

# Overview

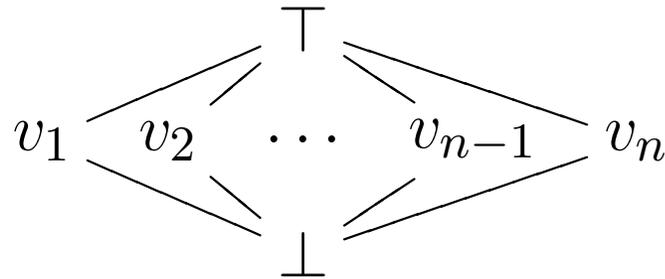
- Abstract Interpretation
  - What is it, intuitively?
  - Relationship to dataflow analysis
- Value ranges
- Fixpoints and infinite lattices
  - Dataflow problems with infinite lattices
  - Widening
  - Narrowing
- Two approaches to generating correct analyses
  - Representation functions
  - Correctness relations

# Abstract Interpretation: Intuitively

- “Execute” the program on an abstract program state
  - Just like writing an interpreter, but...
  - Abstract program state represents all possible program states at a particular program point
  - Covers all possible program inputs
- What to do for multiple incoming control-flow edges?  
Join!
- What to do for program loops? Iterate!

# Relationship to Dataflow Analysis

- Abstract interpretation is a dataflow analysis
  - A different way to construct *correct* analyses
  - Induces a specific ordering on the “worklist”
- Abstract program states are typically complete lattices
  - Trivial join lattice for any domain  $V$  with values  $v_1, v_2, \dots, v_n \in V$  implies an abstract interpretation.



- Will permit lattices with infinite height
- Can combine multiple analyses into a single lattice
- Trivial example: constant propagation

# Generating Analyses

- Start with the values in domain  $V$  you are interested in.  
Example: The integers  
 $\mathbb{Z} = \{\dots, -3, -2, -1, 0, 1, 2, 3, \dots\}$ .
- Next, consider the operations that can be performed on values in  $V$ , e.g.,  $+$ ,  $-$ ,  $*$ ,  $/$ . For  $v_1, v_2 \in V$  we say that  $v_1 \rightsquigarrow v_2$  if the value  $v_1$  can be transformed to  $v_2$ .
- Determine the form of the elements in the lattice  $L$ .
- Construct the operations performed on the elements of the lattice  $L$ . For  $l_1, l_2 \in V$  we say that  $l_1 \triangleright l_2$  if the lattice element  $l_1$  can be transformed to  $l_2$ .

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$$-(\top) \triangleright \top \quad -(\perp) \triangleright \perp \quad -(c) \triangleright -c$$

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  - What would  $+$  look like?

# Value Ranges

- Constant propagation is boring: we can do better.
- **Definition:** A *value range*, denoted  $[a : b]$ , represents all values  $x$  such that:

$$a \in \mathbb{Z} \cup \{-\infty\} \quad b \in \mathbb{Z} \cup \{\infty\} \quad a \leq x \leq b$$

- **Examples:**
  - $[17 : 17]$  represents the value 17.
  - $[17 : 42]$  represents any value between 17 and 42.
  - $[-\infty : -1]$  represents any negative integer.
  - $[0 : \infty]$  represents any non-negative integer.
- Is this representation more or less expressive than in constant propagation?

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- **Definition:**  $[a_1 : b_1] \sqcup [a_2 : b_2] = [\min(a_1, a_2) : \max(b_1, b_2)]$

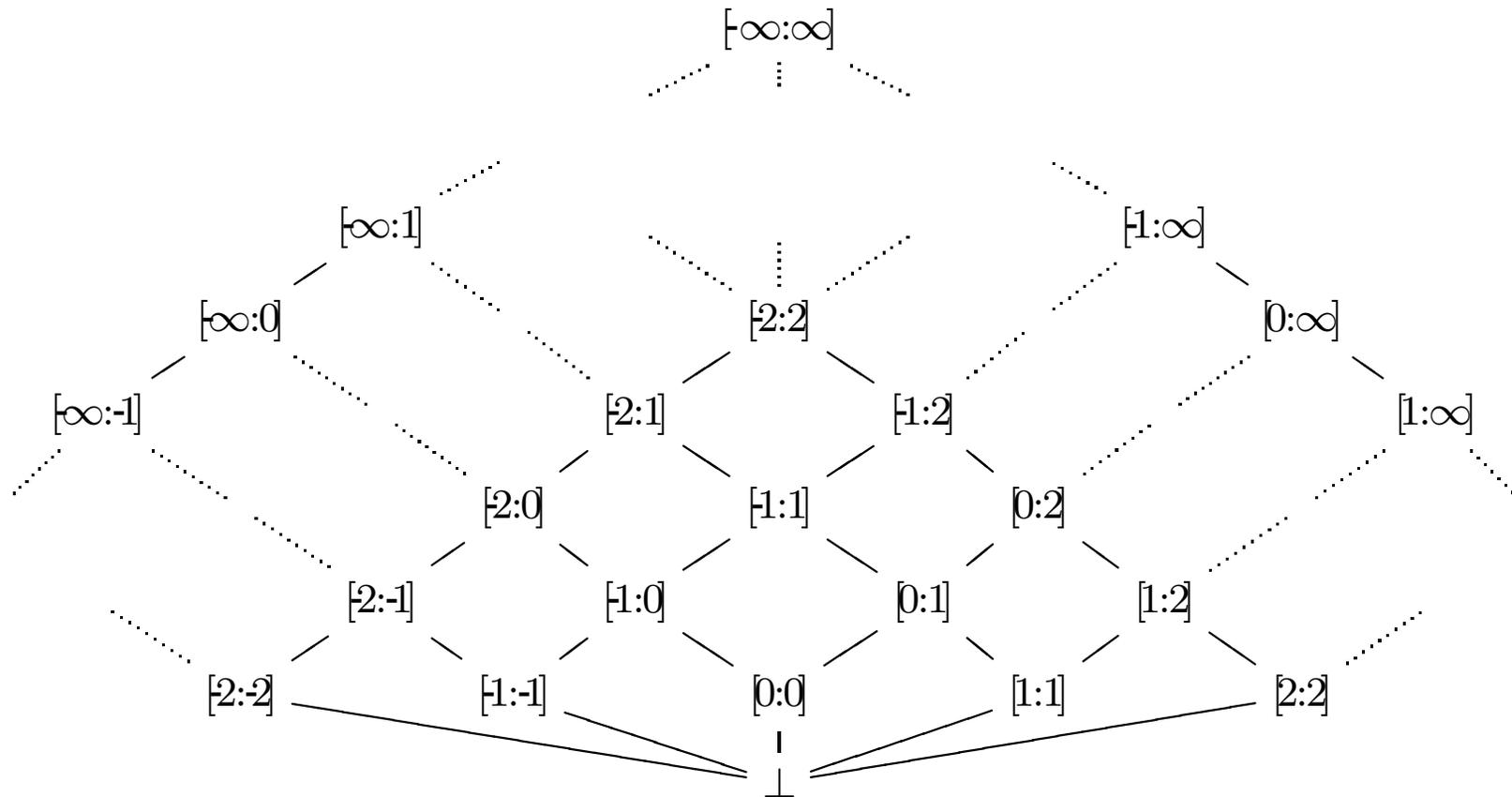
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- How wide is this lattice? How high?

# Value Range Lattice: Graphically

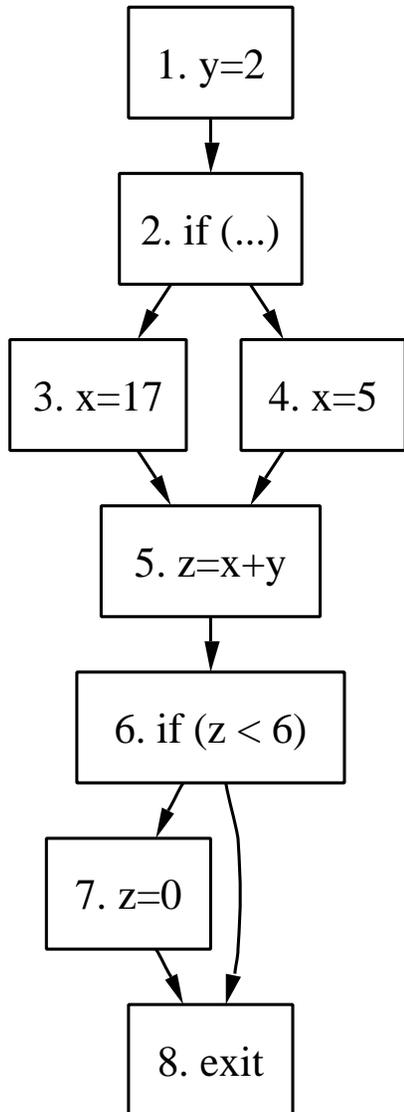


(Marvel at it. It took me *forever* to get right.)

# Value Range Operations

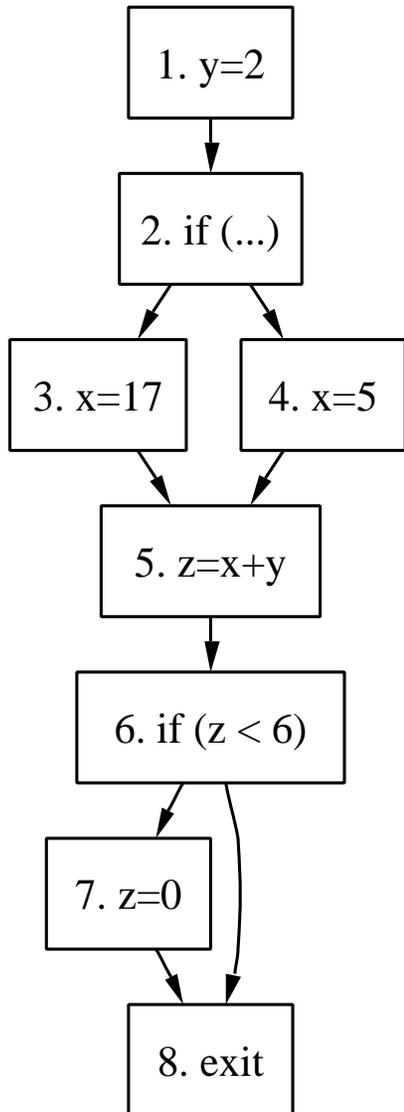
- Negation:  $-[a : b] \triangleright [-b : -a]$ .
- Addition:  $[a_1 : b_1] + [a_2 : b_2] \triangleright [a_1 + a_2 : b_1 + b_2]$
- Subtraction:  $[a_1 : b_1] - [a_2 : b_2] \triangleright [a_1 - b_2 : b_1 - a_2]$
- Multiplication:  $[a_1 : b_1] \cdot [a_2 : b_2] \triangleright$   
 $[\min(a_1 a_2, a_1 b_2, b_1 a_2, b_1 b_2) : \max(a_1 a_2, a_1 b_2, b_1 a_2, b_1 b_2)]$
- Key points to revisit later:
  - We know how to map from elements (integers) in  $V$  to elements (value ranges) in  $L$ .
  - We can prove that the operations on elements of  $V$  are “abstracted” by the operations on elements on  $L$ . Important relationship between  $\rightsquigarrow$  and  $\triangleright$ .
- But now, let’s try some abstract interpretation...

# Abstract Interpretation Example



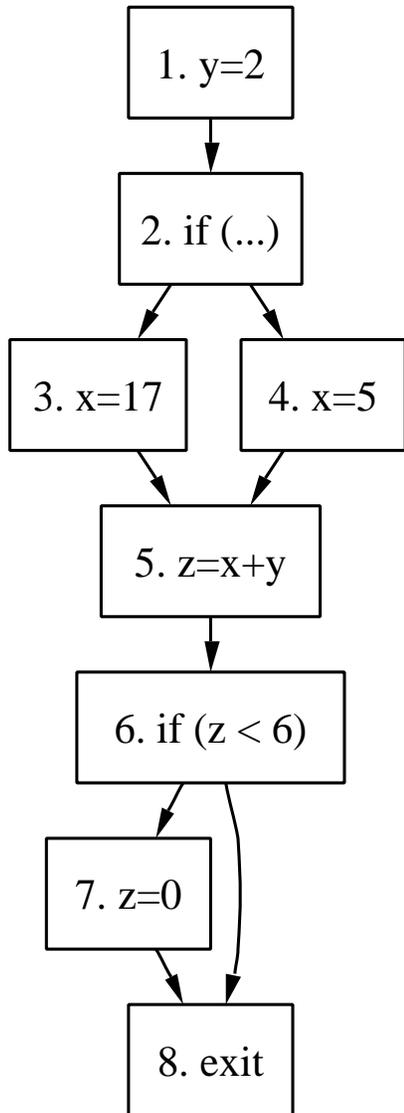
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  - Not much of an improvement.
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  - Start at entry node.

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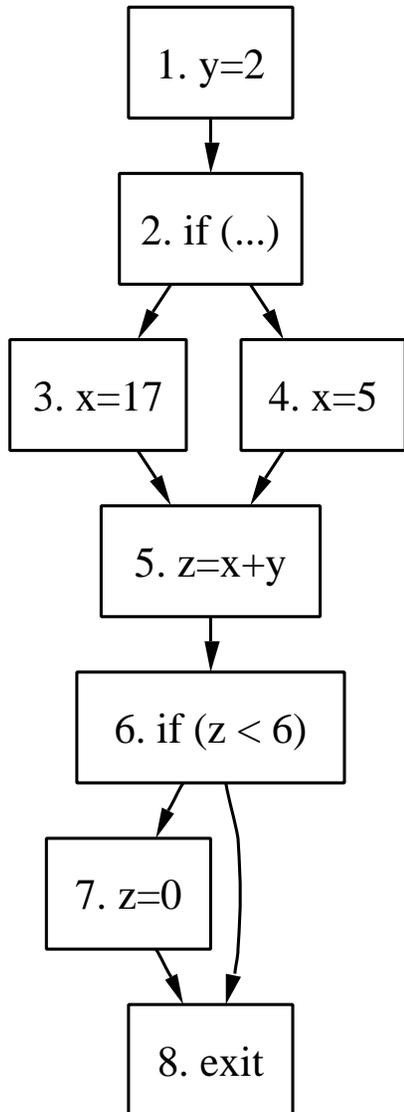
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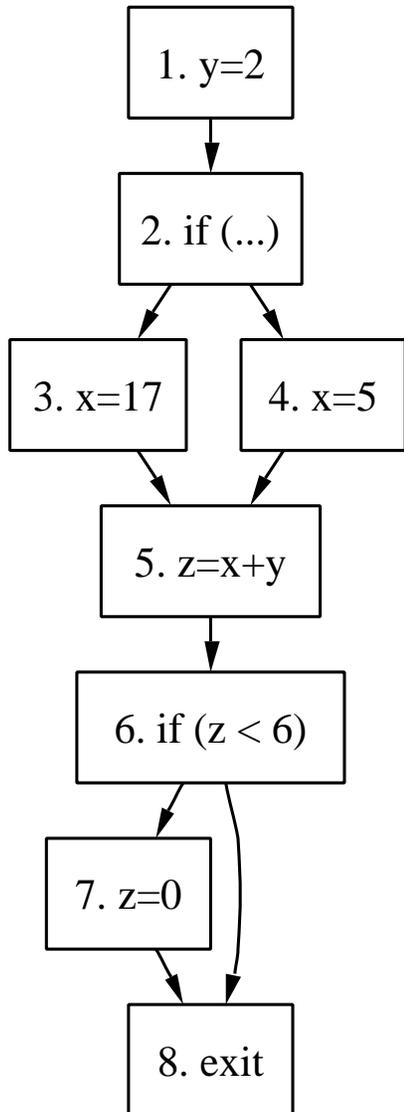
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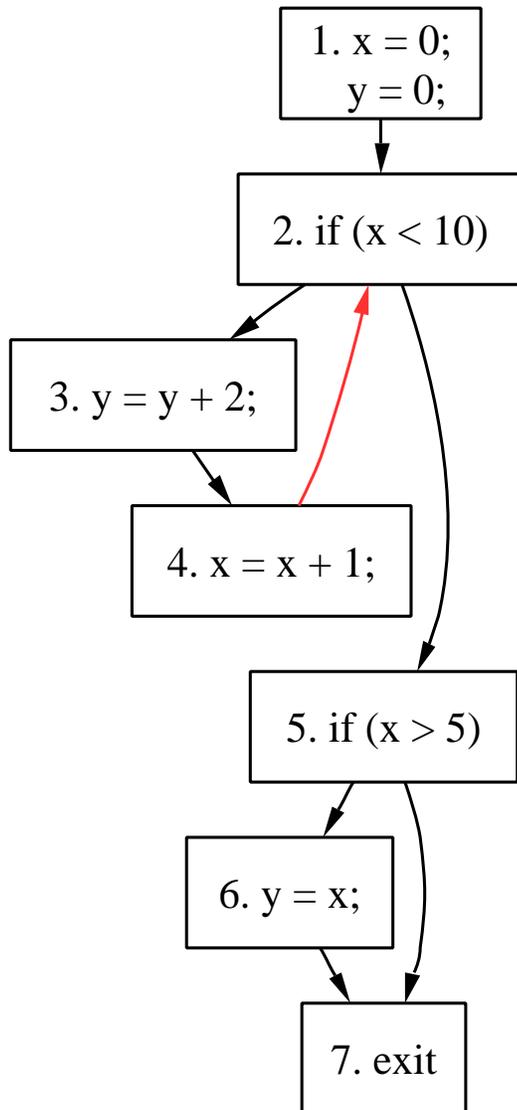
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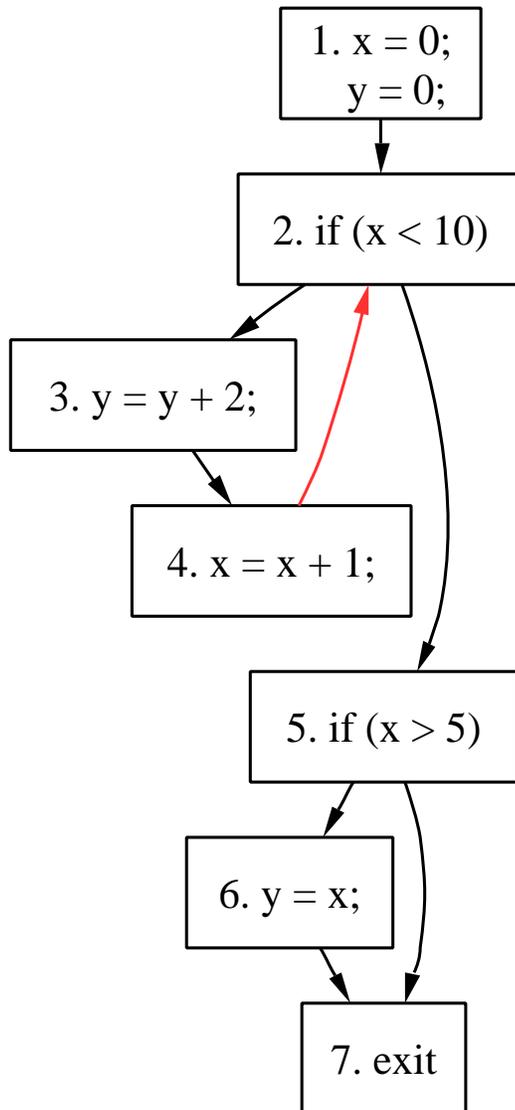
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# Analyzing Loops



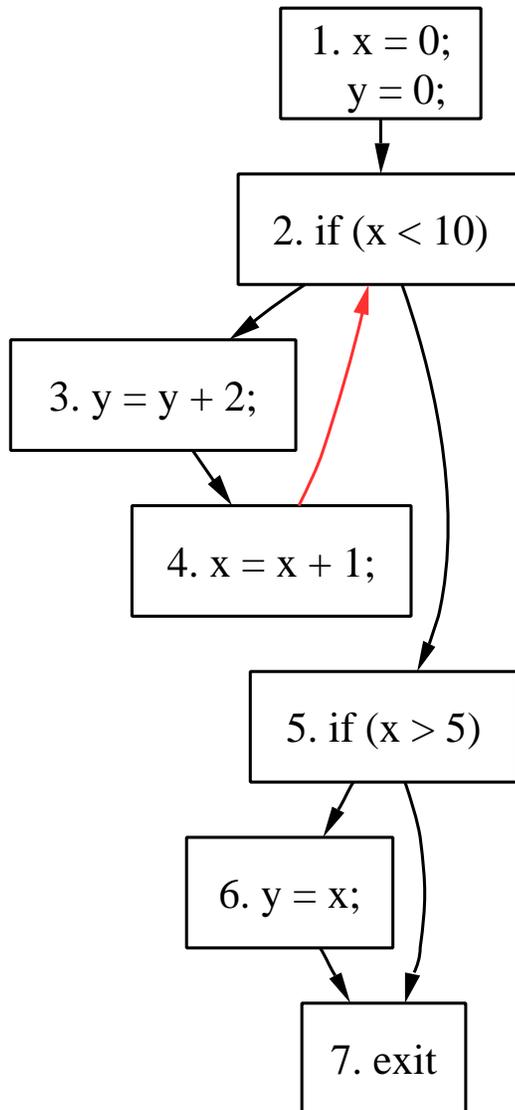
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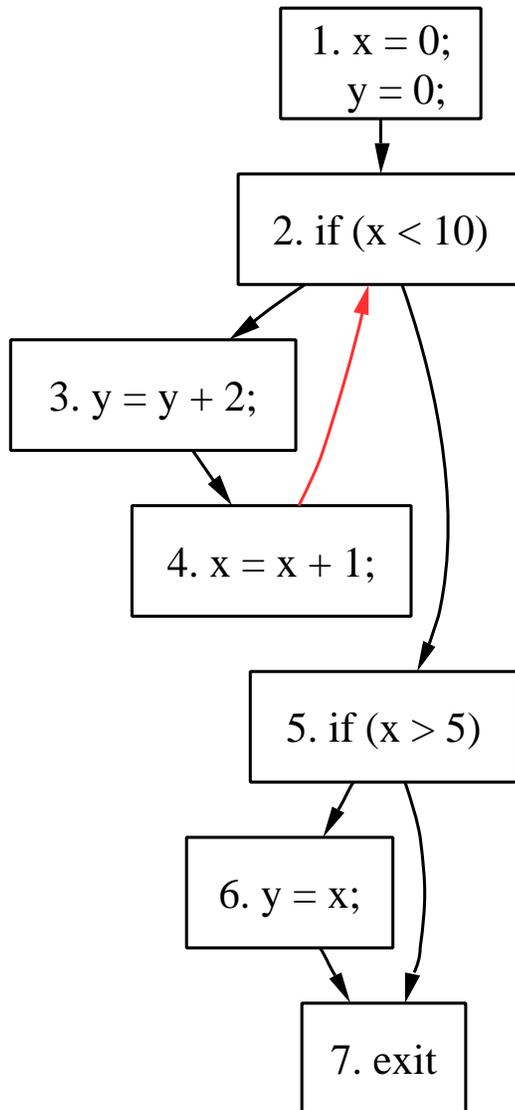
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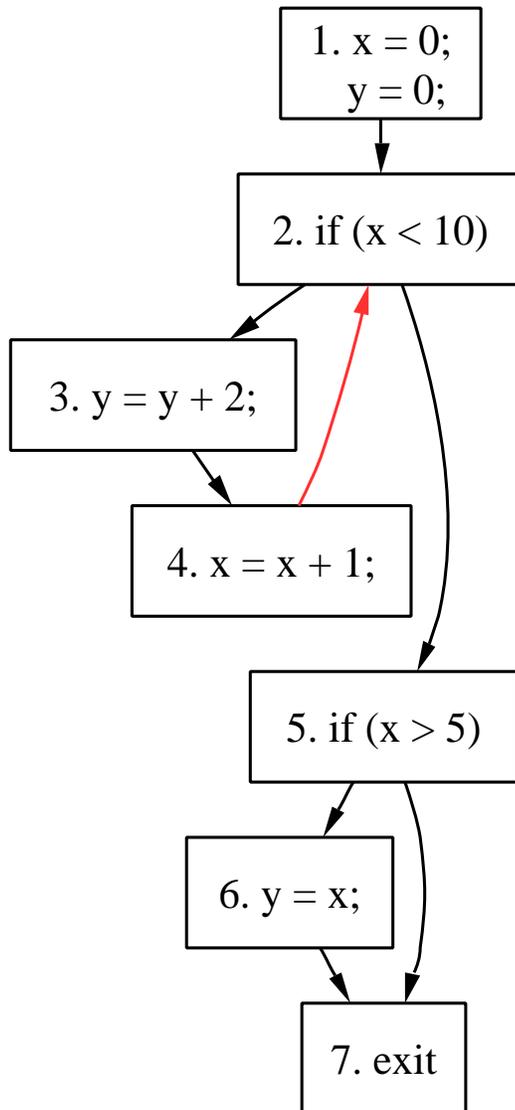
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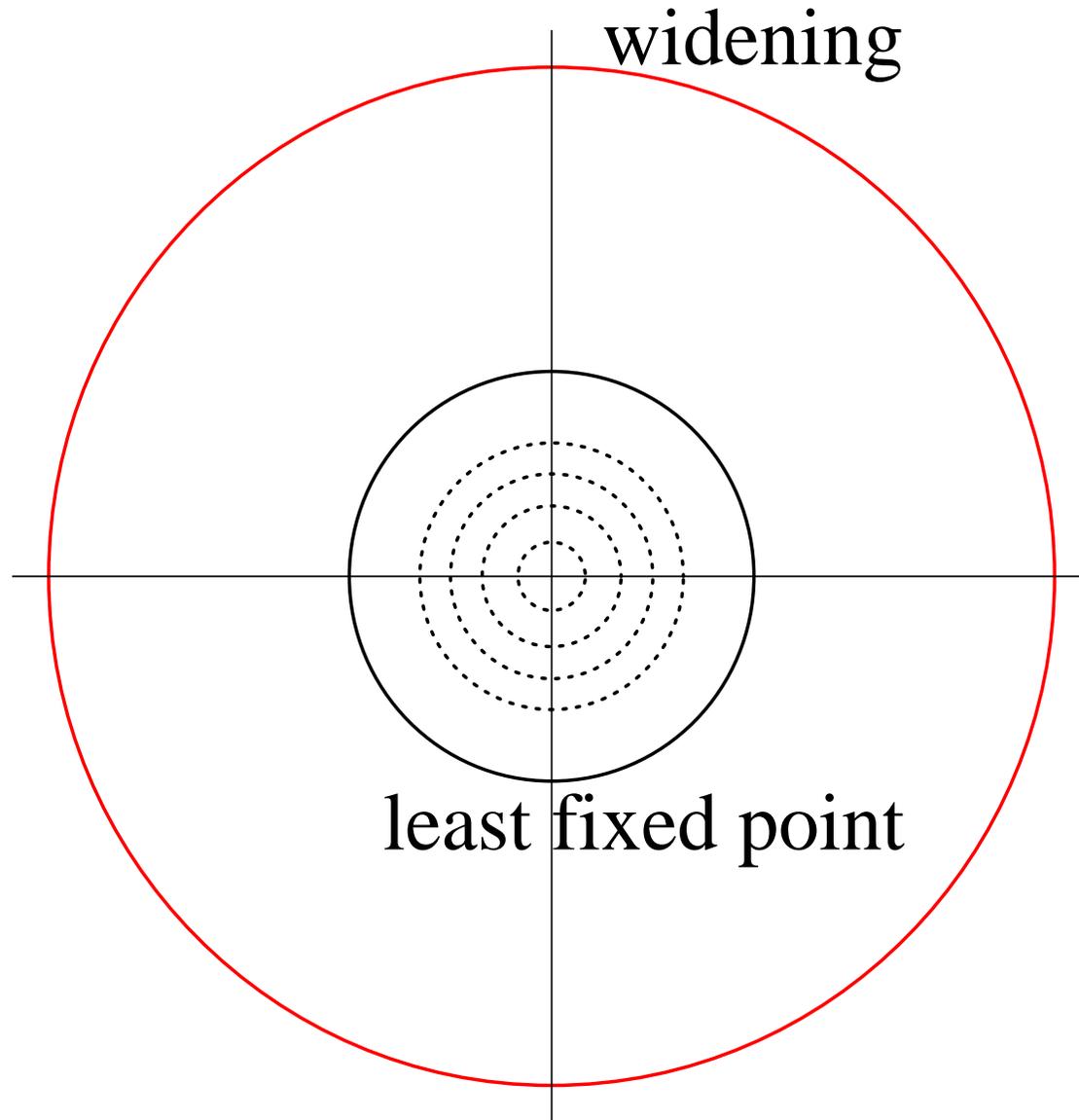
# Widening

- *Widening* reduces the number of iterations around a loop to a finite quantity, even in an infinite lattice.
- Formally,  $\nabla : L \times L \rightarrow L$  is a widening operator iff:
  - It is an upper bound operator, such that
$$\forall l_1, l_2 \in V \quad l_1 \sqsubseteq (l_1 \nabla l_2) \sqsupseteq l_2.$$
  - For all ascending chains of lattice elements  $l_1, l_2, \dots$ , the ascending chain  $l_1 \nabla l_2 \nabla l_3 \nabla \dots$  stabilizes.
- Widening operator for value ranges:

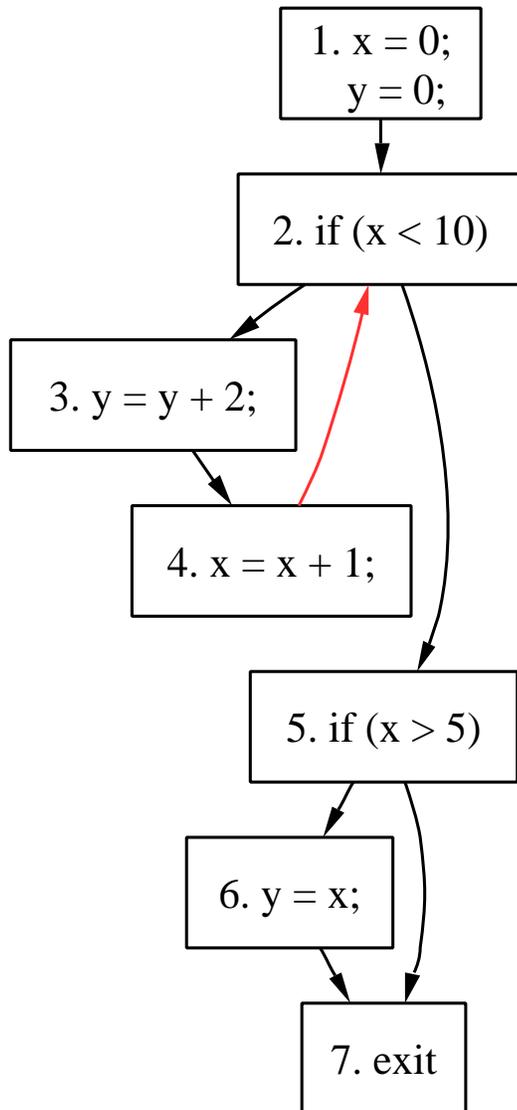
$$[a_1 : b_1] \nabla [a_2 : b_2] = [LB(a_1, a_2) : UB(b_1, b_2)]$$

$$LB(a_1, a_2) = \begin{cases} a_1 & \text{if } a_1 \leq a_2 \\ -\infty & \text{otherwise} \end{cases} \quad UB(b_1, b_2) = \begin{cases} b_1 & \text{if } b_1 \geq b_2 \\ \infty & \text{otherwise} \end{cases}$$

# Widening: Graphically



# Applying Widening Operators



- Apply  $l_1 \nabla l_2$  on back edges.  $l_1$  is the previous value (at the head of the edge) and  $l_2$  is the new value (at the tail of the edge).
- Now we get a fixed point even with our infinite lattice.
- Let's look at  $x$ :
  1.  $[0 : 0] \nabla ([0 : 0] \sqcup [1 : 1]) = [0 : \infty]$ .
  2.  $[0 : \infty] \nabla ([0 : 0] \sqcup [1 : \infty]) = [0 : \infty]$ .

# Deriving Information from Conditions

- Condition `if (x < 10)` tells us something about the value of  $x$  in the `then` and `else` branches.
- If true, we know that  $x \in [-\infty : 9]$ . If false,  $x \in [10 : \infty]$ .
- This information is *in addition to* what we already knew.
  - Meet operation  $l_1 \sqcap l_2$  computes the lattice element when both  $l_1$  and  $l_2$  describe the value.
  - What if the meet is  $\perp$ ?
- Example: We know that  $x \in [0 : \infty]$  (magically).
  - On `then` branch,  $x \in ([0 : \infty] \sqcap [-\infty : 9]) = [0 : 9]$ .
  - On the `else` branch,  $x \in ([0 : \infty] \sqcap [10 : \infty]) = [10 : \infty]$ .

# Narrowing

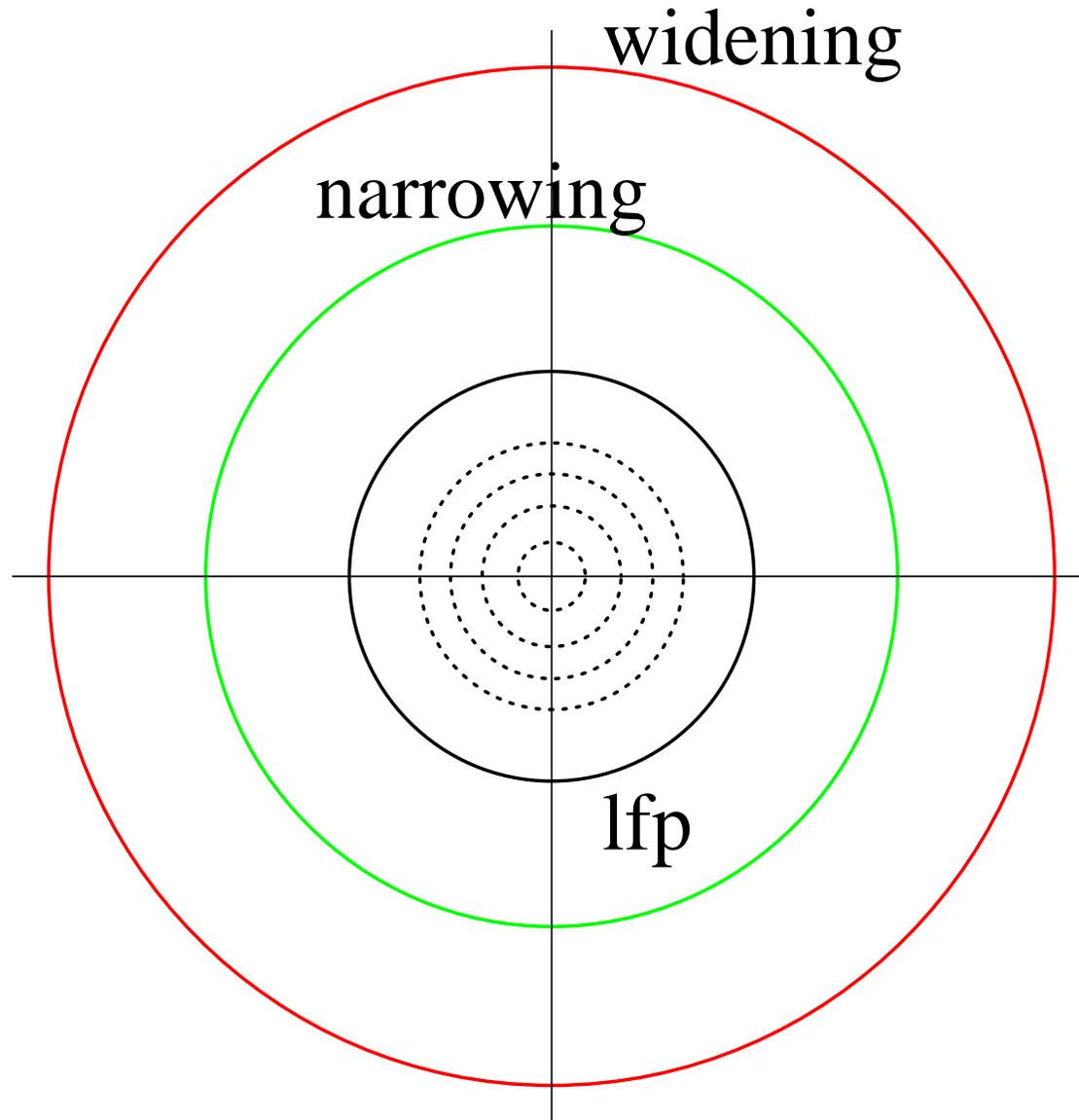
- Apply narrowing after widening to recover some information lost due to widening.
- $\Delta : L \times L \rightarrow L$  is a *narrowing* operator if:
  - $\forall l_1, l_2 \in L \quad l_2 \sqsubseteq (l_1 \Delta l_2) \sqsubseteq l_1$ , and
  - For all descending chains of lattice elements  $l_1, l_2, \dots$ , the descending chain  $l_1 \Delta l_2 \Delta l_3 \Delta \dots$  stabilizes.
- Narrowing operator for value ranges:

$$[a_1 : b_1] \Delta [a_2 : b_2] = [z_1 : z_2]$$

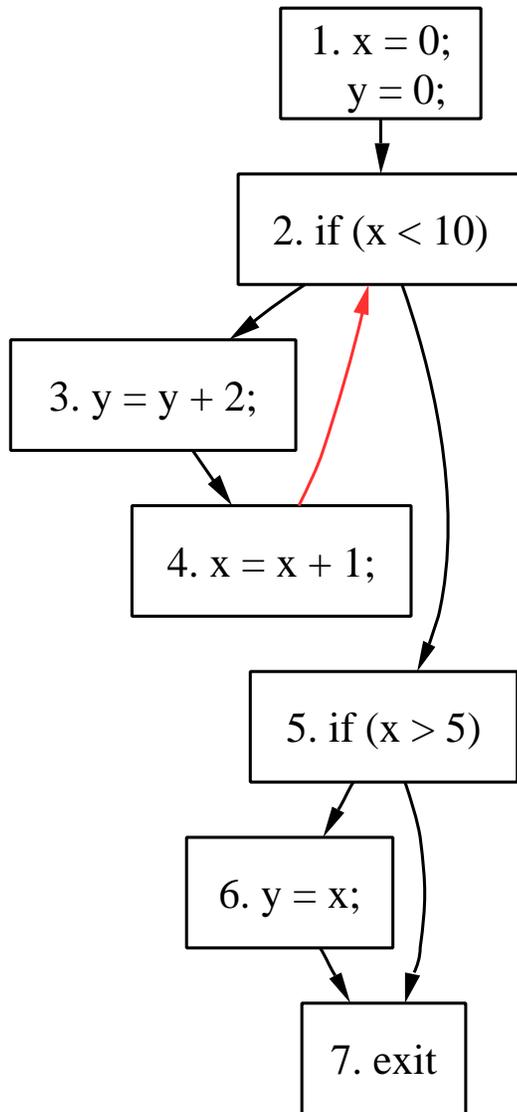
**where**  $z_1 =$  if  $a_1 = -\infty$  then  $a_2$  else  $a_1$ ,

$z_2 =$  if  $b_1 = \infty$  then  $b_2$  else  $b_1$

# Narrowing: Graphically



# Widening/Narrowing Example



- $K = \{0, 1, 2, 5, 10\}$

- Let's look at  $x$  again:

1.  $[0 : 0] \nabla ([0 : 0] \sqcup [1 : 1]) = [0 : \infty]$ .

2.  $[0 : \infty] \nabla ([0 : 0] \sqcup [1 : 10]) = [0 : \infty]$ .  
*Stable.*

3.  $[0 : \infty] \Delta [0 : 10] = [0 : 10]$ . (Interpret the loop)

4.  $[0 : 10] \Delta ([0 : 0] \sqcup [1 : 10]) = [0 : 10]$ .  
*Stable.*

- Now,  $x \in [0 : 9]$  on then branch,  $x \in [10 : 10]$  on else branch!

# A Better Widening Operator

- Let  $K$  be the set of integer constants in the program.
- Define  $\nabla$  as:

$$[a_1 : b_1] \nabla [a_2 : b_2] = [LB(a_1, a_2) : UB(b_1, b_2)]$$

$$LB(a_1, a_2) = \begin{cases} a_1 & \text{if } a_1 \leq a_2 \\ k & \text{if } a_2 < a_1 \wedge k = \max\{k \in K \mid k \leq a_2\} \\ -\infty & \text{if } a_2 < a_1 \wedge \forall k \in K : a_2 < k \end{cases}$$

$$UB(b_1, b_2) = \begin{cases} b_1 & \text{if } b_1 \geq b_2 \\ k & \text{if } b_1 < b_2 \wedge k = \min\{k \in K \mid b_2 \leq k\} \\ -\infty & \text{if } b_1 < b_2 \wedge \forall k \in K : k < b_2 \end{cases}$$

- Precision/efficiency tradeoff: more steps, but better results.

# Generating Correct Analyses

- Have shown how we can create an analysis by abstraction:
  - Abstract the value domain  $V$  with the lattice  $L$
  - Abstract all operations (collectively called  $\rightsquigarrow$ ) with  $\triangleright$ .
- How do we prove that our analysis is correct?
  - Representation functions
  - Correctness relations
- Both methods are equivalent.

# Representation Functions

- Let  $\beta : V \rightarrow L$  be a function that maps any value in  $V$  to its “best” representation in  $L$ .
- Your analysis is correct if the following is true:

$$\beta(v_1) \sqsubseteq l_1 \wedge v_1 \rightsquigarrow v_2 \wedge l_1 \triangleright l_2 \Rightarrow \beta(v_2) \sqsubseteq l_2$$

- Intuitively: If a value can be safely described by a lattice element, then any value it is transformed into can be safely described by the corresponding transformation on the lattice element.
- Can we prove this for value ranges?

# Correctness relations

- Let  $R : V \times L \rightarrow \{\text{true}, \text{false}\}$  be a *correctness* relation.
- Given  $v \in V, l \in L$ ,  $v R l$  is true when  $v$  is described by  $l$ .  
 $1R[-1 : 2] = ?, 7R[17 : 42] = ?$
- General requirement: preservation of correctness

$$v_1 R l_1 \wedge v_1 \rightsquigarrow v_2 \wedge l_1 \triangleright l_2 \Rightarrow v_2 R l_2$$

- Two more conditions for correctness when dealing with lattices:

1. Lattice preserves  $R$ :  $v R l_1 \wedge l_1 \sqsubseteq l_2 \Rightarrow v R l_2$

2. There is always a “best” approximation  $l$  for every  $v$ :

$$(\forall l \in L' \subseteq L : v R l) \Rightarrow v R (\bigsqcap L')$$

- Interesting consequence:  $v R l_1 \wedge v R l_2 \Rightarrow v R (l_1 \sqcap l_2)$

# Combining Analyses

- We mainly talk about a lattice  $L$  for values of a single variable.
- Can take the Cartesian product of several of these lattices to handle multiple variables:  
$$L' = L_1 \times L_2 \times \dots \times L_N.$$
- Variables do not need to be of the same type:  $L_1$  could be a value range lattice,  $L_2$  a boolean lattice, and  $L_3$  a points-to graph lattice.

# Abstract Interpretation Tidbits

- You can read about Galois connections to abstract interpretation in the class text, but it will hurt.
- We've only discussed forward semantics: you can do abstract interpretation backwards, and with meet lattices (everything is dual).
- We only handled the “trivial” case of widening on back edges.
  - What to do about irreducible control-flow graphs?
  - So long as you pick widening edges such that every cycle contains at least one widening edge, abstract interpretation “works”.
  - Bourdoncle studied these *chaotic* iteration strategies. NP-complete problem, but with good heuristics.

# Uses of Value Range Propagation

- Constant propagation, dead-code elimination, etc: can propagate constants and determine when conditions evaluate true or false.
- Array bounds analysis: detect bugs or remove checks that are known to be unnecessary.
- Bit width estimation: limit the sizes of registers when performing hardware synthesis.
- Static branch prediction: produce probabilities that particular branches will be taken.