## 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

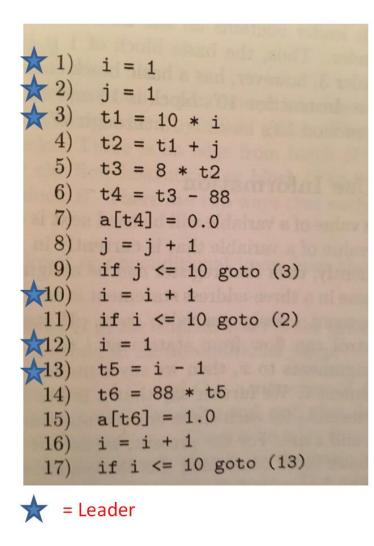
## **Refreshing from Last Lecture**

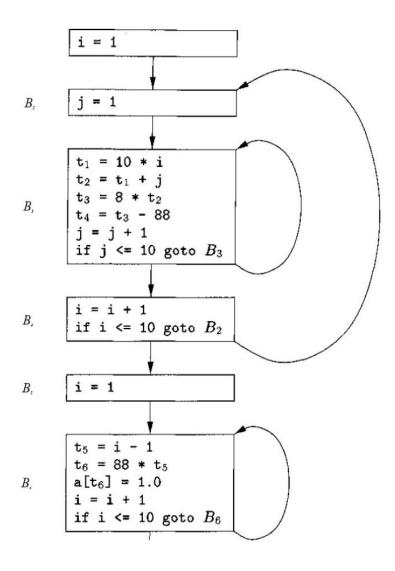
Basic Block Formation

• 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)

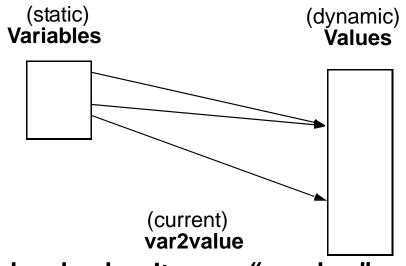




ALSU pp. 529-531

# Value Numbering (VN)

• More explicit with respect to VALUES, and TIME



- each value has its own "number"
  - common subexpression means same value number
- var2value: current map of variable to value
  - 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]
        var
                   = dst
     to VALUES
     v = index of new entry; tv is new temporary for v
     Replace instruction with: tv = VALUES[valnum1].var OP VALUES[valnum2].var
                               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,r3CPY r1 = t1
- b = a-d; SUB t2 = r1, r4CPY r2 = t2
- c = b+c; ADD t3 = r2,r3
  - CPY r3 = t3
- d = a-d; SUB t4 = r1, r4CPY r4 = t4

### **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

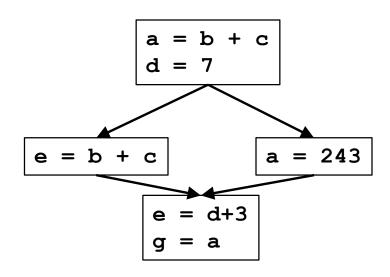
- analyze effect of each basic block
- compose effects of basic blocks to derive information at basic block boundaries
- from basic block boundaries, apply local technique to generate information on instructions

# 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)



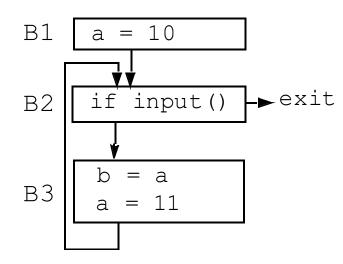
For each variable x determine:

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**

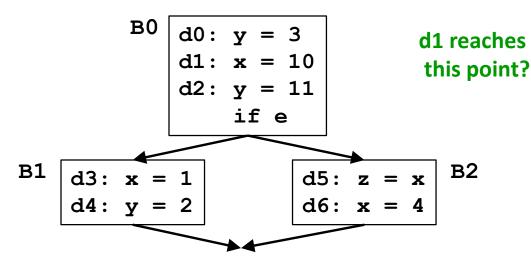
- Effect of a statement: **a** = **b**+**c** 
  - Uses variables (b, c)
  - Kills an old definition (old definition of a)
  - new **definition** (a)
- Compose effects of statements -> Effect of a basic block
  - A locally exposed use in a b.b. is a use of a data item which is not preceded in the b.b.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

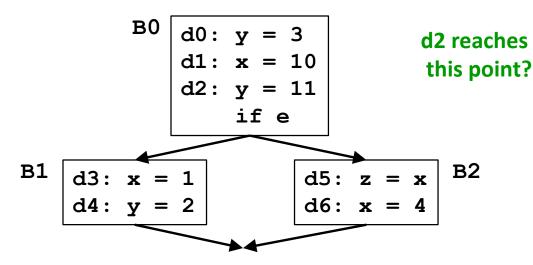
t1 = r1+r2
r2 = t1
t2 = r2+r1
r1 = t2
t3 = r1\*r1
r2 = t3
if r2>100 goto L1

## **Reaching Definitions**



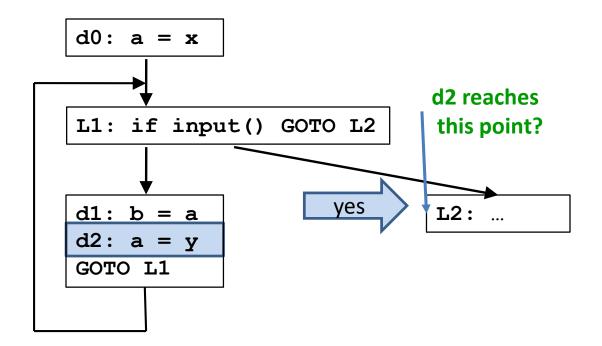
- Every assignment is a **definition**
- A definition d reaches a point p if there exists path from the point immediately following d to p such that d is not killed (overwritten) along that path.
- Problem statement
  - For each point in the program, determine if each definition in the program reaches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

# **Reaching Definitions (2)**

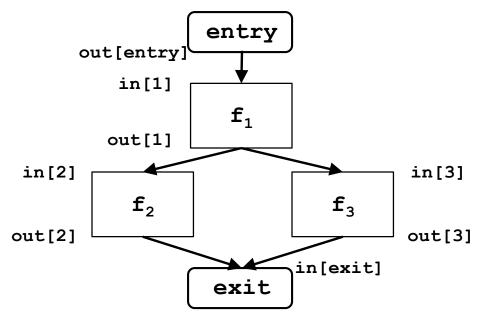


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## **Reaching Definitions (3)**

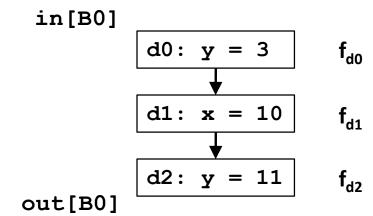


## **Data Flow Analysis Schema**



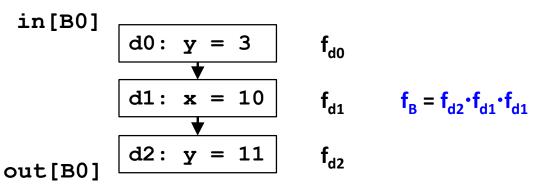
- Build a flow graph (nodes = basic blocks, edges = control flow)
- Set up a set of equations between in[b] and out[b] for all basic blocks b
  - Effect of code in basic block:
    - Transfer function f<sub>b</sub> relates in[b] and out[b], for same b
  - Effect of flow of control:
    - relates out[b<sub>1</sub>], in[b<sub>2</sub>] if b<sub>1</sub> and b<sub>2</sub> are adjacent
- Find a solution to the equations

#### **Effects of a Statement**



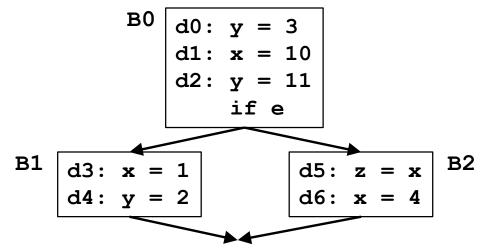
- f<sub>s</sub>: A transfer function of a statement
  - abstracts the execution with respect to the problem of interest
- For a statement s (d: x = y + z) out[s] = f<sub>s</sub>(in[s]) = Gen[s] U (in[s]-Kill[s])
  - Gen[s]: definitions generated: Gen[s] = {d}
  - Propagated definitions: in[s] Kill[s], 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<sub>s</sub>(in[s]) = Gen[s] U (in[s]-Kill[s])
- Transfer function of a basic block B:
  - Composition of transfer functions of statements in B
- $out[B] = f_B(in[B]) = f_{d2}f_{d1}f_{d0}(in[B])$ =  $Gen[d_2] \cup (Gen[d_1] \cup (Gen[d_0] \cup (in[B]-Kill[d_0]))-Kill[d_1])) -Kill[d_2]$ =  $Gen[d_2] \cup (Gen[d_1] \cup (Gen[d_0] - Kill[d_1]) - Kill[d_2]) \cup I_{in[B]} - (Kill[d_0] \cup Kill[d_1] \cup Kill[d_2])$ 
  - = 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

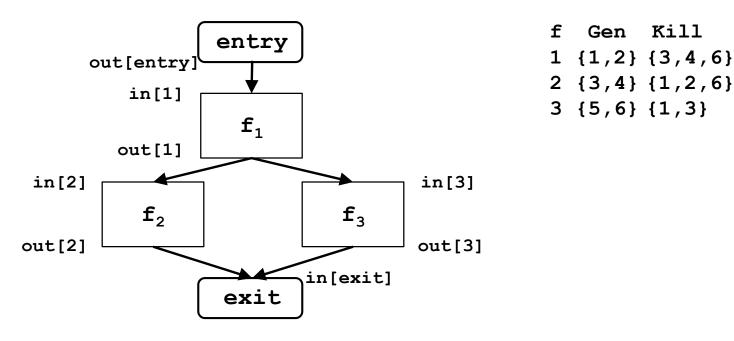


a transfer function f<sub>b</sub> of a basic block b:
 OUT[b] = f<sub>b</sub>(IN[b])

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)**



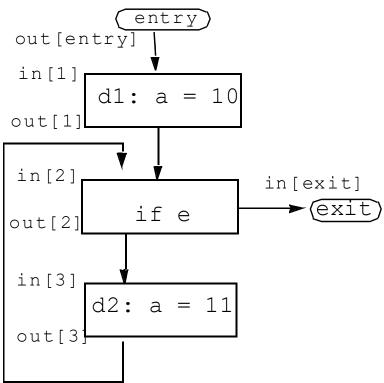
- $out[b] = f_b(in[b])$ •
- Join node: a node with multiple predecessors ullet
- meet operator: ٠

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

Kill

Gen

## **Cyclic Graphs**



- Equations still hold
  - out[b] = f<sub>b</sub>(in[b])
  - in[b] = out[p<sub>1</sub>] U out[p<sub>2</sub>] U ... U out[p<sub>n</sub>], p<sub>1</sub>, ..., p<sub>n</sub> 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] =  $\emptyset$ 

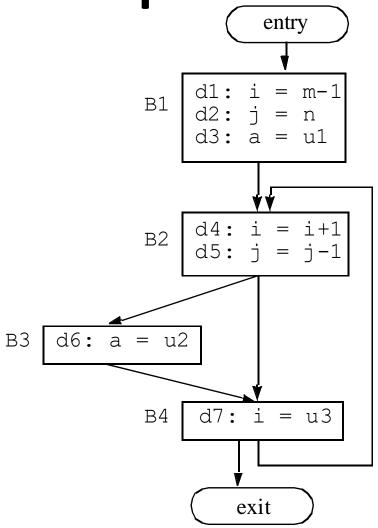
```
// iterate
While (Changes to any out[] occur) {
    For each basic block B other than Entry {
        in[B] = U (out[p]), for all predecessors p of B
        out[B] = f<sub>B</sub>(in[B]) // out[B]=gen[B]U(in[B]-kill[B])
    }
```

#### **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
    }
}
```

#### Example



	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]	001 01 1 1	001 01 1 1
IN[exit]	001 01 1 1	001 01 1 1

# Live Variable Analysis

#### • Definition

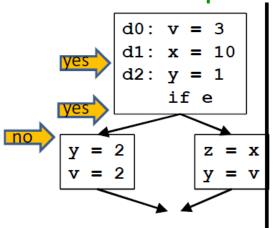
- A variable  $\mathbf{v}$  is **live** at point p if
  - the value of **v** is used along some path in the flow graph starting at *p*.
- Otherwise, the variable is dead.

#### Motivation

• e.g. register allocation

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

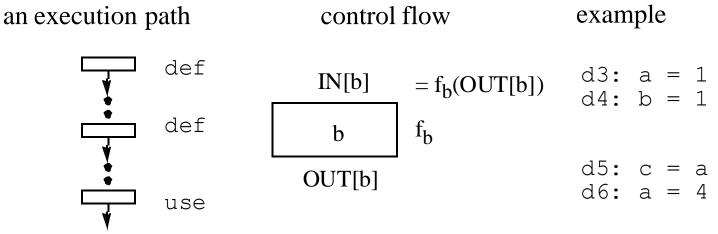
v live at this point?



- 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**

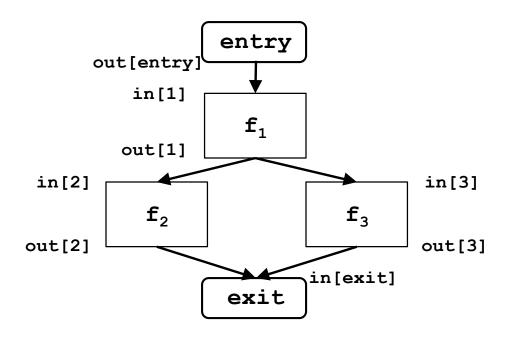
Insight: Trace uses backwards to the definitions



#### • 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])

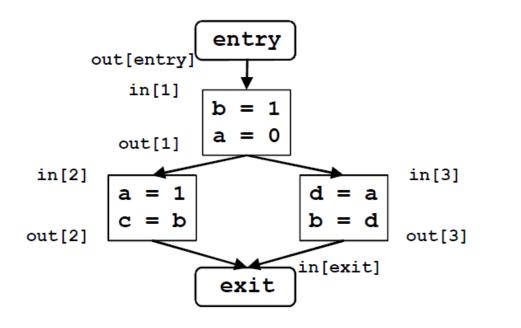
## **Flow Graph**



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

out[b] = in[s<sub>1</sub>] U in[s<sub>2</sub>] U ... U in[s<sub>n</sub>], where s<sub>1</sub>, ..., s<sub>n</sub> 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<sub>b</sub>(out[b])
- Join node: a node with multiple successors
- meet operator:

out[b] = in[s<sub>1</sub>] U in[s<sub>2</sub>] U ... U in[s<sub>n</sub>], where s<sub>1</sub>, ..., s<sub>n</sub> are all successors of b

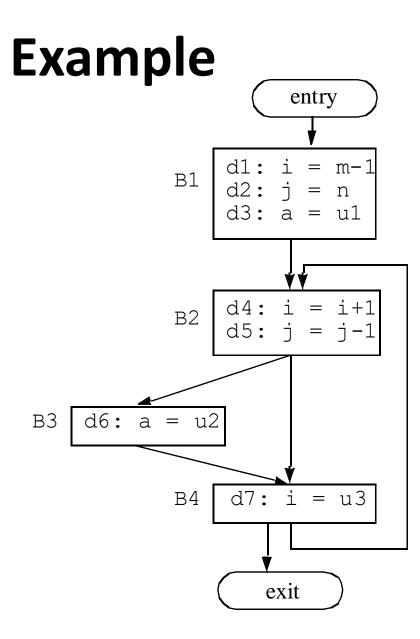
#### **Liveness: Iterative Algorithm**

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  $in[B] = \emptyset$ 

```
// 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

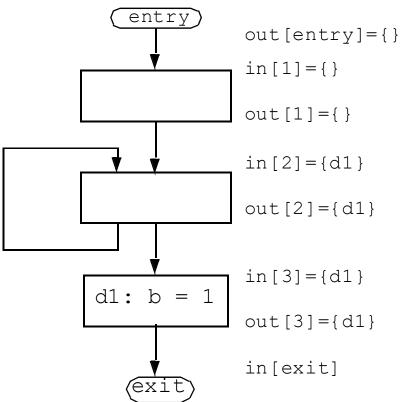
	Reaching Definitions	Live Variables
Domain	Sets of definitions	Sets of variables
Direction	forward: $out[b] = f_b(in[b])$ $in[b] = \land out[pred(b)]$	backward: in[b] = f <sub>b</sub> (out[b]) out[b] = $\land$ in[succ(b)]
Transfer function	$f_b(x) = Gen_b \cup (x - Kill_b)$	$f_b(x) = Use_b \cup (x - Def_b)$
Meet Operation (^)	$\cup$	U
Boundary Condition	out[entry] = $\emptyset$	$in[exit] = \emptyset$
Initial interior points	$out[b] = \emptyset$	$in[b] = \emptyset$

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?

#### 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

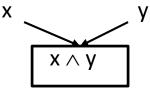
- Data flow problems are defined by
  - Domain of values: V
  - Meet operator ( $V \land V \rightarrow V$ ), initial value
  - A set of transfer functions ( $V \rightarrow V$ )

#### Usefulness of unified framework

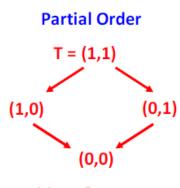
- To answer questions such as correctness, precision, convergence, speed of convergence for a family of problems
  - If meet operators and transfer functions have properties X, then we know Y about the above.
- Reuse code

#### **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  $x \wedge T = x$



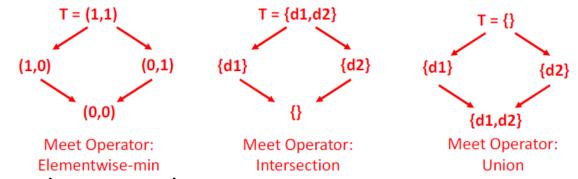
Meet Operator: Elementwise-min

#### • 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**

• Example: let  $V = \{x \mid \text{such that } x \subseteq \{ d_1, d_2 \}\}, \land = \cap$ 

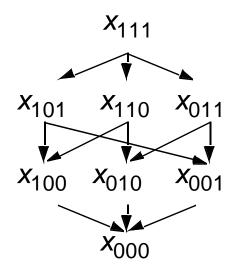


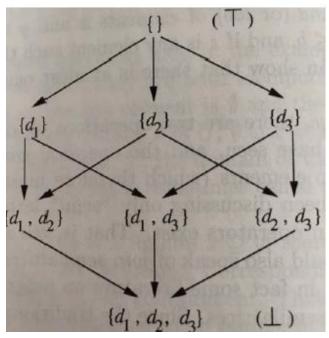
- Top and Bottom elements
  - Top T such that:  $x \wedge T = x$
  - Bottom  $\perp$  such that:  $\mathbf{x} \wedge \perp = \perp$
- Values and meet operator in a data flow problem define a semilattice:
  - there exists a T, but not necessarily a  $\perp$ .
- x, y are ordered:  $x \le y$  then  $x \land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

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#### Example

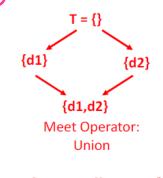
- Reaching definitions:  $f(x) = Gen \cup (x Kill), \land = \cup$ 
  - Definition 1:
    - $x_1 \le x_2$ , Gen  $\cup (x_1 Kill) \le Gen \cup (x_2 Kill)$
  - Definition 2:
    - (Gen  $\cup$  (x<sub>1</sub> Kill) )  $\cup$  (Gen  $\cup$  (x<sub>2</sub> Kill) ) = (Gen  $\cup$  ((x<sub>1</sub>  $\cup$  x<sub>2</sub>) - Kill))

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

- e.g., reaching definition for two definitions in program
- suppose: f<sub>x</sub>: Gen<sub>x</sub> = {d<sub>1</sub>, d<sub>2</sub>}; Kill<sub>x</sub>= {}

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

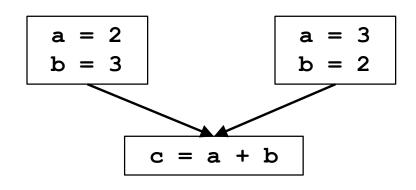
result(second iteration) ≤ result(first iteration)



 $[\mathbf{x}_1 \le x_2 \text{ iff } x_2 \to x_1]$ 

## Distributivity

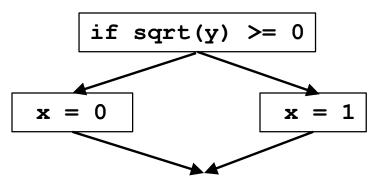
- A framework (F, V,  $\land$ ) is **distributive** if and only if
  - $f(x \land y) = f(x) \land f(y)$
  - i.e. merge input, then apply f is **equal to** apply the transfer function individually then merge result
- Example: Constant Propagation is NOT distributive



#### **Data Flow Analysis**

- Definition
  - Let  $f_1, ..., f_m : \in F$ , where  $f_i$  is the transfer function for node *i* 
    - $f_p = f_{n_k} \cdot \dots \cdot f_{n_1}$ , where **p** is a path through nodes  $n_1, \dots, n_k$
    - $f_p$  = 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.



But determining all possibly executed paths is undecidable

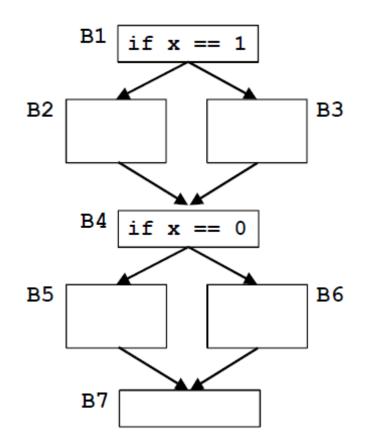
# Meet-Over-Paths (MOP)

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

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

- a path exists as long there is an edge in the code
- consider more paths than necessary
- 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:
  - $FP \leq MFP \leq MOP \leq Perfect-solution$
  - FP, MFP are safe
  - $in(b) \leq MOP(b)$

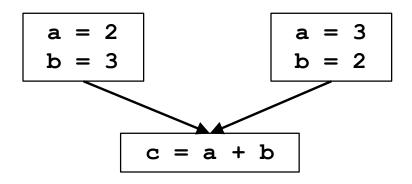
# **Partial Correctness of Algorithm**

- 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:  $IN[n_k] \le f_{n_{k-1}}(f_{n_{k-2}}(\dots f_{n_1}(IN[entry])))$
    - $IN[n_{k+1}] = OUT[n_k] \land ...$

$$\leq \text{OUT}[n_k] \\ \leq f_{n_k}(\text{IN}[n_k]) \\ \leq f_{n_{k-1}}(f_{n_{k-2}}(\dots f_{n_1}(\text{IN}[\text{entry}])))$$

#### Precision

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



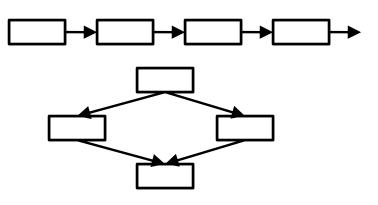
Monotone but not distributive: behaves as if there are additional paths

#### **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 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] = (out[p]), for all predecessors p of i
          oldout = out[i]
          out[i] = f_i(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 2018

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