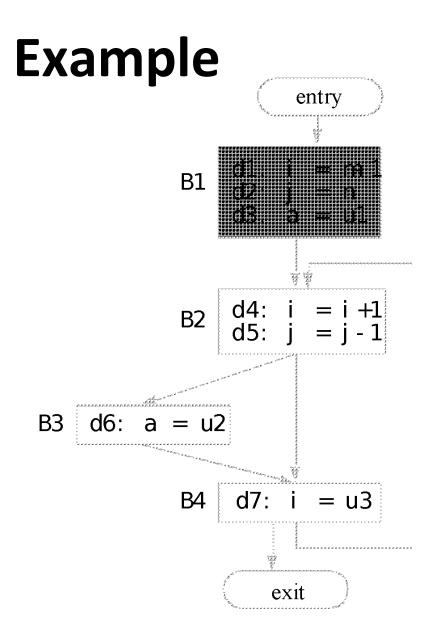
Liveness: Iterative Algorithm

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

```
// Boundary condition
in[Exit] = Ø
```

// Initialization for iterative algorithm
For each basic block B other than Exit
in[B] = Ø

```
// iterate
While (Changes to any in[] occur) {
    For each basic block B other than Exit {
        out[B] = U (in[s]), for all successors s of B
        in[B] = f<sub>B</sub>(out[B]) // in[B]=Use[B]U(out[B]-Def[B])
    }
```



	First Pass	Second Pass
OUT[entry]	{m,n,u1,u2,u3}	{m,n,u1,u2,u3}
IN[B1]	{m,n,u1,u2,u3}	{m,n,u1,u2,u3}
OUT[B1]	{i,j,u2,u3}	{i,j,u2,u3}
IN[B2]	{i,j,u2,u3}	{i,j,u2,u3}
		(; , , , ,)
OUT[B2]	{u2,u3}	{j,u2,u3}
IN[B3]	{u2,u3}	{j,u2,u3}
OUT[B3]	{u3}	{j,u2,u3}
IN[B4]	{u3}	{j,u2,u3}
OUT[B4]	{}	{i,j,u2,u3}

Framework

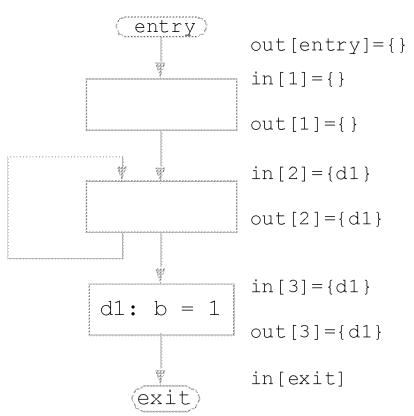
	Reaching Definitions	Live Variables
Domain	Sets of definitions	Sets of variables
Direction	forward: $out[b] = f_b(in[b])$ $in[b] = \land out[pred(b)]$	backward: in[b] = $f_b(out[b])$ out[b] = \land in[succ(b)]
Transfer function	$f_b(x) = Gen_b \cup (x - Kill_b)$	$f_{b}(x) = Use_{b} \cup (x - Def_{b})$
Meet Operation (\wedge)	U	U
Boundary Condition	out[entry] = \emptyset	in[exit] = \emptyset
Initial interior points	out[b] = Ø	in[b] = Ø

Other examples (e.g., Available expressions), defined in ALSU 9.2.6

Thought Problem 1. "Must-Reach" Definitions

- A definition D (a = b+c) <u>must</u> reach point P iff
 - D appears at least once along on all paths leading to P
 - a is not redefined along any path after last appearance of D and before P
- How do we formulate the data flow algorithm for this problem?

Thought Problem 2: A legal solution to (May) Reaching Def?



• Will the worklist algorithm generate this answer?

Questions

Correctness

- equations are satisfied, if the program terminates.
- Precision: how good is the answer?
 - is the answer ONLY a union of all possible executions?
- Convergence: will the analysis terminate?
 - or, will there always be some nodes that change?
- Speed: how fast is the convergence?
 - how many times will we visit each node?

Foundations of Data Flow Analysis

- 1. Meet operator
- 2. Transfer functions
- 3. Correctness, Precision, Convergence
- 4. 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 \land V \rightarrow V), initial value
- A set of transfer functions ($V \rightarrow V$)

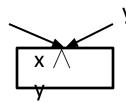
Usefulness of unified framework

- 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

Meet Operator

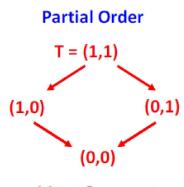
• Properties of the meet operator

• commutative: $x \land y = y \land x$



- idempotent: $x \land x = x$
- associative: $x \land (y \land z) = (x \land y) \land z$
- there is a Top element T such that $x \land T = x$

Х

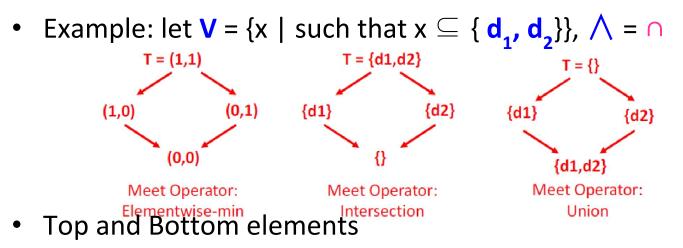


Meet Operator: Elementwise-min

• Meet operator defines a partial ordering on values

- $x \le y$ if and only if $x \land y = x$ (y -> x in diagram)
 - Transitivity: if $x \le y$ and $y \le z$ then $x \le z$
 - Antisymmetry: if $x \le y$ and $y \le x$ then x = y
 - Reflexitivity: $x \le x$

Partial Order



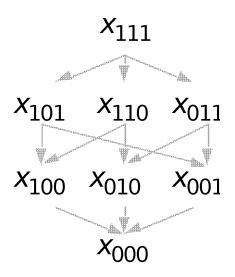
- Top T such that: $x \wedge T = x$
- Bottom \perp such that: $\mathbf{x} \land \perp = \perp$
- Values and meet operator in a data flow problem define a semi-lattice:
 - there exists a T, but not necessarily a \perp .
- x, y are ordered: $x \le y$ then $x \land y = x$ (y -> x in diagram)
- what if x and y are not ordered?

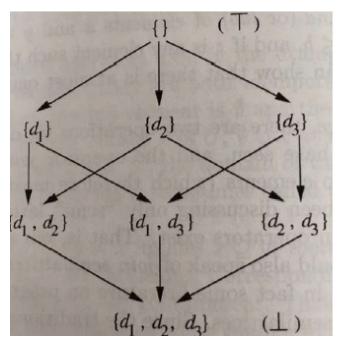
• $x \land y \le x, x \land y \le y$, and if $w \le x, w \le y$, then $w \le x \land y$

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?

Height n – number of definitions

- Important property: finite descending chain
- Can an infinite lattice have a finite descending chain? yes
- Example: Constant Propagation/Folding
 - To determine if a variable is a constant
- Data values
 - undef, ... -1, 0, 1, 2, ..., not-a-constant

Transfer Functions

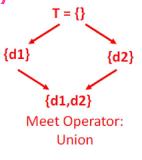
- Basic Properties $f: V \rightarrow V$
 - Has an identity function
 - There exists an f such that f (x) = x, for all x.
 - Closed under composition
 - if $f_1, f_2 \in F$, then $f_1 \cdot f_2 \in F$

Monotonicity

- A framework (*F, V,* \land) is monotone if and only if
 - $x \le y$ implies $f(x) \le f(y)$
 - 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
 - $f(x \land y) \leq f(x) \land f(y)$
 - 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: $f(x) = Gen \cup (x Kill), \land = \cup$
 - Definition 1:
 - $x_1 \le x_2$, Gen U $(x_1 Kill) \le Gen U (x_2 Kill)$
 - Definition 2:
 - (Gen U (x₁ Kill)) U (Gen U (x₂ Kill)) = (Gen U ((x₁ U x₂) - Kill))



 $[\mathbf{x}_1 \le \mathbf{x}_2 \text{ iff } \mathbf{x}_2 \to \mathbf{x}_1]$

Note: Monotone framework does not mean that f(x) ≤ x

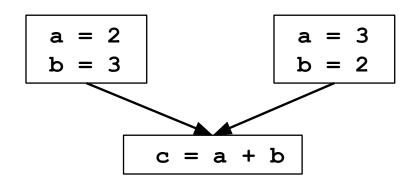
- e.g., reaching definition for two definitions in program
- suppose: f_x: Gen_x = {d₁, d₂}; Kill_x = {}

If input(second iteration) ≤ input(first iteration)

result(second iteration) ≤ result(first iteration)

Distributivity

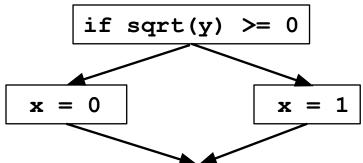
- A framework (*F, V,* \land) is **distributive** if and only if
 - $f(x \land y) = f(x) \land f(y)$
 - i.e. merge input, then apply f is **equal to** apply the transfer function individually then merge result
- Example: Constant Propagation is NOT distributive



Data Flow Analysis

- Definition
 - Let $f_1, ..., f_m := F$, where f_i is the transfer function for node *i* • $f_p = f_{nk} \cdot ... \cdot f_{n1}$, where *p* is a path through nodes $n_1, ..., n_k$
 - $f_n =$ identify function, if p is an empty path
- Ideal data flow answer:
 - For each node n:

 $\bigwedge f_{p_i}$ (T), for all possibly executed paths p_i reaching n.



• But determining all possibly executed paths is undecidable

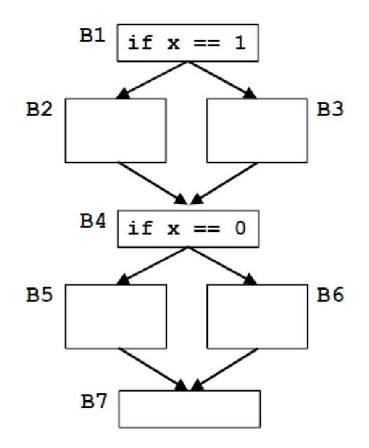
Meet-Over-Paths (MOP)

- Error in the conservative direction
- Meet-Over-Paths (MOP):
 - For each node *n*:

 $MOP(n) = \bigwedge f_{pi}(T)$, for all paths p_i reaching n

- a path exists as long there is an edge in the code
- consider more paths than necessary
- MOP = Perfect-Solution \land Solution-to-Unexecuted-Paths
- MOP ≤ 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

MOP Example



Assume: B2 & B3 do not update x

Ideal: Considers only 2 paths B1-B2-B4-B6-B7 (i.e., x=1) B1-B3-B4-B5-B7 (i.e., x=0)

MOP: Also considers unexecuted paths B1-B2-B4-B5-B7 B1-B3-B4-B6-B7

Solving Data Flow Equations

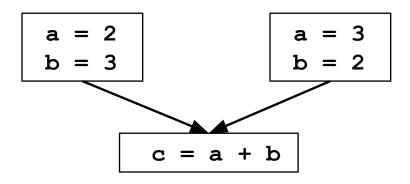
- Example: Reaching definitions
 - out[entry] = {}
 - Values = {subsets of definitions}
 - Meet operator: \cup
 - in[b] = ∪ out[p], for all predecessors p of b
 - Transfer functions: $out[b] = gen_b \cup (in[b] kill_b)$
- Any solution satisfying equations = Fixed Point Solution (FP)
- Iterative algorithm
 - initializes out[b] to {}
 - if converges, then it computes Maximum Fixed Point (MFP):
 - MFP is the largest of all solutions to equations
- Properties:
 - FP ≤ MFP ≤ MOP ≤ 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] ≤ 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, $p_k = (n_1, ..., n_k)$, then true for path of length k+1: $p_{k+1} = (n_1, ..., n_{k+1})$
 - Assume: $IN[n_k] \le f_{nk-1}(f_{nk-2}(...f_{n1}^{k+1}(IN[entry])))$
 - $IN[n_{k+1}] = OUT[n_k] \land ...$ $\leq OUT[n_k]$ $\leq f_{nk} (IN[n_k])$ $\leq f_{nk-1} (f_{nk-2} (... f_{n1} (IN[entry])))$

Precision

 If data flow framework is distributive, then if the algorithm converges, IN[b] = MOP[b]



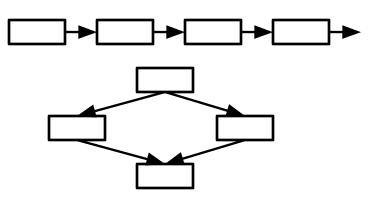
Monotone but not distributive: behaves as if there are additional paths

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

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] = \langle (out[p]), for all predecessors p of i
          oldout = out[i]
           out[i] = f<sub>i</sub>(in[i])
          if oldout # 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 -> 4 -> 5 -> 7 -> 2 -> 4 ...
- 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

Conclusions

- Dataflow analysis examples
 - Reaching definitions
 - Live variables
- Dataflow formation definition
 - Meet operator
 - Transfer functions
 - Correctness, Precision, Convergence
 - Efficiency

CSC D70: Compiler Optimization Dataflow Analysis

Prof. Gennady Pekhimenko University of Toronto Winter 2019

The content of this lecture is adapted from the lectures of Todd Mowry and Phillip Gibbons