

Abstract Interpretation

(Cousot, Cousot 1977)

also known as

Data-Flow Analysis

(Kildall 1973)

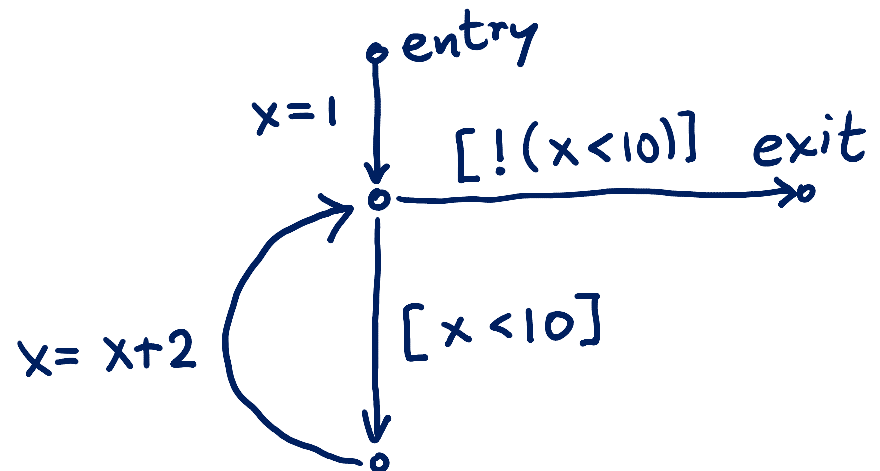
Goal of Data-Flow Analysis

Automatically compute information about the program

- Use it to report errors to user (like type errors)
- Use it to optimize the program

Works on control-flow graphs:
(like flow-charts)

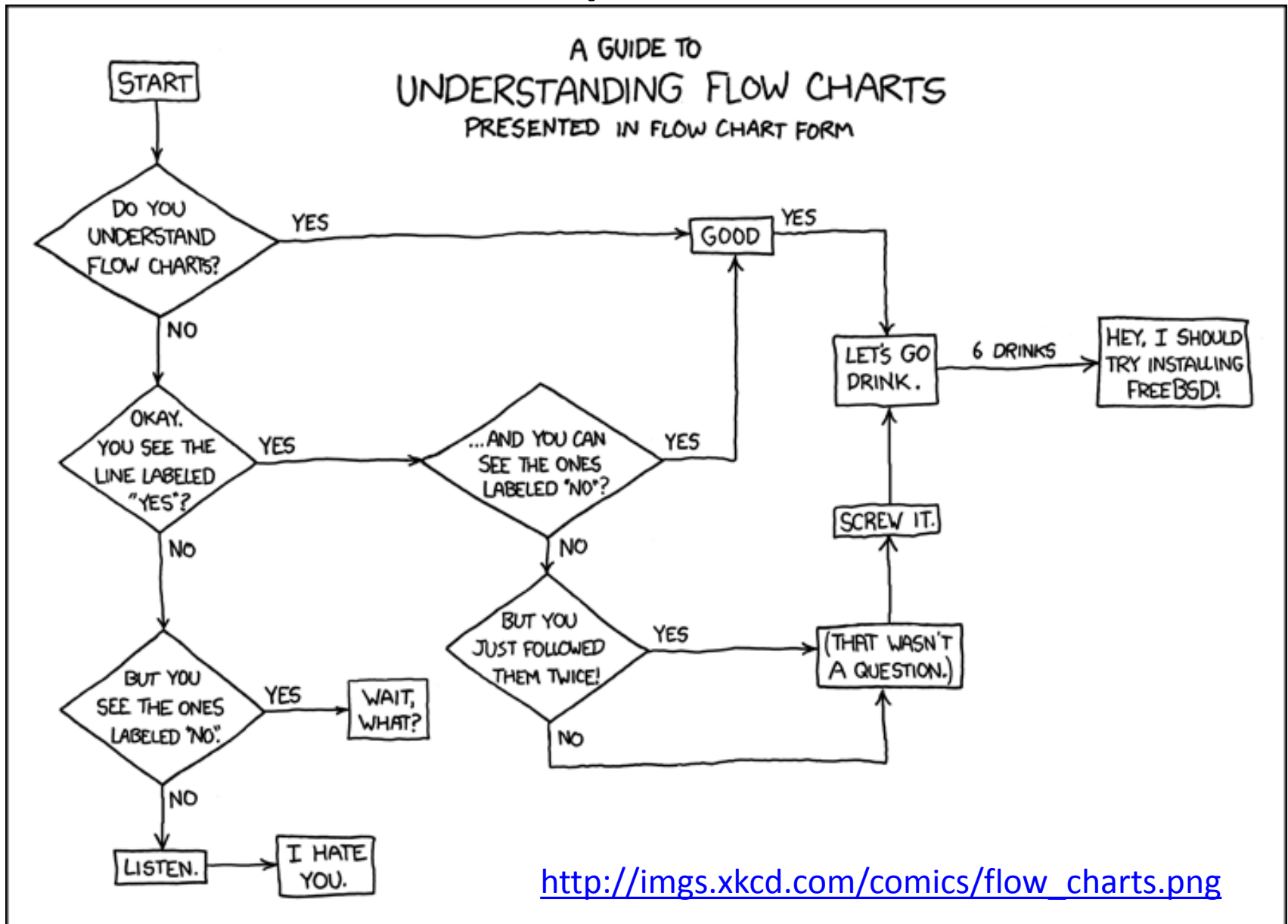
```
x = 1  
while (x < 10) {  
  x = x + 2  
}
```



Why Constant Propagation

```
int a, b, step, i;
boolean c;
a = 0;
b = a + 10;
step = -1;
if (step > 0) {
    i = a;
} else {
    i = b;
}
c = true;
while (c) {
    print(i);
    i = i + step; // can emit decrement
    if (step > 0) {
        c = (i < b);
    } else {
        c = (i > a); // can emit better instruction here
    } // insert here (a = a + step), redo analysis
}
```

Control-Flow Graphs: Like Flow Charts



Control-Flow Graph: (V,E)

Set of nodes, V

Set of edges, which have statements on them

$$(v_1, st, v_2) \in E$$

means there is edge from v_1 to v_2 labeled with statement st .

$x = 1$

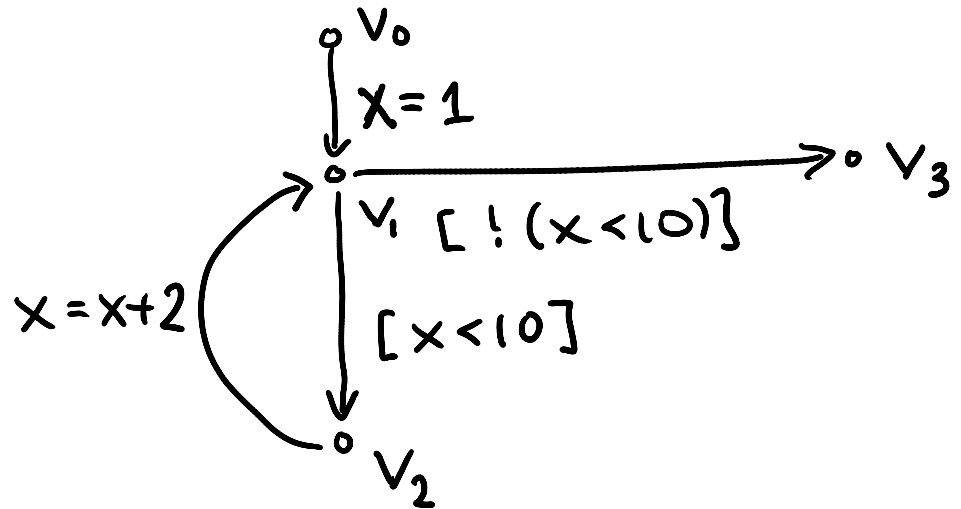
while ($x < 10$) {

$x = x + 2$

}

$V = \{v_0, v_1, v_2, v_3\}$

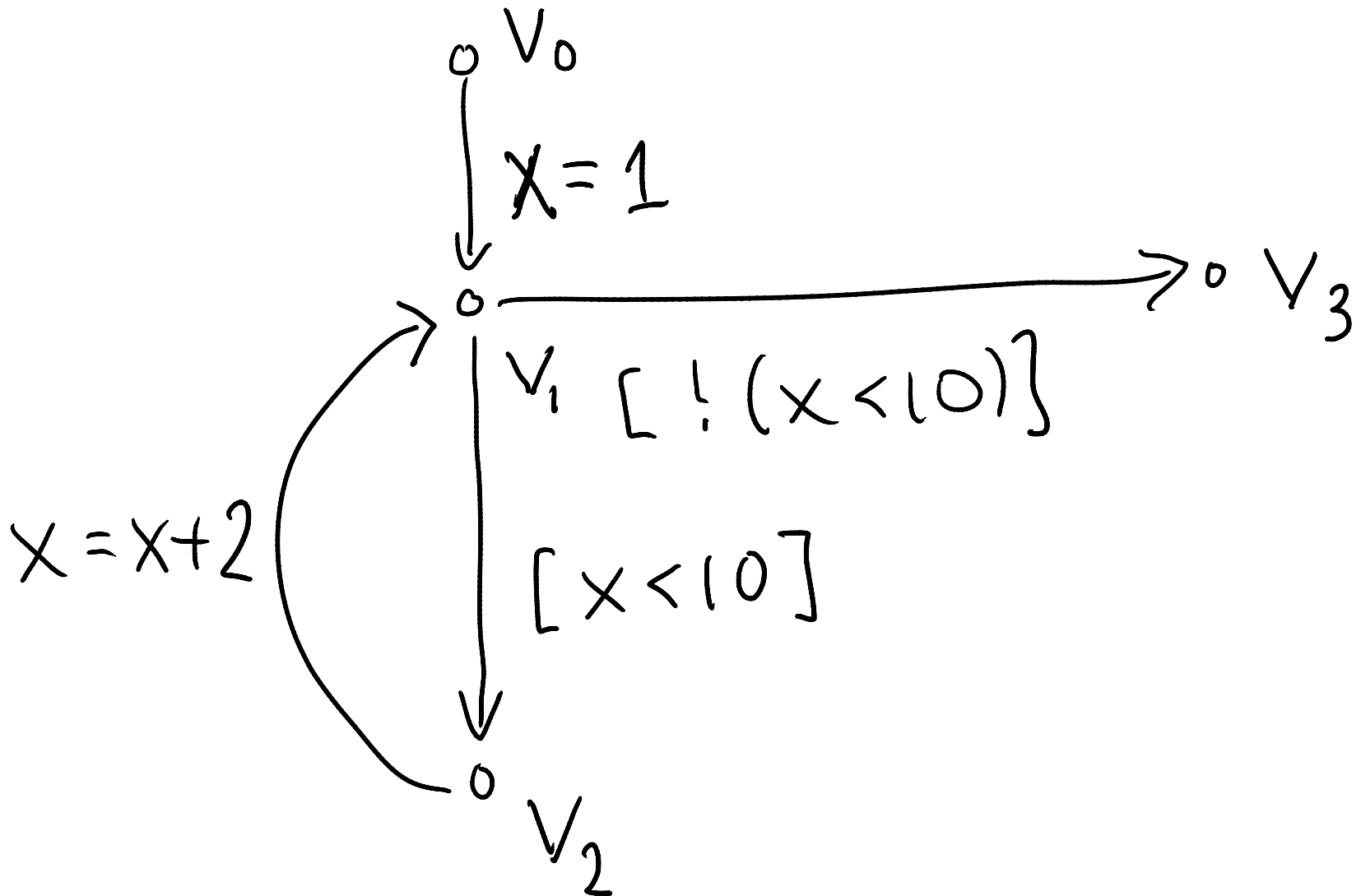
$E = \{(v_0, x=1, v_1), (v_1, [x < 10], v_2),$
 $(v_2, x=x+2, v_1), (v_1, [!(x < 10)], v_3)\}$



Interpretation and Abstract Interpretation

- Control-Flow graph is similar to AST
- We can
 - interpret control flow graph
 - generate machine code from it (e.g. LLVM, gcc)
 - abstractly interpret it: do not push values, but **approximately compute supersets of possible values** (e.g. intervals, types, etc.)

Compute Range of x at Each Point



What we see today

1. How to compile abstract syntax trees into control-flow graphs
2. Lattices, as structures that describe abstractly sets of program states (facts)
3. Transfer functions that describe how to update facts

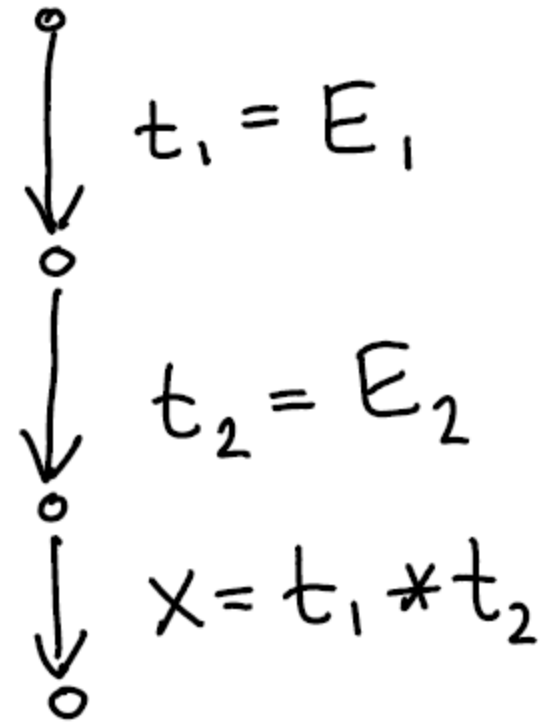
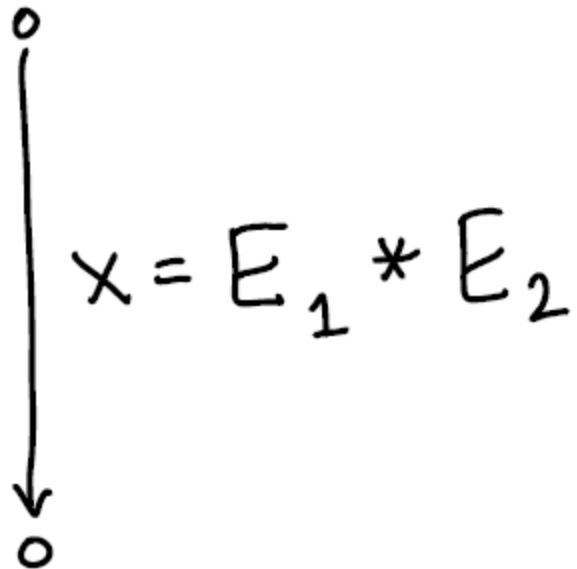
Generating Control-Flow Graphs

- Start with graph that has one entry and one exit node and label is entire program
- Recursively decompose the program to have more edges with simpler labels
- When labels cannot be decomposed further, we are done

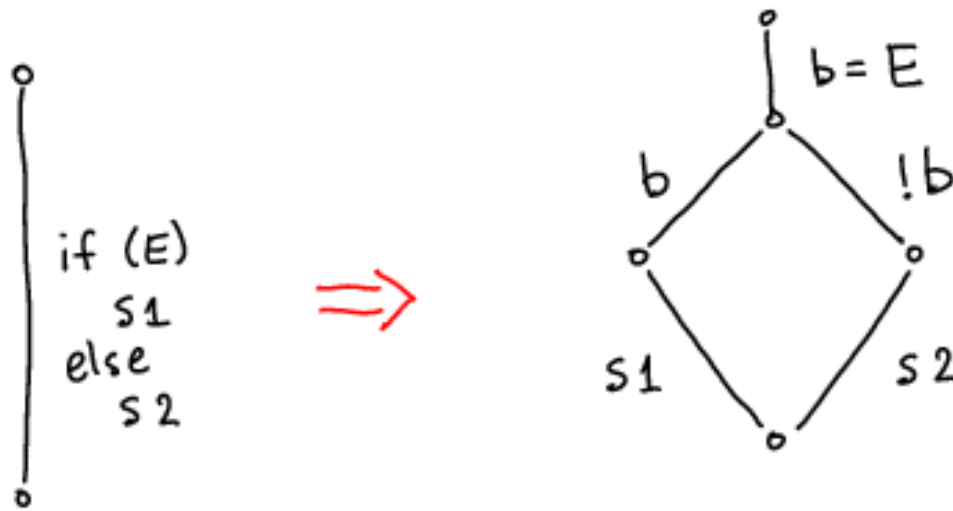
Flattening Expressions

for simplicity and ordering of side effects

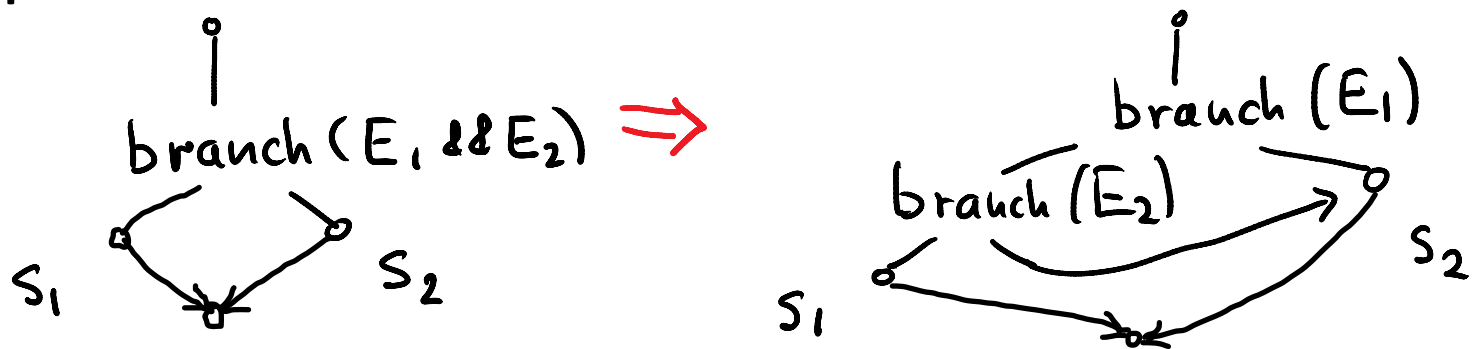
E_1, E_2 - complex expressions
 t_1, t_2 - fresh variables



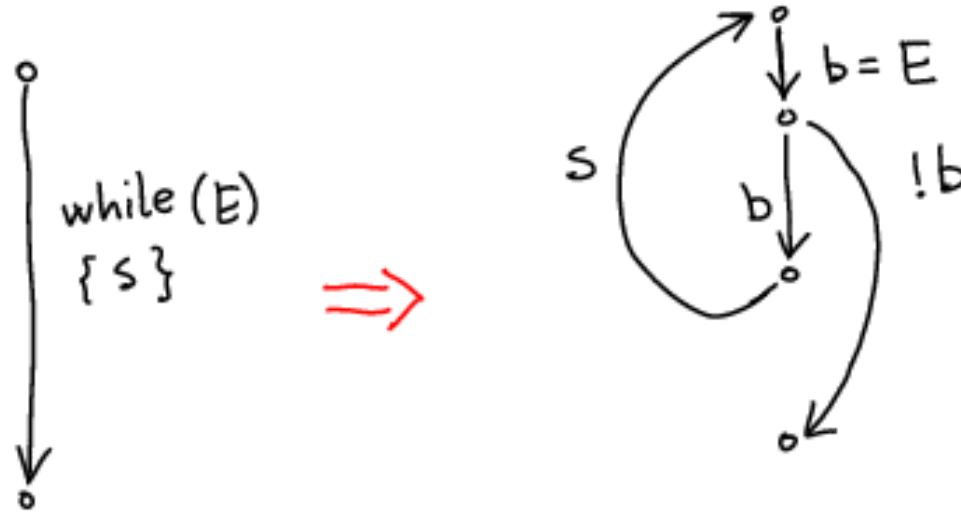
If-Then-Else



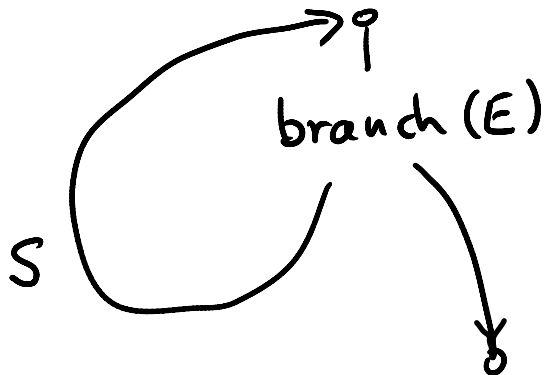
Better translation uses the "branch" instruction approach: have two destinations



While



Better translation uses the "branch" instruction

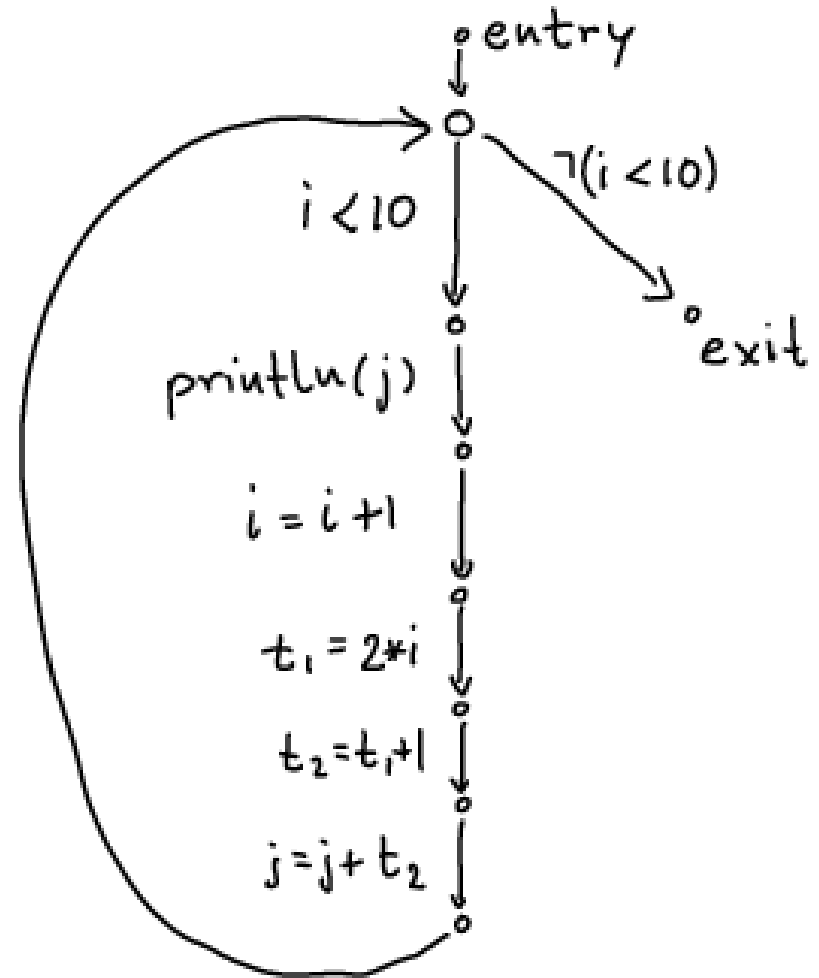


Example 1: Convert to CFG

```
while (i < 10) {  
    println(j);  
    i = i + 1;  
    j = j + 2*i + 1;  
}
```

Example 1 Result

```
while (i < 10) {  
  println(j);  
  i = i + 1;  
  j = j + 2*i + 1;  
}
```

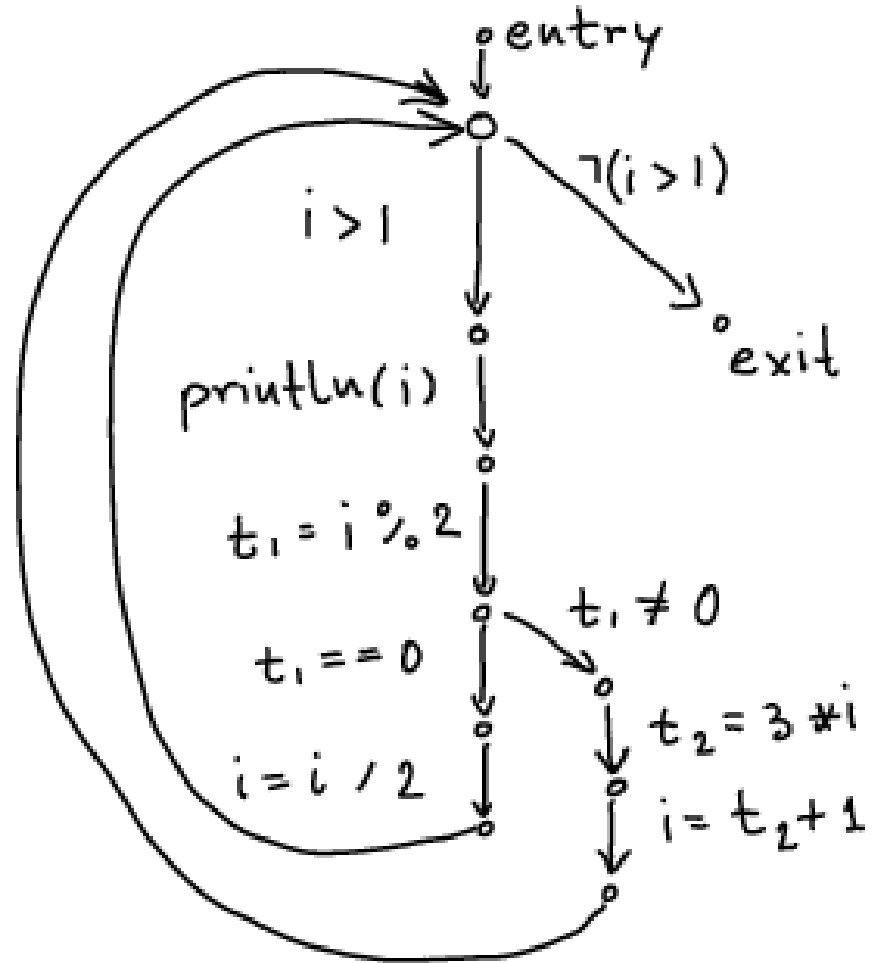


Example 2: Convert to CFG

```
int i = n;  
while (i > 1) {  
    println(i);  
    if (i % 2 == 0) {  
        i = i / 2;  
    } else {  
        i = 3*i + 1;  
    }  
}
```

Example 2 Result

```
int i = n;  
while (i > 1) {  
    println(i);  
    if (i % 2 == 0) {  
        i = i / 2;  
    } else {  
        i = 3*i + 1;  
    }  
}
```



Translation Functions

$$\begin{aligned} [s_1 ; s_2] v_{\text{source}} v_{\text{target}} &= \\ [s_1] v_{\text{source}} v_{\text{fresh}} & \\ [s_2] v_{\text{fresh}} v_{\text{target}} & \end{aligned}$$

$$\begin{aligned} \text{insert}(v_s, \text{stmt}, v_t) &= \\ \text{cfg} = \text{cfg} + (v_s, \text{stmt}, v_t) & \end{aligned}$$

$$\begin{aligned} [\text{branch}(x < y)] v_{\text{source}} v_{\text{true}} v_{\text{false}} &= \\ \text{insert}(v_{\text{source}}, [x < y], v_{\text{true}}); & \\ \text{insert}(v_{\text{source}}, [!(x < y)], v_{\text{false}}) & \end{aligned}$$

$$[x = y + z] v_s v_t = \text{insert}(v_s, x = y + z, v_t)$$

when y, z are constants or variables

Analysis Domain (D)

Lattices

Abstract Interpretation

Generalizes Type Inference

Type Inference

- computes types
- type rules
 - can be used to compute types of expression from subtypes
- types fixed for a variable

Abstract Interpretation

- computes **facts** from a domain
 - types
 - intervals
 - formulas
 - set of initialized variables
 - set of live variables
- transfer functions
 - compute facts for one program point from facts at previous program points
- facts change as the values of vars change (*flow-sensitivity*)

scalac computes types. Try in REPL:

```
class C
```

```
class D extends C
```

```
class E extends C
```

```
val p = false
```

```
val d = new D()
```

```
val e = new E()
```

```
val z = if (p) d else e
```

```
val u = if (p) (d,e) else (d,d)
```

```
val v = if (p) (d,e) else (e,d)
```

```
val f1 = if (p) ((d1:D) => 5) else ((e1:E) => 5)
```

```
val f2 = if (p) ((d1:D) => d) else ((e1:E) => e)
```

Finds "Best Type" for Expression

```
class C
```

```
class D extends C
```

```
class E extends C
```

```
val p = false
```

```
val d = new D() // d:D
```

```
val e = new E() // e:E
```

```
val z = if (p) d else e // z:C
```

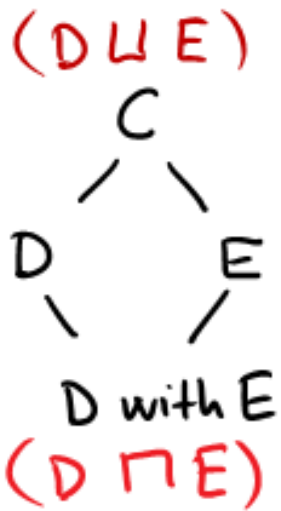
```
val u = if (p) (d,e) else (d,d) // u:(D,C)
```

```
val v = if (p) (d,e) else (e,d) // v:(C,C)
```

```
val f1 = if (p) ((d1:D) => 5) else ((e1:E) => 5) // f1: ((D with E) => Int)
```

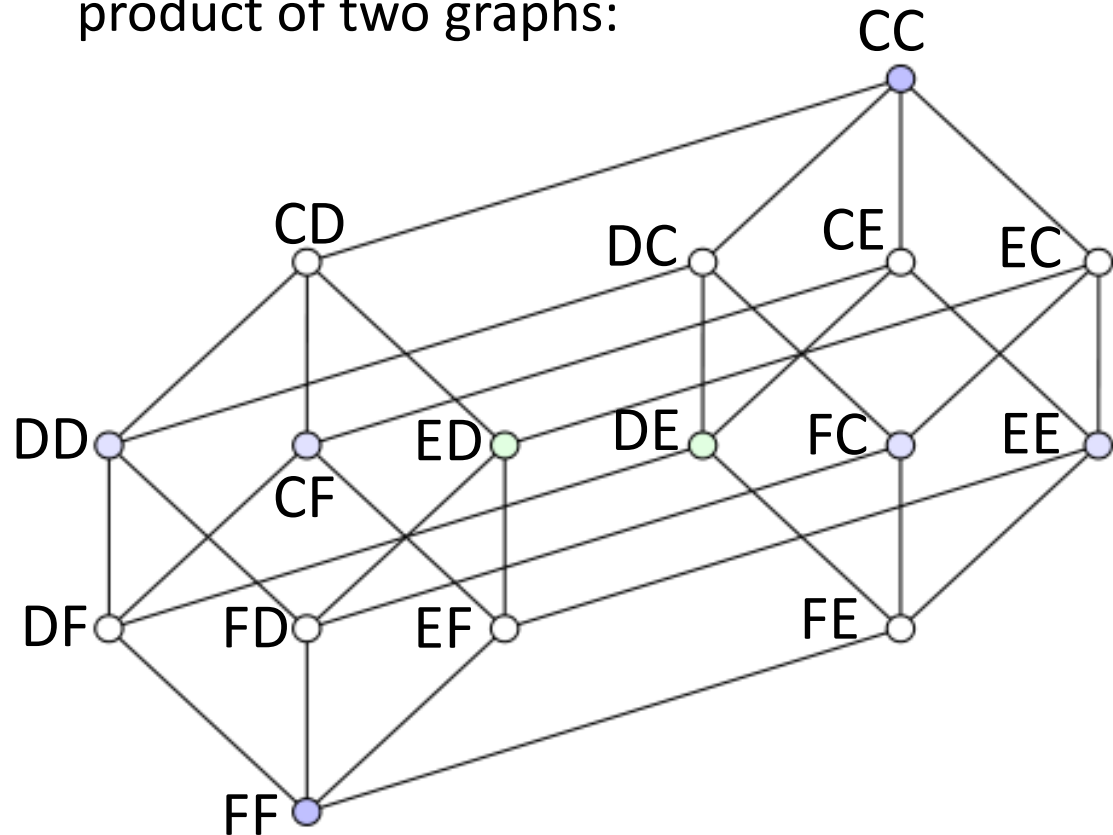
```
val f2 = if (p) ((d1:D) => d) else ((e1:E) => e) // f2: ((D with E) => C)
```

Subtyping Relation in this Example



class C
class D extends C
class E extends C

product of two graphs:

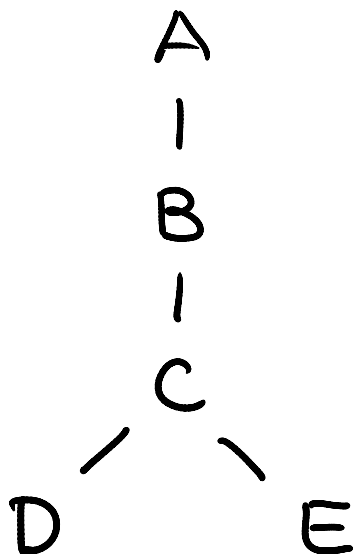


each relation can be visualized in 2D

– two relations: naturally shown in 4D (hypercube)

we usually draw larger elements higher

Least Upper Bound (lub, join)



A,B,C are all upper bounds on both D and E (they are above each of them in the picture, they are supertypes of D and supertypes of E). Among these upper bounds, C is the least one (the most specific one).

We therefore say C is the **least upper bound**,

$$C = D \sqcup E$$

In any partial order \leq , if S is a set of elements (e.g. $S=\{D,E\}$) then:

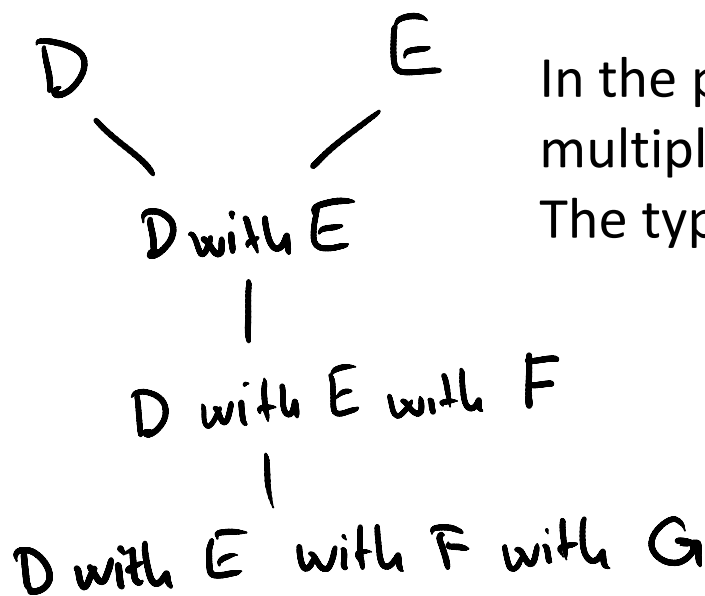
U is **upper bound** on S iff $x \leq U$ for every x in S .

U_0 is the **least upper bound (lub)** of S , written $U_0 = \bigsqcup S$, or $U_0 = \text{lub}(S)$ iff:

U_0 is upper bound and

if U is any upper bound on S , then $U_0 \leq U$

Greatest Lower Bound (glb, meet)



In the presence of traits or interfaces, there are multiple types that are subtypes of both D and E. The type (D with E) is the largest of them.

$$D \sqcap E$$

In any partial order \leq , if S is a set of elements (e.g. $S=\{D,E\}$) then:

L is **lower bound** on S iff $L \leq x$ for every x in S.

L_0 is the **greatest upper bound (glb)** of S, written $L_0 = \bigsqcup S$, or $L_0 = \text{glb}(S)$, iff:

m_0 is upper bound and

if m is any upper bound on S, then $m_0 \leq m$

Computing lub and glb for tuple and function types

$$(x_1, y_1) \sqcup (x_2, y_2) = (x_1 \sqcup x_2, y_1 \sqcup y_2)$$

$$(x_1, y_1) \sqcap (x_2, y_2) = (x_1 \sqcap x_2, y_1 \sqcap y_2)$$

$$(x_1 \rightarrow y_1) \sqcup (x_2 \rightarrow y_2) = (x_1 \sqcap y_1) \rightarrow (y_1 \sqcup y_2)$$

$$(x_1 \rightarrow y_1) \sqcap (x_2 \rightarrow y_2) = (x_1 \sqcup y_1) \rightarrow (y_1 \sqcap y_2)$$

Lattice

Partial order: binary relation \leq (subset of some D^2) which is

- reflexive: $x \leq x$
- anti-symmetric: $x \leq y \wedge y \leq x \rightarrow x = y$
- transitive: $x \leq y \wedge y \leq z \rightarrow x \leq z$

Lattice is a partial order in which every **two-element** set has **lub** and **glb**

- Lemma: if (D, \leq) is lattice and D is finite, then lub and glb exist for every finite set

Idea of Why Lemma Holds

- $\text{lub}(x_1, \text{lub}(x_2, \dots, \text{lub}(x_{n-1}, x_n)))$ is $\text{lub}(\{x_1, \dots, x_n\})$
- $\text{glb}(x_1, \text{glb}(x_2, \dots, \text{glb}(x_{n-1}, x_n)))$ is $\text{glb}(\{x_1, \dots, x_n\})$
- lub of all elements in D is maximum of D
 - by definition, $\text{glb}(\{x\})$ is the maximum of D
- glb of all elements in D is minimum of D
 - by definition, $\text{lub}(\{x\})$ is the minimum of D

Graphs and Partial Orders

- If the domain is finite, then partial order can be represented by directed graphs
 - if $x \leq y$ then draw edge from x to y
- For partial order, no need to draw $x \leq z$ if $x \leq y$ and $y \leq z$. So we only draw non-transitive edges
- Also, because always $x \leq x$, we do not draw those self loops
- Note that the resulting graph is acyclic: if we had a cycle, the elements must to be equal

Defining Abstract Interpretation

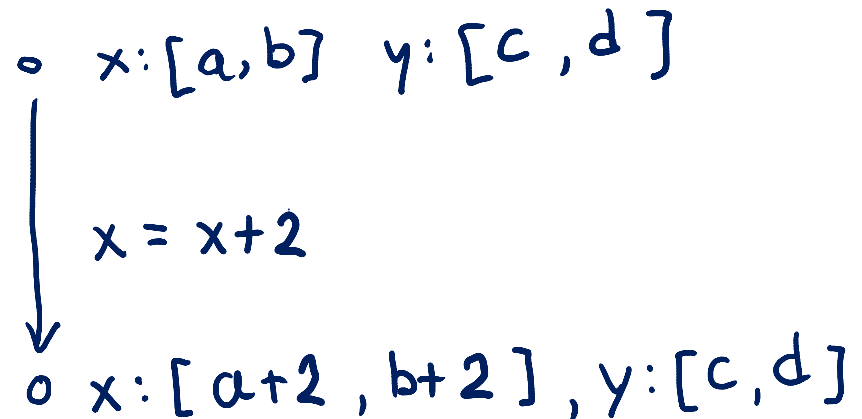
Abstract Domain D describing which information to compute – this is often a lattice

- inferred types for each variable: $x:T1, y:T2$
- interval for each variable $x:[a,b], y:[a',b']$

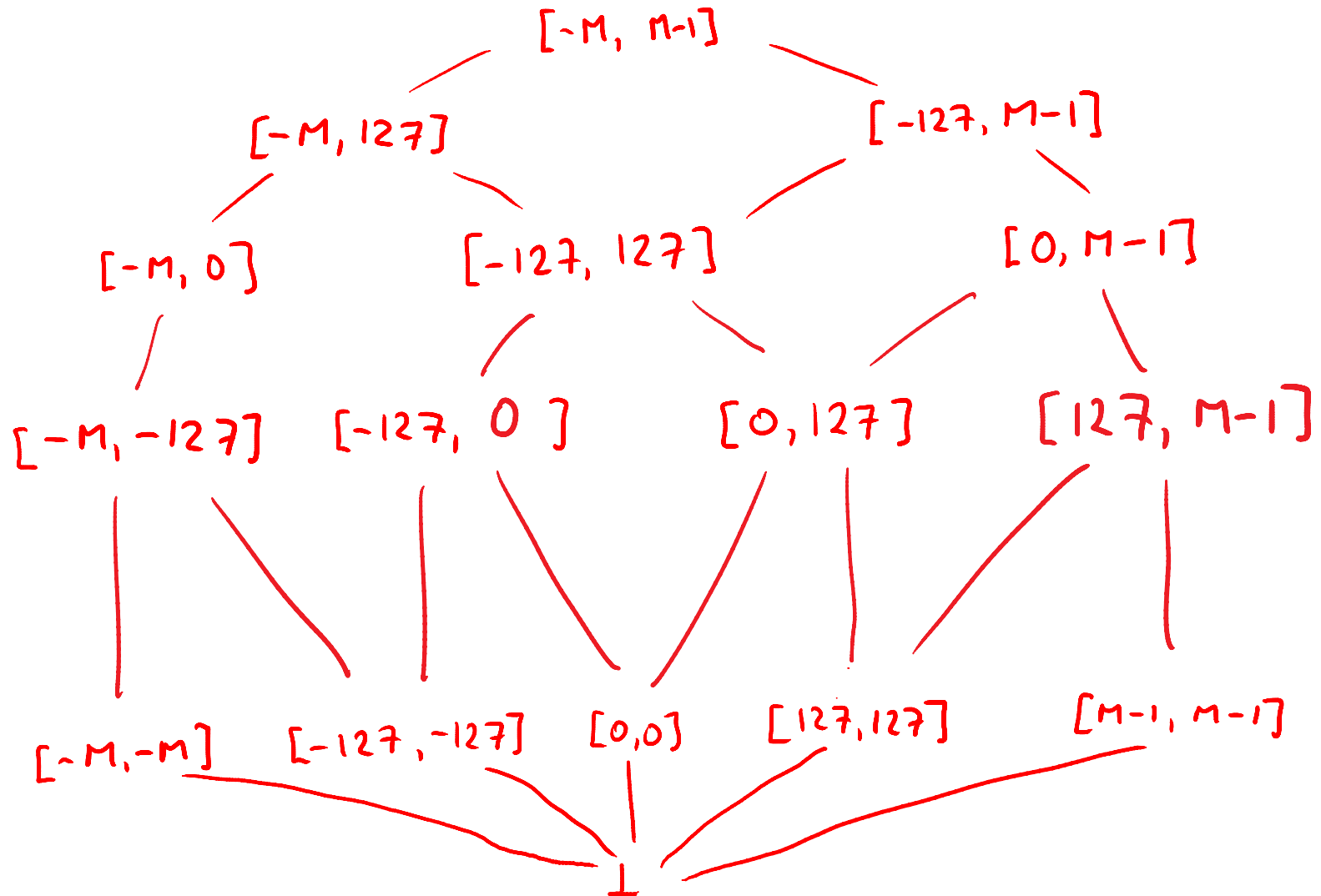
Transfer Functions, $[[st]]$ for each statement **st**, how this statement affects the facts

– Example:

$$\begin{aligned} &[[x = x + 2]](x:[a,b], \dots) \\ &= (x:[a+2, b+2], \dots) \end{aligned}$$



Domain of Intervals $[a,b]$ where $a,b \in \{-M, -127, 0, 127, M-1\}$



For now, we consider arbitrary integer bounds for intervals

- Really 'Int' should be BigInt, as in Haskell, Go
- Often we must analyze machine integers
 - need to correctly represent (and/or warn about) overflows and underflows
 - fundamentally same approach as for unbounded integers
- For efficiency, many analysis do not consider arbitrary intervals, but only a subset of them
- For now, we consider as the domain
 - empty set (denoted \perp , pronounced “bottom”)
 - all intervals $[a,b]$ where a,b are integers and $a \leq b$, or where we allow $a = -\infty$ and/or $b = \infty$
 - set of all integers $[-\infty, \infty]$ is denoted T , pronounced “top”

Find Transfer Function: Plus

Suppose we have only two integer variables: x, y

◦ $x: [a, b] \quad y: [c, d]$
↓
◦ $x: [a', b'] \quad y: [c', d']$

$x = x + y$

If $a \leq x \leq b \quad c \leq y \leq d$

and we execute $x = x + y$

then $x' = x + y$
 $y' = y$

so

$a + c \leq x' \leq$

$b + d$
 $c \leq y' \leq d$

So we can let

$$a' = a + c \quad b' = b + d$$

$$c' = c \quad d' = d$$

Find Transfer Function: Minus

Suppose we have only two integer variables: x, y

$$\begin{array}{l} \bullet \quad x: [a, b] \quad y: [c, d] \\ \downarrow \\ \circ \quad x: [a', b'] \quad y: [c', d'] \end{array}$$

$y = x - y$

If

and we execute $y = x - y$

then

So we can let

$$\begin{array}{ll} a' = a & b' = b \\ c' = a - d & d' = b - c \end{array}$$

Transfer Functions for Tests

Tests e.g. $[x > 1]$ come from translating if, while into CFG

$x: [-10, 10]$

```
if (x > 1) {
```

$x:$

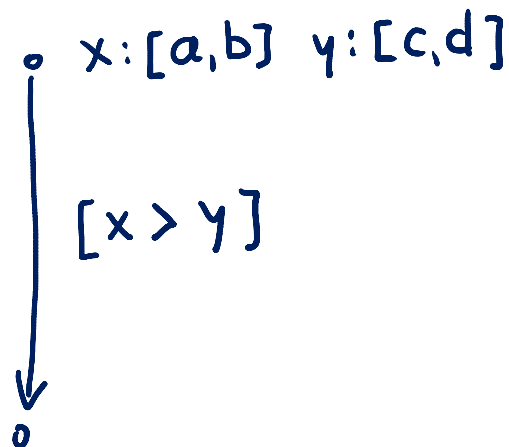
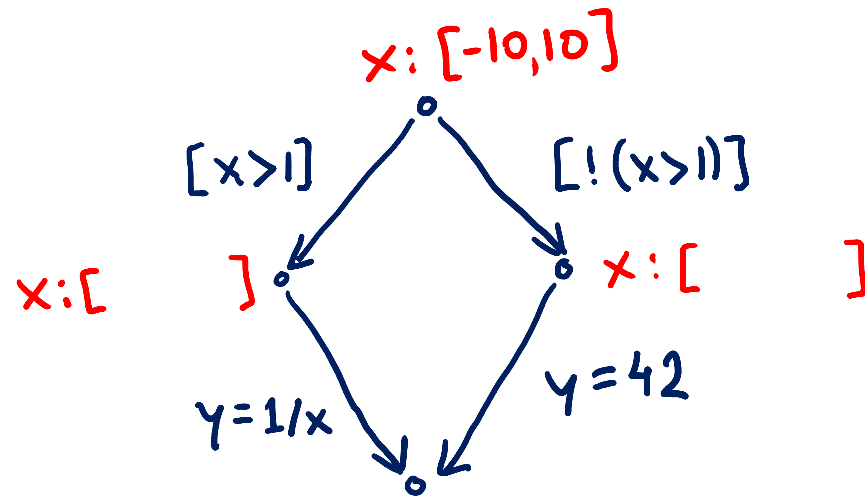
```
  y = 1 / x
```

```
} else {
```

$x:$

```
  y = 42
```

```
}
```



Joining Data-Flow Facts

$x: [-10, 10]$ $y: [-1000, 1000]$

if ($x > 0$) {

$x:$ $y:$

$y = x + 100$

$x:$ $y:$

} else {

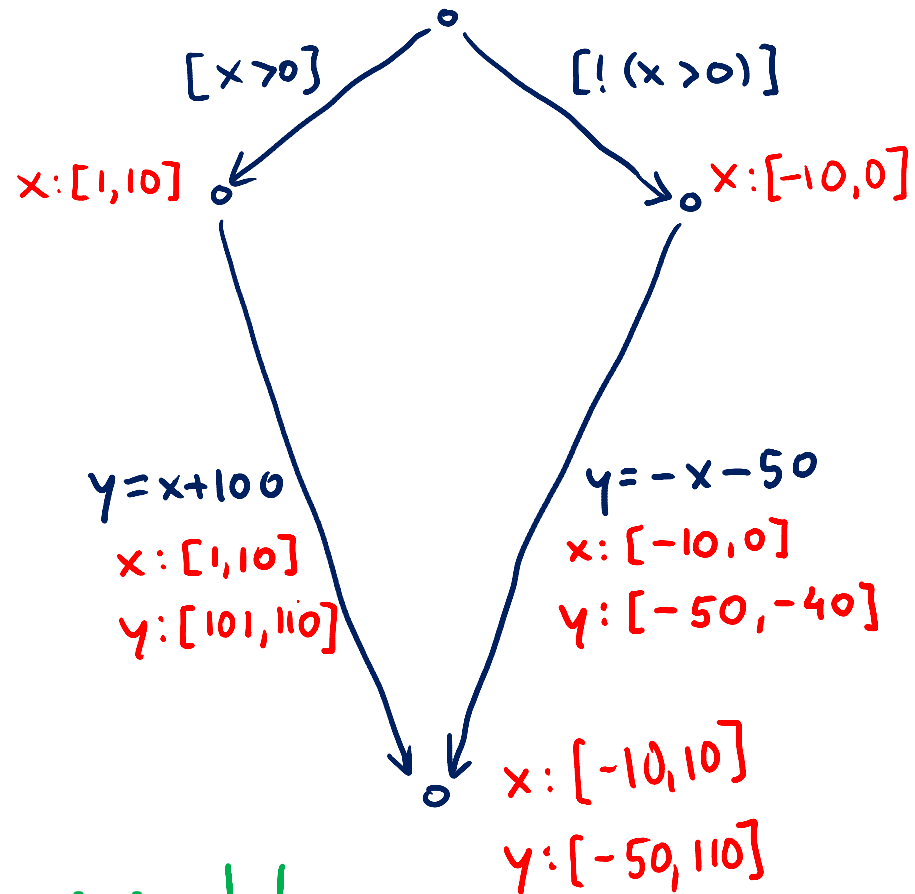
$x:$ $y:$

$y = -x - 50$

$x:$ $y:$

}

$x:$ $y:$



join \sqcup

$$[a, b] \sqcup [c, d] = [\min(a, c), \max(b, d)]$$

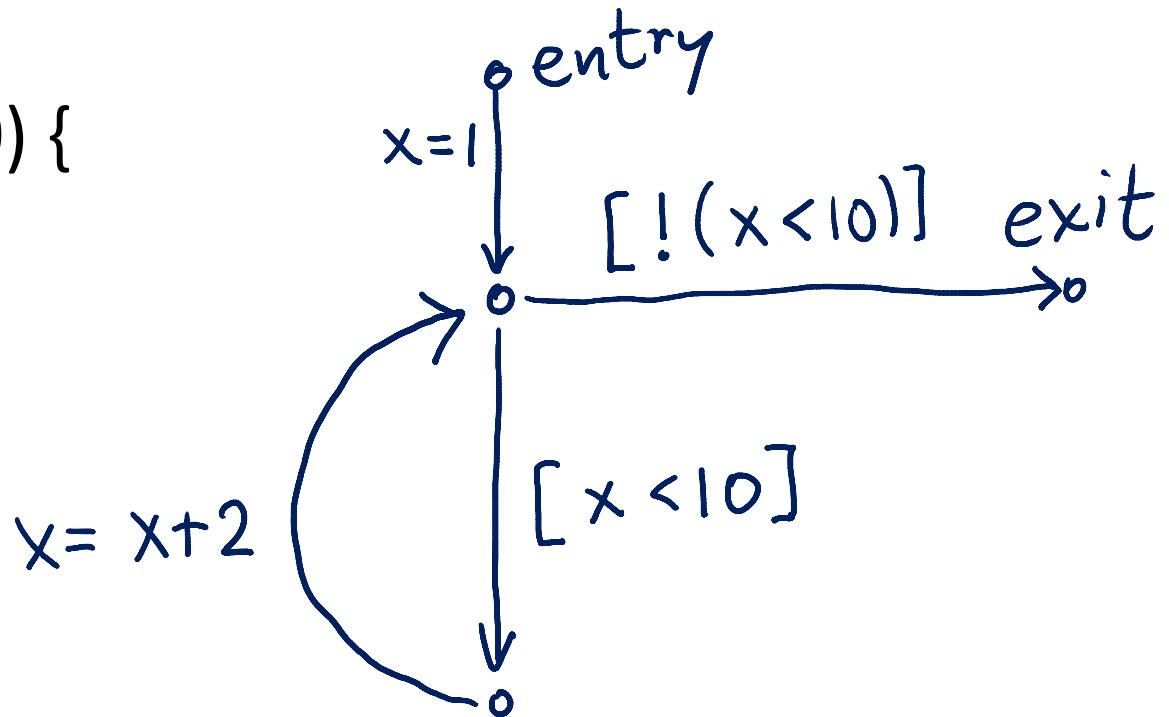
Handling Loops: Iterate Until Stabilizes

$x = 1$

while ($x < 10$) {

$x = x + 2$

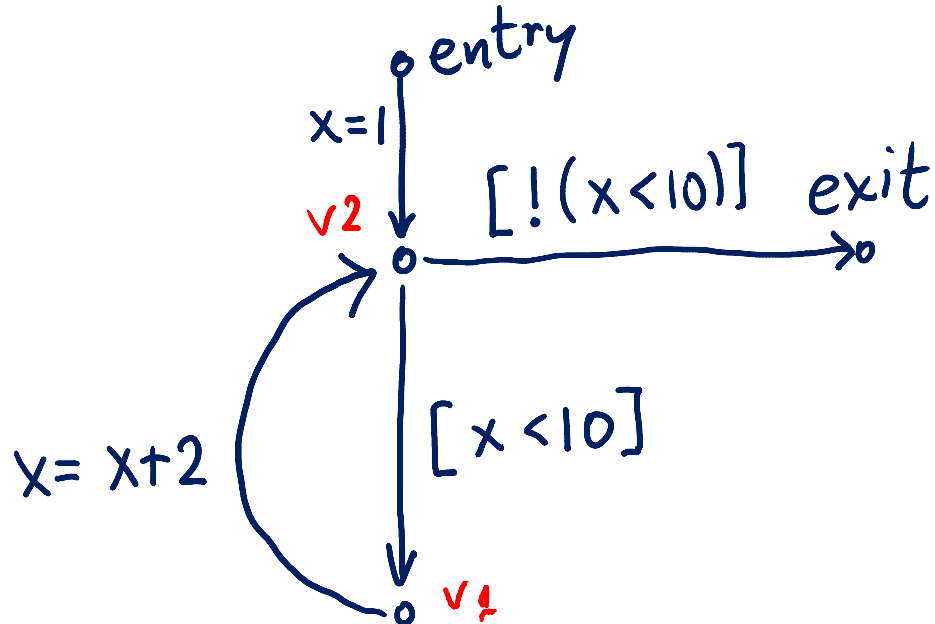
}



Analysis Algorithm

```
var facts : Map[Node,Domain] = Map.withDefault(empty)
facts(entry) = initialValues
while (there was change)
  pick edge (v1,statmt,v2) from CFG
    such that facts(v1) has changed
  facts(v2)=facts(v2) join transferFun(statmt, facts(v1))
}
```

Order does not matter for the end result, as long as we do not permanently neglect any edge whose source was changed.



```

var facts : Map[Node,Domain] = Map.withDefault(empty)
var worklist : Queue[Node] = empty
  def assign(v1:Node,d:Domain) = if (facts(v1)!=d) {
    facts(v1)=d
    for (stmt,v2) <- outEdges(v1) { worklist.add(v2) }
  }
assign(entry, initialValues)
while (!worklist.isEmpty) {
  var v2 = worklist.getAndRemoveFirst
  update = facts(v2)
  for (v1,stmt) <- inEdges(v2)
    { update = update join transferFun(facts(v1),stmt) }
  assign(v2, update)
}

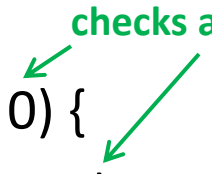
```

Work List Version

Run range analysis, prove **error** is unreachable

```
int M = 16;
int[M] a;
x := 0;
while (x < 10) {
  x := x + 3;
}
if (x >= 0) {
  if (x <= 15)
    a[x]=7;
  else
    error;
} else {
  error;
}
```

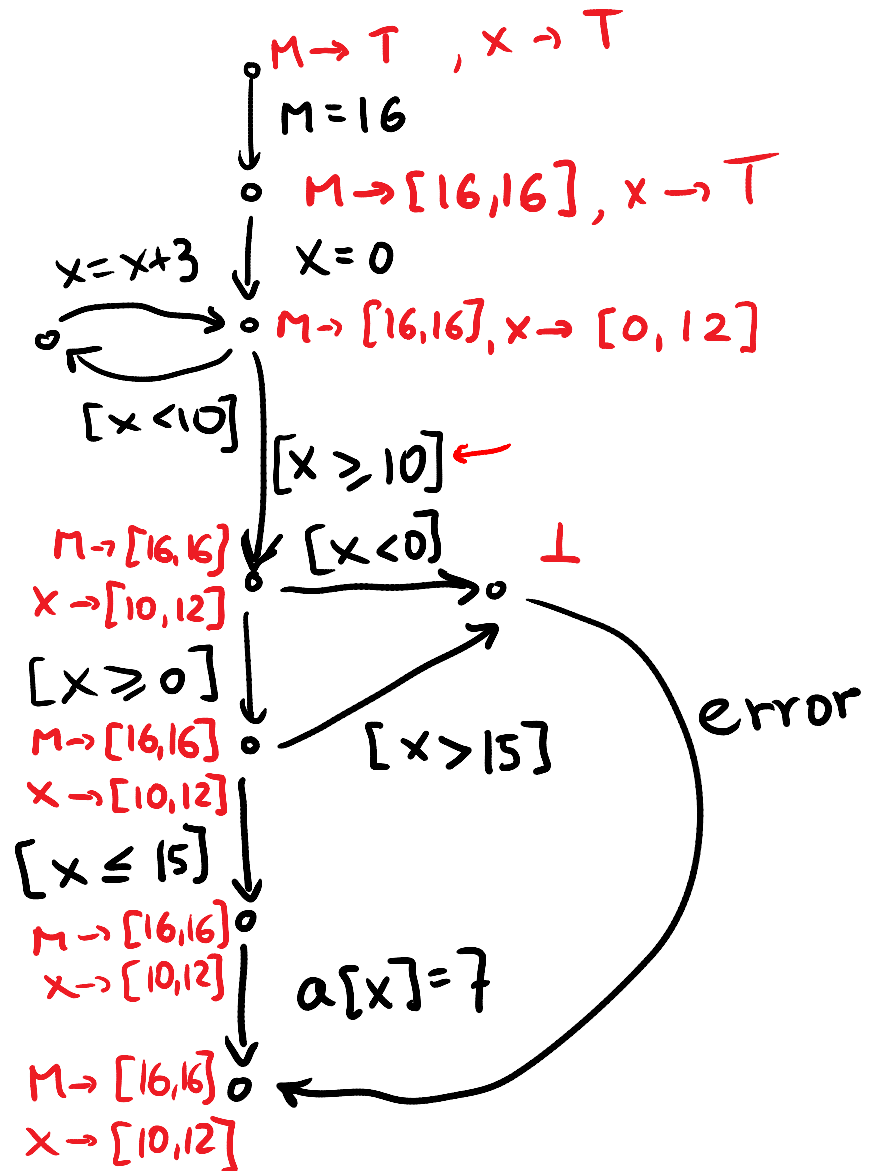
checks array accesses



Range analysis results

```
int M = 16;
int[M] a;
x := 0;
while (x < 10) {
  x := x + 3;
}
if (x >= 0) {
  if (x <= 15)
    a[x]=7;
  else
    error;
} else {
  error;
}
```

checks array accesses



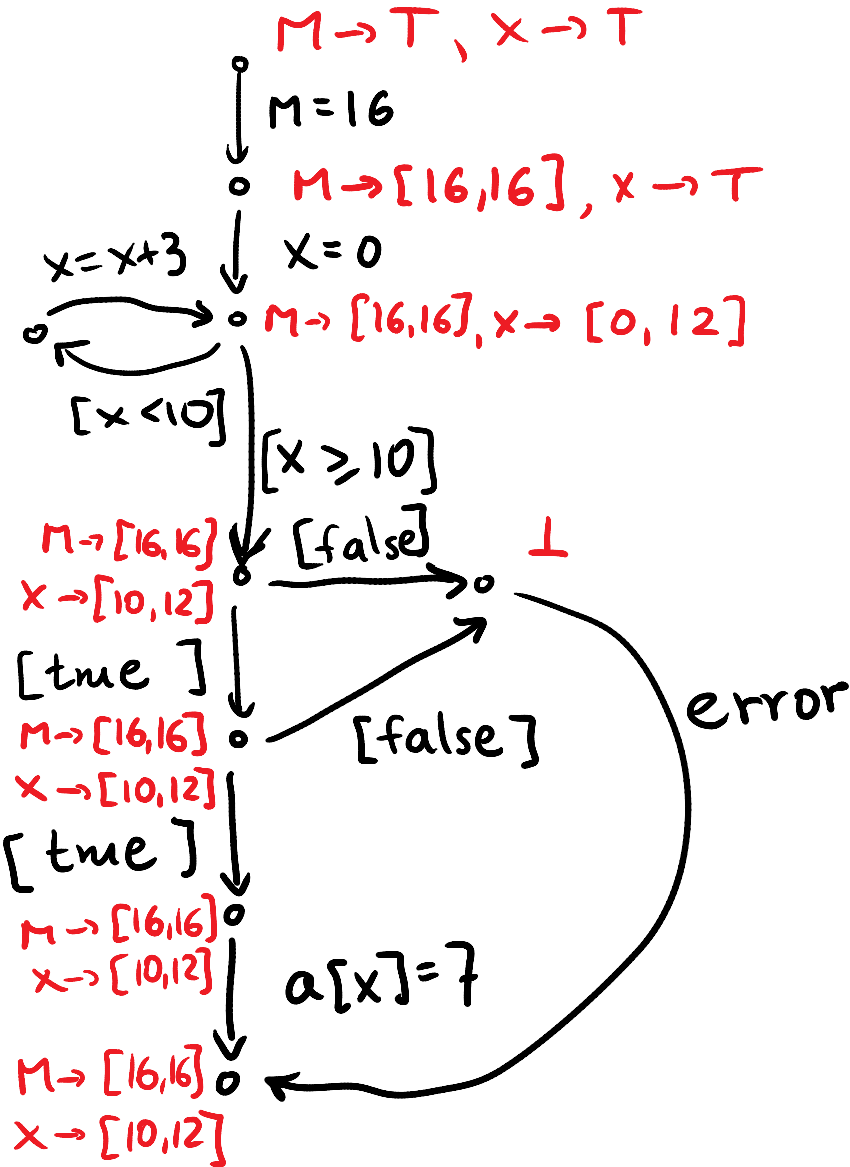
Simplified Conditions

```

int M = 16;
int[M] a;
x := 0;
while (x < 10) {
  x := x + 3;
}
if (x >= 0) {
  if (x <= 15)
    a[x]=7;
  else
    error;
} else {
  error;
}
    
```

checks array accesses

$M \rightarrow [16, 16]$
 $x \rightarrow [0, 9]$

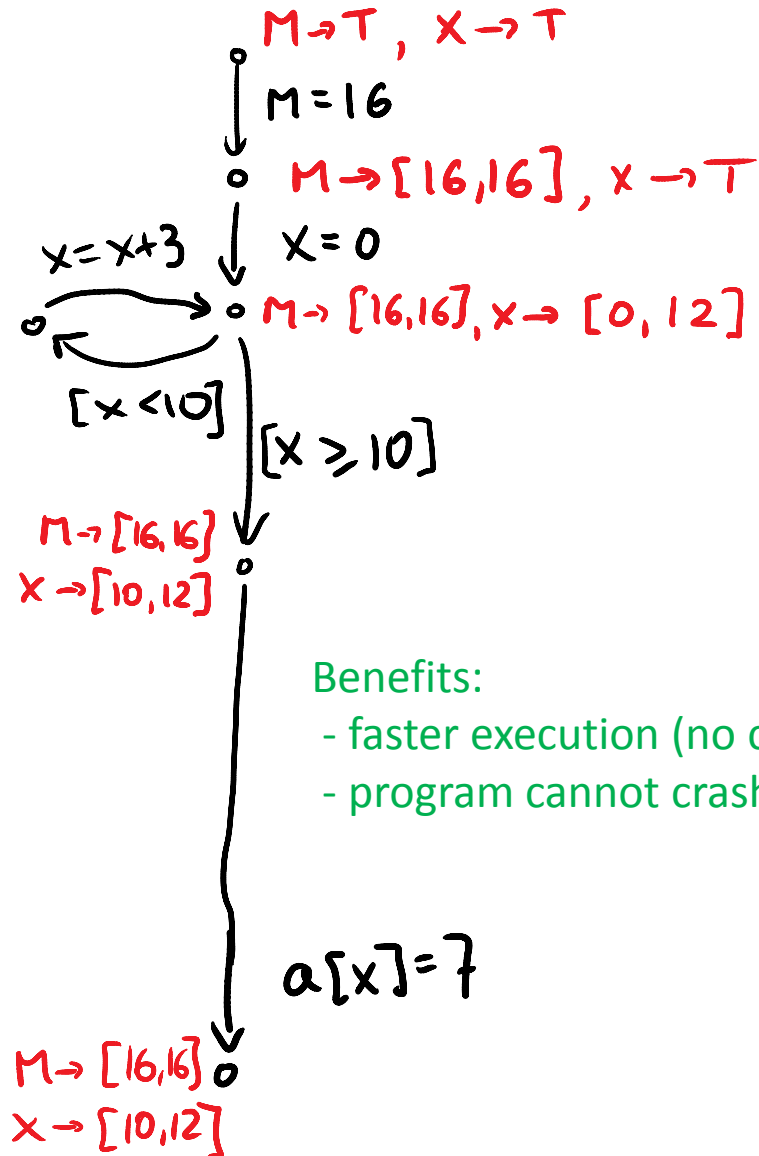


Remove Trivial Edges, Unreachable Nodes

```
int M = 16;
int[M] a;
x := 0;
while (x < 10) {
  x := x + 3;
}
if (x >= 0) {
  if (x <= 15)
    a[x]=7;
  else
    error;
} else {
  error;
}
```

checks array accesses

$M \rightarrow [16, 16]$
 $x \rightarrow [0, 9]$



Benefits:

- faster execution (no checks)
- program cannot crash with error

Apply Range Analysis and Simplify

```
int a, b, step, i;
boolean c;
a = 0;
b = a + 10;
step = -1;
if (step > 0) {
    i = a;
} else {
    i = b;
}
c = true;
while (c) {
    process(i);
    i = i + step;
    if (step > 0) {
        c = (i < b);
    } else {
        c = (i > a);
    }
}
```

For booleans, use this lattice: $D_b = \{ \{\}, \{\underline{\text{false}}\}, \{\underline{\text{true}}\}, \{\underline{\text{false}}, \underline{\text{true}}\} \}$
with ordering given by set subset relation.