#### Introduction to Abstract Interpretation

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### Course Recap

#### What you have learned so far

#### **Operational Semantics**

- How will a given program behave on a given input?
- This is the ground truth for any analysis

#### Types

- Annotations describe properties of the data that can be refered by a variable.
- Easy to describe properties that are global to the execution, but only one variable at a time (at least with the machinery we have seen here)
- Properties are fixed a priori by the type system designer
- •Actual analysis is cheap
- Annotations can often be inferred

#### **Program Logics**

- •Annotations describe properties of the state at a given point in the program.
- Easy to describe complex properties of the overall program state, but messy to describe properties that hold over time
- Logic provides a rich language for properties
- •Actual analysis can be expensive
- •Annotations are hard to infer

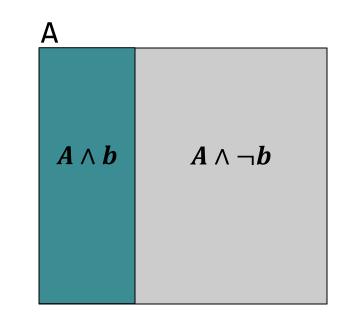
#### Some motivation

```
{true}
y=0;
while(x<10){
    x = x+1;
    y = y+2;
}
{even(y)}</pre>
```

 $\vdash \{A \land b\}c \{A\}$  $\vdash \{A\}while \ b \ do \ c \ \{A \land not \ b\}$ 

What is the loop invariant? Intuition:

- The loop invariant is a set of states
- C transforms elements in  $A \wedge b$  to other elements in A.



# Simplifying the problem

{true}
y=0;
while(x<10){
 x = x+1;
 y = y+2;
}
{even(y)}</pre>

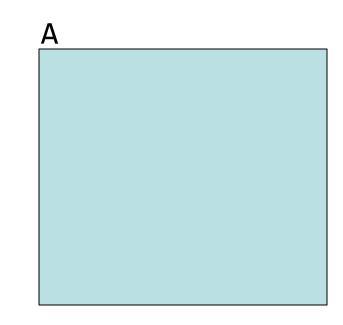
 $\vdash \{A \land b\}c \{A\}$  $\vdash \{A\}while \ b \ do \ c \ \{A \land not \ b\}$ 

#### This rule is strictly weaker

- Many correct programs can't be proved with it

#### Simpler Intuition:

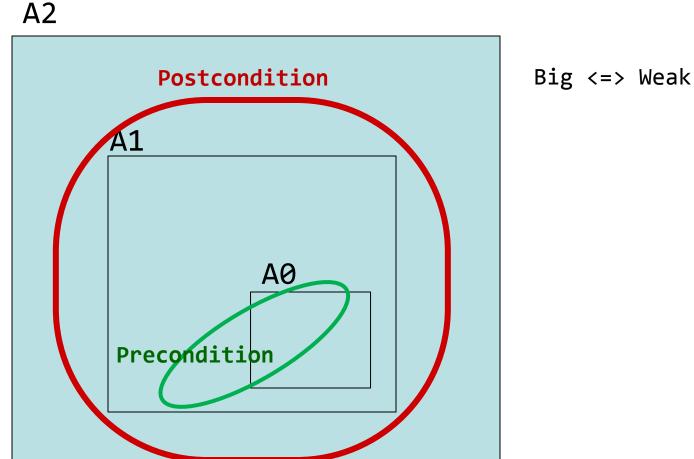
- The loop invariant is a set of states
- C transforms elements in A to other elements in A.



### Discovering the invariant

There may be many candidates for A

- True is always an invariant



### Discovering the invariant

We want a set *A* such that  $\dashv$  {*A*}*c* {*A*}

- It should be small enough to prove the postcondition (strong)
- But big enough to prove the precondition (weak)

Let  $F(P) = wpc(c, P) \land Post$ 

- Then what we want is a greatest fixpoint solution of A=F(A)

Convergence properties

- Can we always find such solutions?

Forward vs. Backward

- When is it better to use wpc vs. spc?

Precision

- How do we minimize the loss of precision?

### Partial Orders

#### Set P

Partial order  $\leq$  such that  $\forall x, y, z \in P$ 

- $x \le x$  (reflexive)
- $x \le y$  and  $y \le x$  implies x = y (asymmetric)
- $x \le y$  and  $y \le z$  implies  $x \le z$

(transitive)

Can use partial order to define

- Upper and lower bounds
- Least upper bound
- Greatest lower bound

### **Upper Bounds**

#### If $S \subseteq P$ then

- $x \in P$  is an upper bound of S if  $\forall y \in S. y \le x$
- $x \in P$  is the least upper bound of S if
  - x is an upper bound of S, and
  - $x \le y$  for all upper bounds y of S
- $\vee$  join, least upper bound, lub, supremum, sup
  - ${\scriptstyle \vee}$  S is the least upper bound of S
  - $x \lor y$  is the least upper bound of  $\{x,y\}$
- Often written as ⊔ as well

#### Lower Bounds

#### If $S \subseteq P$ then

- $x \in P$  is a lower bound of S if  $\forall y \in S. x \leq y$
- $x \in P$  is the greatest lower bound of S if
  - x is a lower bound of S, and
  - $y \le x$  for all lower bounds y of S
- $\wedge$  meet, greatest lower bound, glb, infimum, inf
  - $\bullet\ \wedge$  S is the greatest lower bound of S
  - $x \land y$  is the greatest lower bound of  $\{x,y\}$
- Often written as ⊓ as well

### Covering

x < y if  $x \le y$  and  $x \neq y$ 

x is covered by y (y covers x) if

- x < y, and
- $x \le z < y$  implies x = z

Conceptually,

- y covers x if there are no elements between x and y

#### Lattices

If  $x \land y$  and  $x \lor y$  exist for all  $x,y \in P$ then P is a lattice

If  $\land$ S and  $\lor$ S exist for all S  $\subseteq$  P then P is a complete lattice

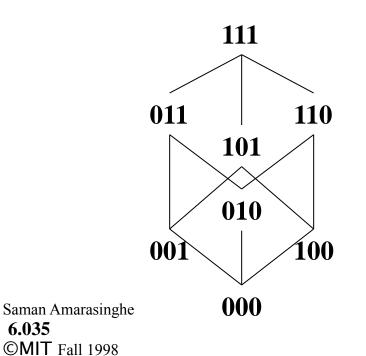
All finite lattices are complete Example of a lattice that is not complete

- Integers I
- For any x,  $y \in I$ ,  $x \lor y = max(x,y)$ ,  $x \land y = min(x,y)$
- But  $\lor$  I and  $\land$  I do not exist
- $I \cup \{+\infty, -\infty\}$  is a complete lattice

#### Example

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 $P = \{000, 001, 010, 011, 100, 101, 110, 111\}$ (standard boolean lattice, also called hypercube)  $x \le y$  if (x bitwise and y) = x



#### Hasse Diagram

- If y covers x
  - Line from y to x
  - y above x in diagram

#### Top and Bottom

Greatest element of P (if it exists) is top (T) Least element of P (if it exists) is bottom ( $\perp$ )

#### Connection Between $\leq$ , $\land$ , and $\lor$

The following 3 properties are equivalent:

- $x \le y$
- $\mathbf{x} \lor \mathbf{y} = \mathbf{y}$
- $\mathbf{x} \wedge \mathbf{y} = \mathbf{x}$

#### Chains

A set S is a chain if  $\forall x, y \in S. \ y \le x \text{ or } x \le y$ 

P has no infinite chains if every chain in P is finite

#### Product Latices

Given two latices L and Q, the product can easily be made a latice

 $(l_1, q_1) \sqsubseteq (l_2, q_2) \Leftrightarrow l_1 \sqsubseteq l_2 \text{ and } q_1 \sqsubseteq q_2$ 

For vectors of L, defining a latice is also easy

$$\langle l_1, l_2, \dots, l_k \rangle \subseteq \langle t_1, t_2, \dots, t_k \rangle \Leftrightarrow \forall_{i \in [1,k]} l_i \subseteq t_i$$

### Back to our problem

<pre>{true} y=0; while(x&lt;10){     x = x+1;     y = y+2; }</pre>	<i>x</i> =	$ \left(\begin{array}{c} \top \\ odd \\ even \\ \bot \end{array}\right) $	defini defini	odd or evei itely odd tely even o cares	n
} {even(y)}			-	_	
A latice of predicates: $(x = 1 \text{ or } p \text{ odd } T)$			odd		
- $\langle (x = \bot, even, odd, \top) \rangle$ • Ex: $\langle x = even, y = odd \rangle \sqsubseteq \langle$	$x = \top, y$	$= odd \rangle$	odd	even	
What does this have to d	o witł	n our pi	ر ?oblem		

1 1

#### Latices and fixpoints

Order Preserving (Monotonic) Function:  $x \sqsubseteq y \Rightarrow f(x) \sqsubseteq f(y)$ 

Now, let  $x_{\perp}$  be the least fixed point of  $f: L \to L$ - so  $f(x_{\perp}) = x_{\perp}$ 

Now, let  $x_0 = \bot$  and  $x_i = f(x_{i-1})$ 

- By induction,  $x_i \sqsubseteq x_{\perp}$
- Also, the chain  $x_i$  is an ascending chain
- If L has no infinite ascending chains, sooner or later  $x_i = x_{i+1} = x_{\perp}$

Same trick works for greatest fixed point!

- But then you have to start with  $x_0 = T$ 

### Back to our problem

<pre>{true} y=0; while(x&lt;10){     x = x+1;     y = y+2;</pre>	$x = \begin{cases} \top \\ odd \\ even \\ \bot \end{cases}$	Could be odd or even definitely odd definitely even who cares		L
} {even(y)}	-	-	T	
A latice of predicates	•	K		
- <(x = $\perp$ , even, odd, $\top$ )>		odd	even	
• Ex: $\langle x = even, y = odd \rangle \sqsubseteq$	$\langle x = \top, y = odd \rangle$			
		-	L	
We now have a recipe to	e	-		
$\Lambda = 1 \circ n \circ \sigma = E(D) - \mu \circ \sigma \circ (\sigma D)$	) A Doct in mono	tonio in ou	in lation	

As long as  $F(P) = wpc(c, P) \land Post$  is monotonic in our latice -

. .

#### Finding a fixpoint

$$x = \begin{cases} odd \\ even \\ \bot \end{cases}$$

Т

Could be odd or even definitely odd definitely even who cares

 $F(P) = wpc(c, P) \land Post$ 

- $P_0 = \{x = T, y = T\}$
- $P_1 = \{x = T, y = even\}$
- $P_2 = \{x = \top, y = even\}$
- Success!

## Complicating things a bit

$$\{x = T, y = T\}$$
  
y=0; t=1; } c0  
while(x<10){  
x = x+1;  
y = y+2; } c1  
if(x=5){  
t=t+2; } c2  
} c2  
} else{  
y = t+1; } c3  
}  
x = T, y = even}

 $\frac{\vdash \{A \land b\}c_1 \{B\}}{\vdash \{A\} if \ b \ then \ c_1 else \ c_2 \{B\}}$ 

Relaxed Rule

 $\vdash \{A \land b\}c_1 \{B\} \vdash \{A \land not \ b\}c_2 \{B\} \\ \vdash \{A\}if \ b \ then \ c_1else \ c_2 \{B\}$ 

 $F(P) = wpc(c, P) \land Post$ = wpc(c1, wpc(c2, P) \land wpc(c3, P)) \land Post

### Dataflow equations

Big <=> Weak So  $A \Rightarrow B$ is equivalent to  $A \sqsubseteq B$ 

 $\{x = \top, y = \top\}$  <-P1 - C0 y=0; t=1; while (x < 10)<-P2 x = x+1;- C1 y = y+2;<-P3 if(x=5){ **C**2 t=t+2; }else{ y = t+1;- C3 <-P2

 $F(P) = wpc(c, P) \land Post$ =  $wpc(c1, wpc(c2, P) \land wpc(c3, P)) \land Post$  $p1 \sqsubseteq wpc(c0, p2)$  $p2 \sqsubseteq wpc(c1, p3)$  $p3 \sqsubseteq wpc(c2, p2) \land wpc(c3, p2)$  $p2 \sqsubseteq p5$ 

 $p2 \sqsubseteq wpc(c1, p3) \land p5$ 

 ${x = T, y = even}$  <-P5

#### Dataflow equations

$$\{x = T, y = T\} < -P1 y=0; t=1; \ \ C0 while(x<10){ < -P2 x = x+1; y = y+2; \ \ C1 if(x=5){ < -P3 t=t+2; \ \ C2 }else{ y = t+1; \ \ C3 } {x = T, y = even} < -P5$$

 $p1 \sqsubseteq wpc(c0, p2)$  $p2 \sqsubseteq wpc(c1, p3) \land p5$  $p3 \sqsubseteq wpc(c2, p2) \land wpc(c3, p2)$ 

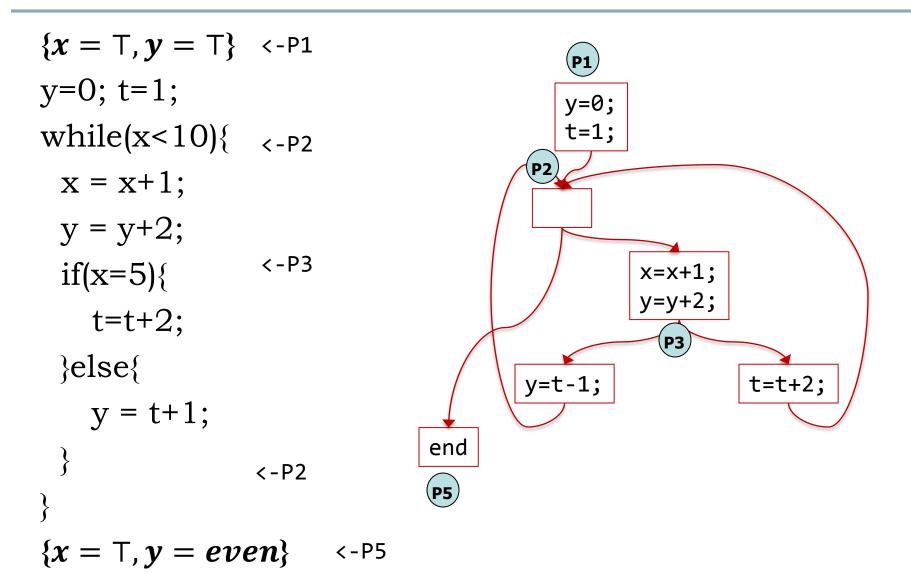
### **Dataflow Analysis**

General Analysis Framework

- Developed by Kildall in 1973
- Traditionally used for compiler optimization

Frame analysis question as a set of equations on a  $\underline{CFG}$ 

### **Control Flow Graph**



## **Control Flow Graph**

Very general program representation

- Easy to represent unstructured control flow
- Widely used by most program analysis tools for imperative languages

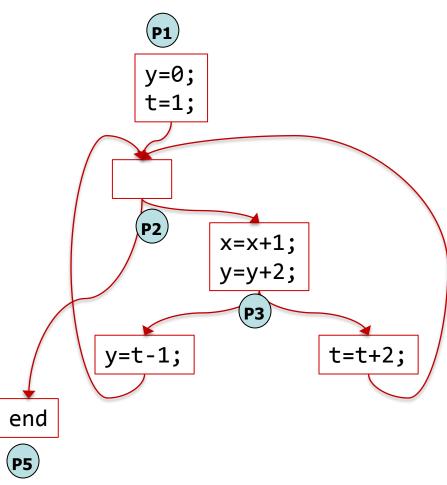
### Solution strategy

For every basic block we have an equation of the form

- $Out \subseteq F(in)$
- Use meet (∧) when many edges meet together

We can solve through "Chaotic Iteration"

- Keep a list of nodes to update
- Pick one CFG node at a time
- Update out from new in
- If out changed, add its children to the list



## Computing transfer function

So far we defined it in terms of weakest precondition.

- Or alternatively, strongest postcondition
- Too general and expensive!

We can hard-code a transfer function specific to the lattice

- For finite lattices they can be implemented cheaply in terms of bitvector operations

We can build lattices for arbitrary facts about the program

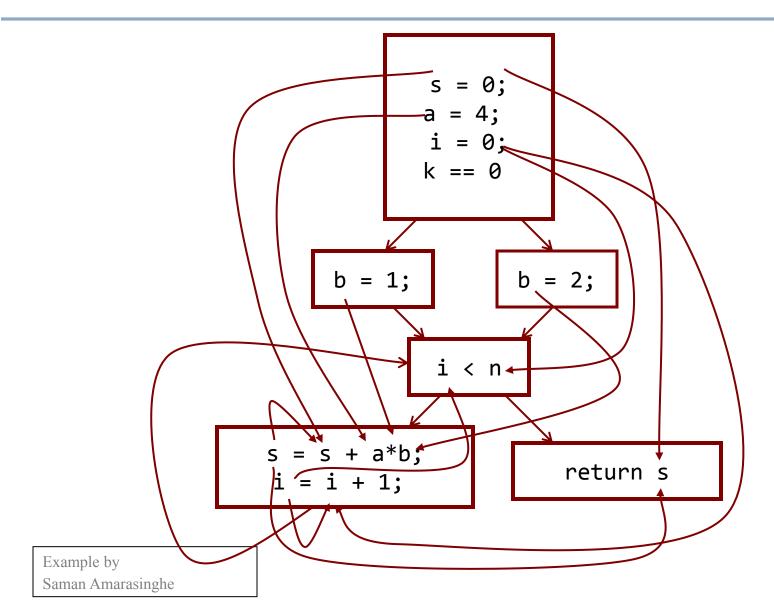
- Need to make sure our transfer functions are monotonic

## Example: Reaching Definitions

#### Concept of definition and use

- a = x+y
- is a definition of a
- is a use of x and y
- A definition reaches a use if
  - value written by definition
  - may be read by use

### **Reaching Definitions**



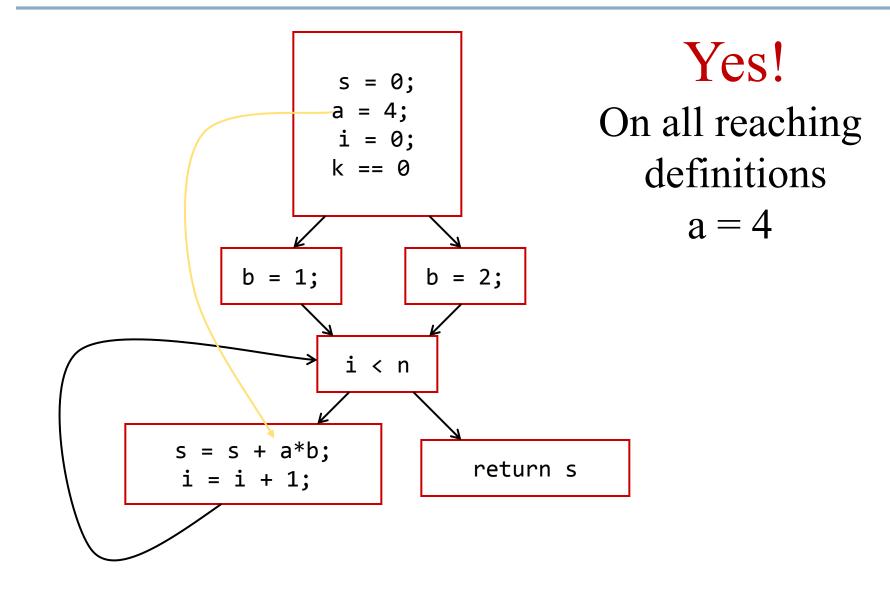
#### Reaching Definitions and Constant Propagation

Is a use of a variable a constant?

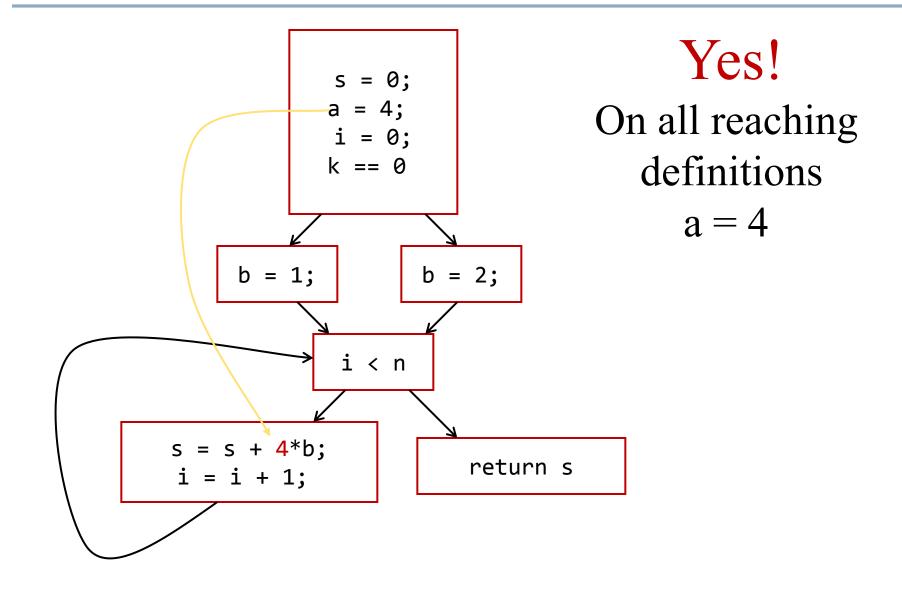
- Check all reaching definitions
- If all assign variable to same constant
- Then use is in fact a constant

Can replace variable with constant

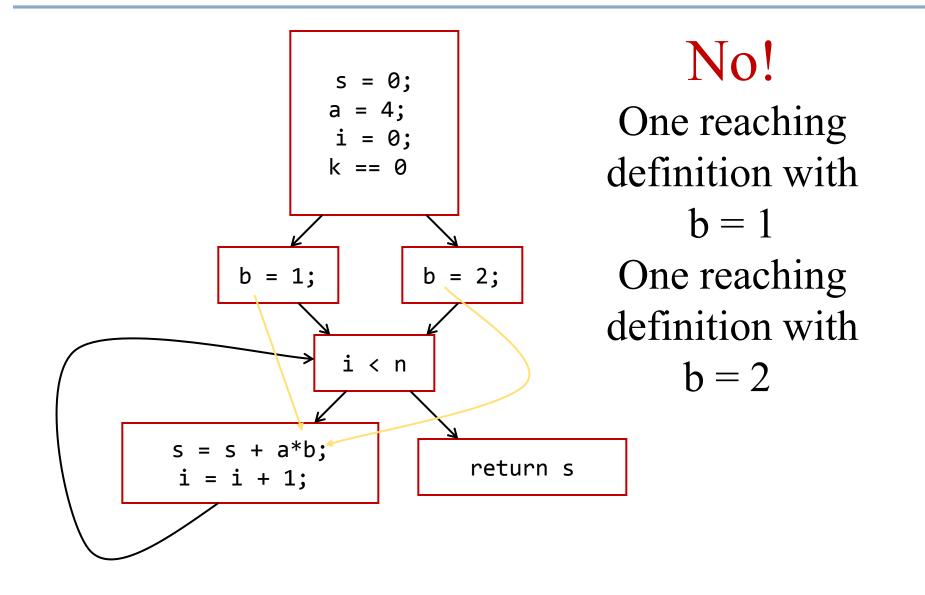
#### Is a Constant in s = s+a\*b?



### **Constant Propagation Transform**



#### Is b Constant in s = s+a\*b?

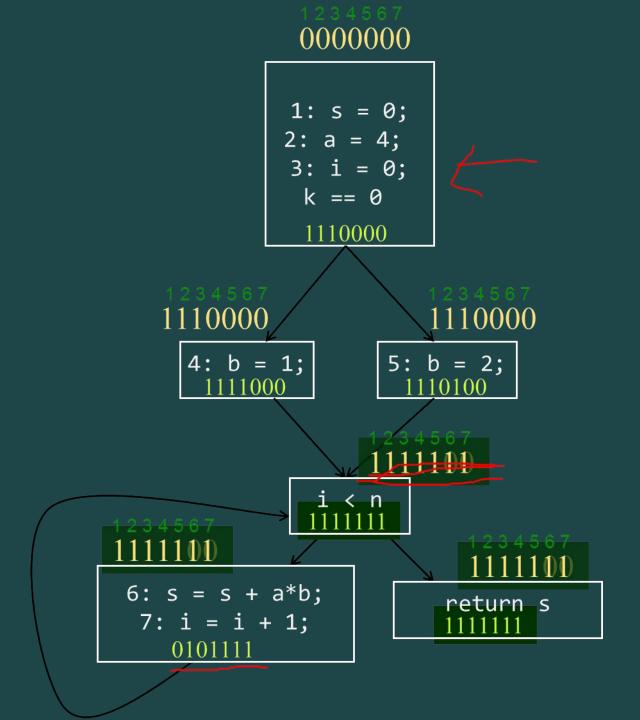


# **Computing Reaching Definitions**

#### Compute with sets of definitions

- represent sets using bit vectors
- each definition has a position in bit vector
- At each basic block, compute
  - definitions that reach start of block
  - definitions that reach end of block

Do computation by simulating execution of program until reach fixed point



### Transfer functions

#### Each basic block has

- IN set of definitions that reach beginning of block
- OUT set of definitions that reach end of block
- GEN set of definitions generated in block
- KILL set of definitions killed in block

GEN[s = s + a\*b; i = i + 1;] = 0000011

KILL[s = s + a\*b; i = i + 1;] = 1010000

Analyzer scans each basic block to derive GEN and KILL sets for each function

### **Dataflow Equations**

 $IN[b] = OUT[b1] U \dots U OUT[bn]$ 

- where b1, ..., bn are predecessors of b in CFG
- OUT[b] = (IN[b] KILL[b]) U GEN[b]
- IN[entry] = 0000000

Result: system of equations

## Solving Equations

Use fixed point algorithm Initialize with solution of OUT[b] = 0000000 Repeatedly apply equations

- IN[b] = OUT[b1] U ... U OUT[bn]
- OUT[b] = (IN[b] KILL[b]) U GEN[b]

Until reach fixed point

Until equation application has no further effect

Use a worklist to track which equation applications may have a further effect

### Questions

Does the algorithm halt?

- yes, because transfer function is monotonic
- if increase IN, increase OUT
- in limit, all bits are 1

If bit is 0, does the corresponding definition ever reach basic block?

If bit is 1, is does the corresponding definition always reach the basic block?

#### 6.820 Fundamentals of Program Analysis Fall 2015

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