

# Lecture 36: Code Optimization

[Adapted from notes by R. Bodik and G. Necula]

# Introduction to Code Optimization

*Code optimization* is the usual term, but is grossly misnamed, since code produced by “optimizers” is not optimal in any reasonable sense. *Program improvement* would be more appropriate.

Topics:

- Basic blocks
- Control-flow graphs (CFGs)
- Algebraic simplification
- Constant folding
- Static single-assignment form (SSA)
- Common-subexpression elimination (CSE)
- Copy propagation
- Dead-code elimination
- Peephole optimizations

# Basic Blocks

- A *basic block* is a maximal sequence of instructions with:
  - no labels (except at the first instruction), and
  - no jumps (except in the last instruction)
- Idea:
  - Cannot jump into a basic block, except at the beginning.
  - Cannot jump within a basic block, except at end.
  - Therefore, each instruction in a basic block is executed after all the preceding instructions have been executed

# Basic-Block Example

- Consider the basic block

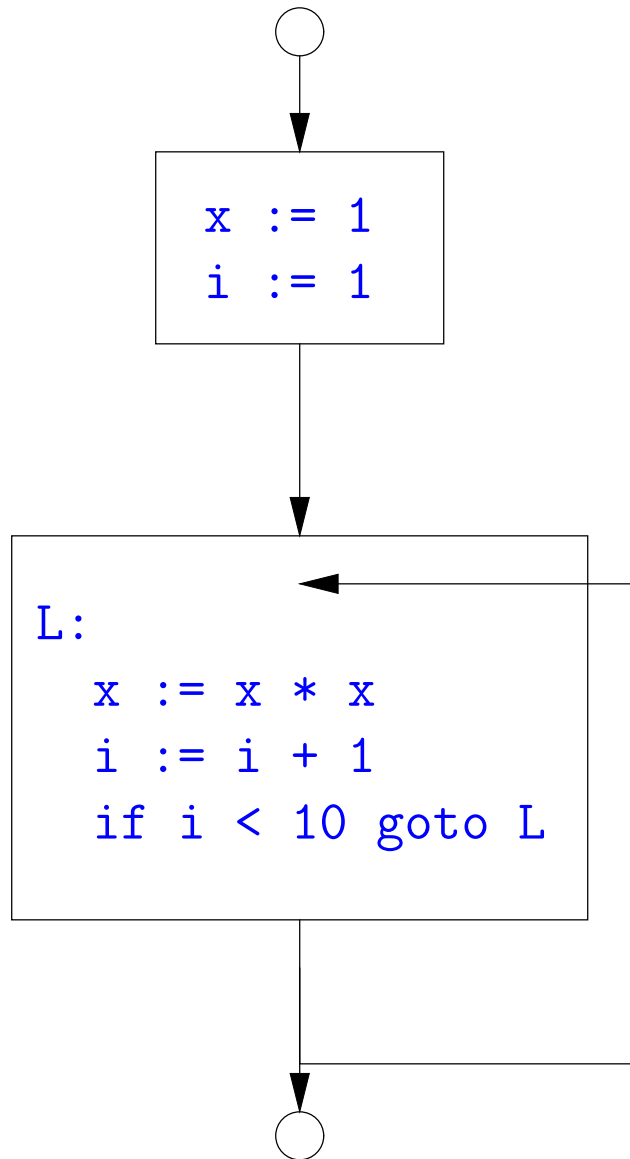
```
1. L1:  
2. t := 2 * x  
3. w := t + x  
4. if w > 0 goto L2
```

- No way for (3) to be executed without (2) having been executed right before
- We can change (3) to  $w := 3 * x$
- Can we eliminate (2) as well?

# Control-Flow Graphs (CFGs)

- A control-flow graph is a directed graph with basic blocks as nodes
- There is an edge from block  $A$  to block  $B$  if the execution can flow from the last instruction in  $A$  to the first instruction in  $B$ :
  - The last instruction in  $A$  can be a jump to the label of  $B$ .
  - Or execution can fall through from the end of block  $A$  to the beginning of block  $B$ .

# Control-Flow Graphs: Example



- The body of a method (or procedure) can be represented as a *CFG*
- There is one initial node
- All "return" nodes are terminal

# Optimization Overview

- Optimization seeks to improve a program's utilization of some resource:
  - Execution time (most often)
  - Code size
  - Network messages sent
  - Battery power used, etc.
- Optimization should not depart from the programming language's semantics
- So if the semantics of a particular program is deterministic, optimization must not change the answer.
- On the other hand, some program behavior is undefined (e.g., what happens when an unchecked rule in *C* is violated), and in those cases, optimization may cause differences in results.

# A Classification of Optimizations

- For languages like C and Java there are three granularities of optimizations
  1. *Local optimizations*: Apply to a basic block in isolation.
  2. *Global optimizations*: Apply to a control-flow graph (single function body) in isolation.
  3. *Inter-procedural optimizations*: Apply across function boundaries.
- Most compilers do (1), many do (2) and very few do (3)
- Problem is expense: (2) and (3) typically require superlinear time. Can usually handle that when limited to a single function, but gets problematic for larger program.
- In practice, generally *don't* implement fanciest known optimizations: some are hard to implement (esp., hard to get right), some require a lot of compilation time.
- The goal: maximum improvement with minimum cost.



# Local Optimizations: Algebraic Simplification

- Some statements can be deleted

`x := x + 0`

`x := x * 1`

- Some statements can be simplified or converted to use faster operations:

Original	Simplified
<code>x := x * 0</code>	<code>x := 0</code>
<code>y := y ** 2</code>	<code>y := y * y</code>
<code>x := x * 8</code>	<code>x := x &lt;&lt; 3</code>
<code>x := x * 15</code>	<code>t := x &lt;&lt; 4; x := t - x</code>

(on some machines << is faster than \*; but not on all!)

# Local Optimization: Constant Folding

- Operations on constants can be computed at compile time.
- Example:  $x := 2 + 2$  becomes  $x := 4$ .
- Example: `if 2 < 0 jump L` becomes a no-op.
- When might constant folding be dangerous?

# Global Optimization: Unreachable code elimination

- Basic blocks that are not reachable from the entry point of the CFG may be eliminated.
- Why would such basic blocks occur?
- Removing unreachable code makes the program smaller (sometimes also faster, due to instruction-cache effects, but this is probably not a terribly large effect.)

# Single Assignment Form

- Some optimizations are simplified if each assignment is to a temporary that has not appeared already in the basic block.
- Intermediate code can be rewritten to be in *(static) single assignment (SSA) form*:

$x := a + y$	$x := a + y$
$a := x$	$a1 := x$
$x := a * x$	$x1 := a1 * x$
$b := x + a$	$b := x1 + a1$

where  $x1$  and  $a1$  are fresh temporaries.

# Common SubExpression (CSE) Elimination in Basic Blocks

- A *common subexpression* is an expression that appears multiple times on a right-hand side in contexts where the operands have the same values in each case (so that the expression will yield the same value).
- Assume that the basic block on the left is in single assignment form.

$x := y + z$

...

$w := y + z$

$x := y + z$

...

$w := x$

- That is, if two assignments have the same right-hand side, we can replace the second instance of that right-hand side with the variable that was assigned the first instance.
- How did we use the assumption of single assignment here?

# Copy Propagation

- If  $w := x$  appears in a block, can replace all subsequent uses of  $w$  with uses of  $x$ .
- Example:

$b := z + y$	$b := z + y$
$a := b$	$a := b$
$x := 2 * a$	$x := 2 * b$

- This does not make the program smaller or faster but might enable other optimizations. For example, if  $a$  is not used after this statement, we need not assign to it.
- Or consider:

$b := 13$	$b := 13$
$x := 2 * b$	$x := 2 * 13$

which immediately enables constant folding.

- Again, the optimization, as described, won't work unless the block is in single assignment form.

## Another Example of Copy Propagation and Constant Folding

a := 5	a := 5	a := 5	a := 5	a := 5
x := 2 * a	x := 2 * 5	x := 10	x := 10	x := 10
y := x + 6	y := x + 6	y := 10 + 6	y := 16	y := 16
t := x * y	t := x * y	t := 10 * y	t := 10 * 16	t := 160

# Dead Code Elimination

- If that statement  $w := rhs$  appears in a basic block and  $w$  does not appear anywhere else in the program, we say that the statement is *dead* and can be eliminated; it does not contribute to the program's result.
- Example: ( $a$  is not used anywhere else)

$b := z + y$	$b := z + y$	$b := z + y$
$a := b$	$a := b$	
$x := 2 * a$	$x := 2 * b$	$x := 2 * b$

- How have I used SSA here?



# Applying Local Optimizations

- As the examples show, each local optimization does very little by itself.
- Typically, optimizations interact: performing one optimization enables others.
- So typical optimizing compilers repeatedly perform optimizations until no improvement is possible, or it is no longer cost effective.

## An Example: Initial Code

```
a := x ** 2
b := 3
c := x
d := c * c
e := b * 2
f := a + d
g := e * f
```

## An Example II: Algebraic simplification

```
a := x ** 2
b := 3
c := x
d := c * c
e := b * 2
f := a + d
g := e * f
```

## An Example II: Algebraic simplification

a := x \* x

b := 3

c := x

d := c \* c

e := b + b

f := a + d

g := e \* f

# An Example: Copy propagation

```
a := x * x
b := 3
c := x
d := c * c
e := b + b
f := a + d
g := e * f
```

# An Example: Copy propagation

```
a := x * x
b := 3
c := x
d := x * x
e := 3 + 3
f := a + d
g := e * f
```

## An Example: Constant folding

```
a := x * x
b := 3
c := x
d := x * x
e := 3 + 3
f := a + d
g := e * f
```

## An Example: Constant folding

```
a := x * x
b := 3
c := x
d := x * x
e := 6
f := a + d
g := e * f
```



# An Example: Common Subexpression Elimination

a := x \* x

b := 3

c := x

d := x \* x

e := 6

f := a + d

g := e \* f

# An Example: Common Subexpression Elimination

a := x \* x

b := 3

c := x

d := a

e := 6

f := a + d

g := e \* f

# An Example: Copy propagation

```
a := x * x
b := 3
c := x
d := a
e := 6
f := a + d
g := e * f
```

# An Example: Copy propagation

```
a := x * x  
b := 3  
c := x  
d := a  
e := 6  
f := a + a  
g := 6 * f
```

## An Example: Dead code elimination

```
a := x * x  
b := 3  
c := x  
d := a  
e := 6  
f := a + a  
g := 6 * f
```

## An Example: Dead code elimination

$a := x * x$

$f := a + a$

$g := 6 * f$

This is the final form.

# Peephole Optimizations on Assembly Code

- The optimizations presented before work on intermediate code.
- *Peephole optimization* is a technique for improving assembly code directly
  - The "*peephole*" is a short subsequence of (usually contiguous) instructions, either contiguous, or linked together by the fact that they operate on certain registers that no intervening instructions modify.
  - The optimizer replaces the sequence with another equivalent, but (one hopes) better one.
  - Write peephole optimizations as replacement rules
$$i1; \dots; in \Rightarrow j1; \dots; jm$$
possibly plus additional constraints. The *j*'s are the improved version of the *i*'s.

## Peephole optimization examples:

- We'll use the notation '@A' for pattern variables.

- Example:

`movl %@a %@b; L: movl %@b %@a ⇒ movl %@a %@b`

assuming `L` is not the target of a jump.

- Example:

`addl $@k1, %@a; movl @k2(%@a), %@b  
⇒ movl @k1+@k2(%@a), %@b`

assuming `%@a` is "dead".

- Example (PDP11):

`mov #@I, @I(@ra) ⇒ mov (r7), @I(@ra)`

This is a real hack: we reuse the value `I` as both the immediate value and the offset from `ra`. On the PDP11, the program counter is `r7`.

- As for local optimizations, peephole optimizations need to be applied repeatedly to get maximum effect.



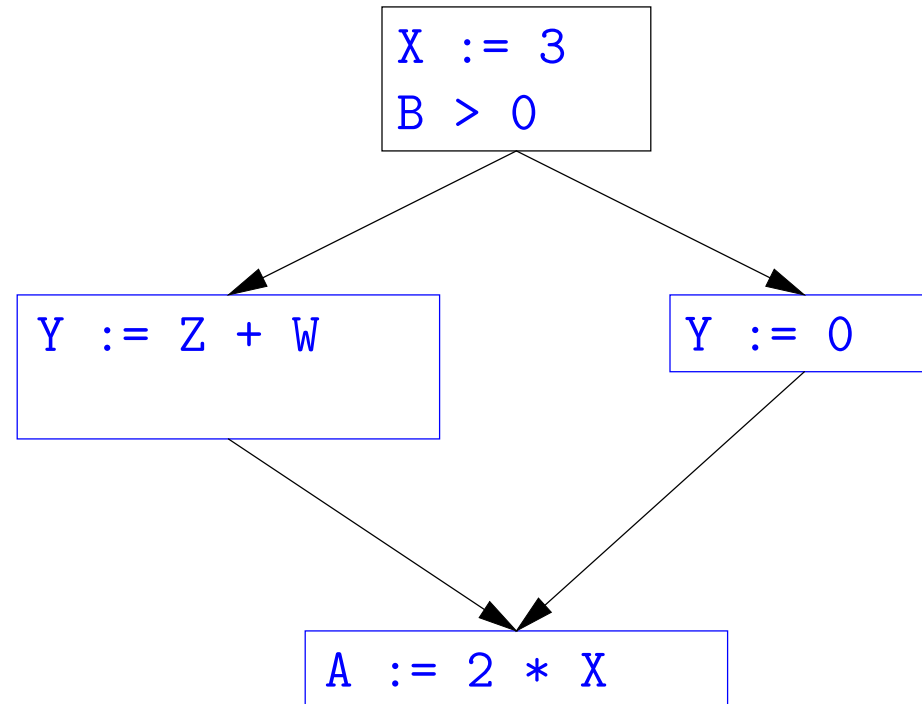
# Problems:

- Serious problem: what to do with pointers? Problem is *aliasing*: two names for the same variable:
  - As a result, *\*t* may change even if local variable *t* does not and we never assign to *\*t*.
  - Affects language design: rules about overlapping parameters in Fortran, and the **restrict** keyword in C.
  - Arrays are a special case (address calculation): is *A[i]* the same as *A[j]*? Sometimes the compiler can tell, depending on what it knows about *i* and *j*.
- What about global variables and calls?
  - Calls are not exactly jumps, because they (almost) always return.
  - Can modify global variables used by caller

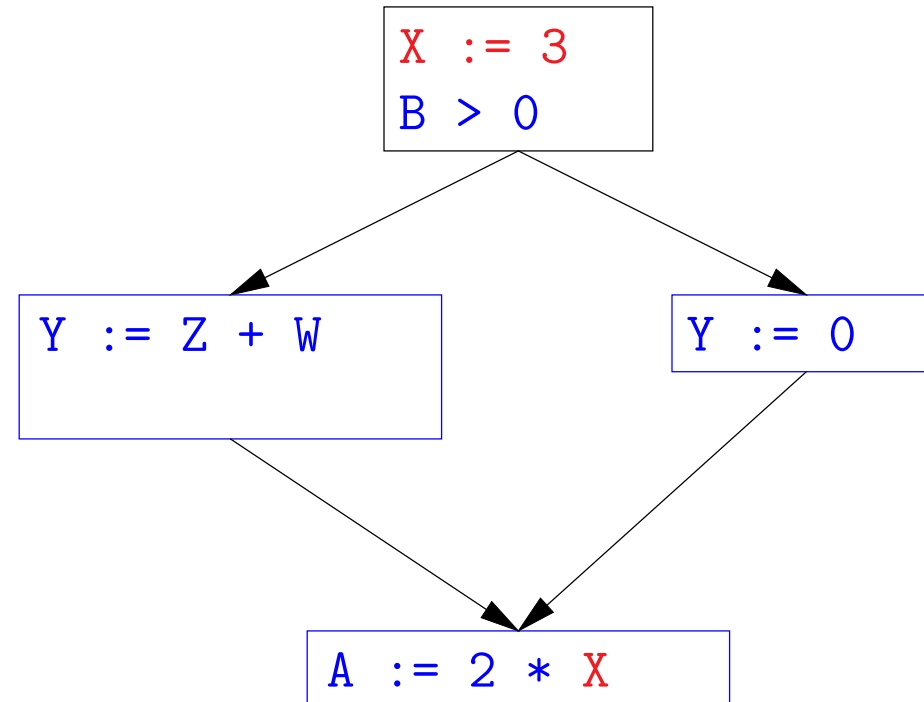
# Global Optimization

- *Global optimization* refers to program optimizations that encompass multiple basic blocks in a function.
- (I have used the term *galactic optimization* to refer to going beyond function boundaries, but it hasn't caught on; we call it just *interprocedural optimization*.)
- Since we can't use the usual assumptions about basic blocks, global optimization requires *global flow analysis* to see where values can come from and get used.
- The overall question is: When can local optimizations (from the last lecture) be applied across multiple basic blocks?

# A Simple Example: Copy Propagation

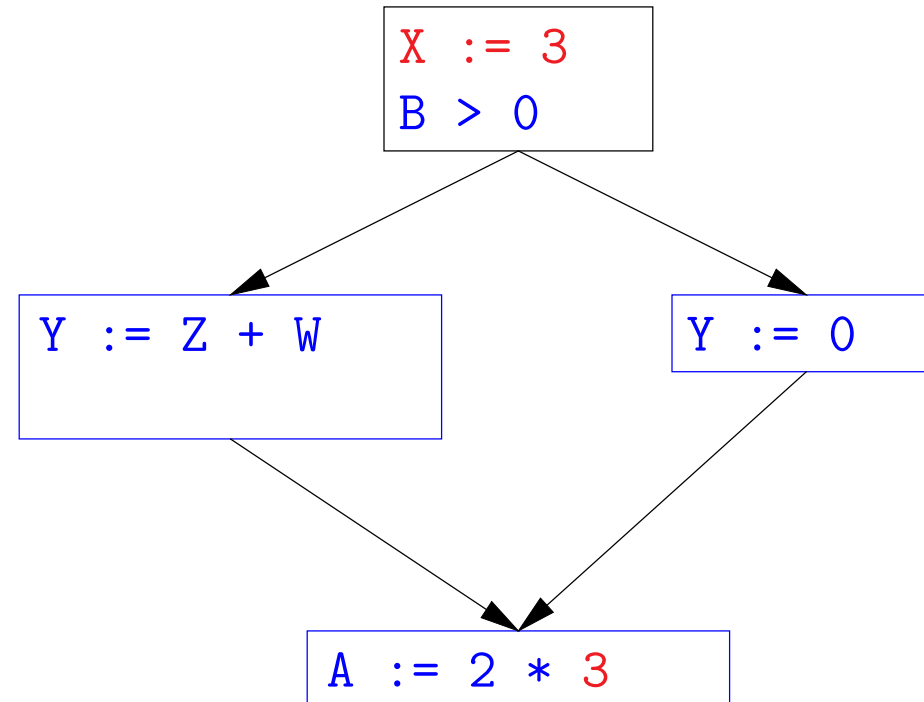


# A Simple Example: Copy Propagation



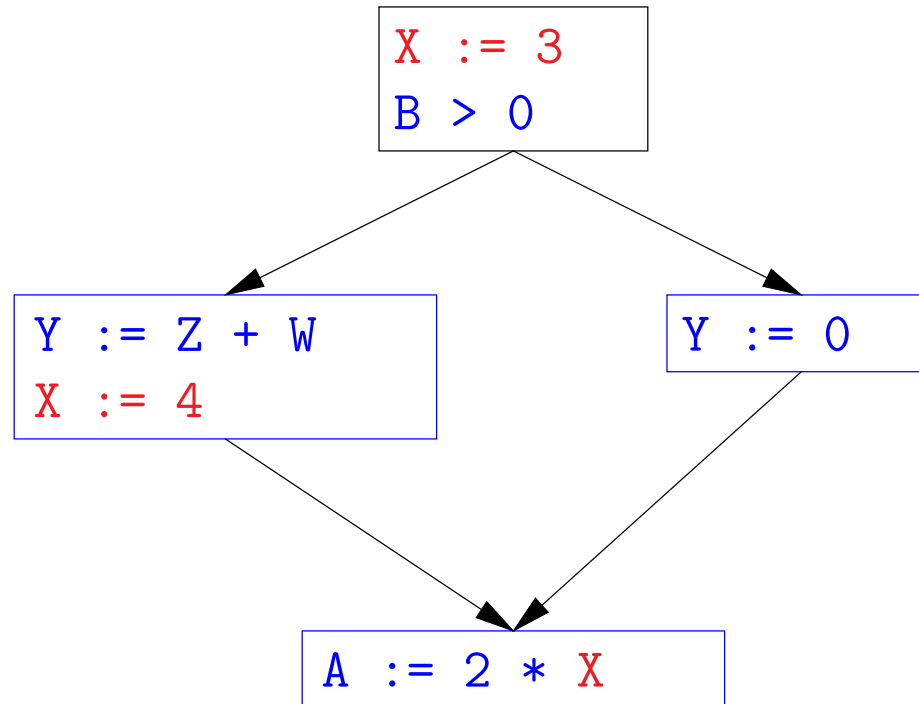
- Without other assignments to `X`, it is valid to treat the red parts as if they were in the same basic block.

# A Simple Example: Copy Propagation



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# A Simple Example: Copy Propagation



- Without other assignments to `X`, it is valid to treat the red parts as if they were in the same basic block.
- But as soon as *one* other block on the path to the bottom block assigns to `X`, we can no longer do so.
- It is correct to apply copy propagation to a variable `x` from an assignment statement `A: x := ...` to a given use of `x` in statement `B` only if the last assignment to `x` in every path from `A` to `B` is `A`.

# Issues

- This correctness condition is not trivial to check
- “All paths” includes paths around loops and through branches of conditionals
- Checking the condition requires global analysis: an analysis of the entire control-flow graph for one method body.
- This is typical for optimizations that depend on some property  $P$  at a particular point in program execution.
- Indeed, property  $P$  is typically undecidable, so program optimization is all about making *conservative* (but not cowardly) approximations of  $P$ .

# Undecidability of Program Properties

- Rice's "theorem:" Most interesting dynamic properties of a program are undecidable. E.g.,

- Does the program halt on all (some) inputs? (Halting Problem)
- Is the result of a function **F** always positive? (Consider

```
def F(x):  
    H(x)  
    return 1
```

Result is positive iff **H** halts.)

- Syntactic properties are typically decidable (e.g., "How many occurrences of **x** are there?").
- Theorem does not apply in absence of loops



# Conservative Program Analyses

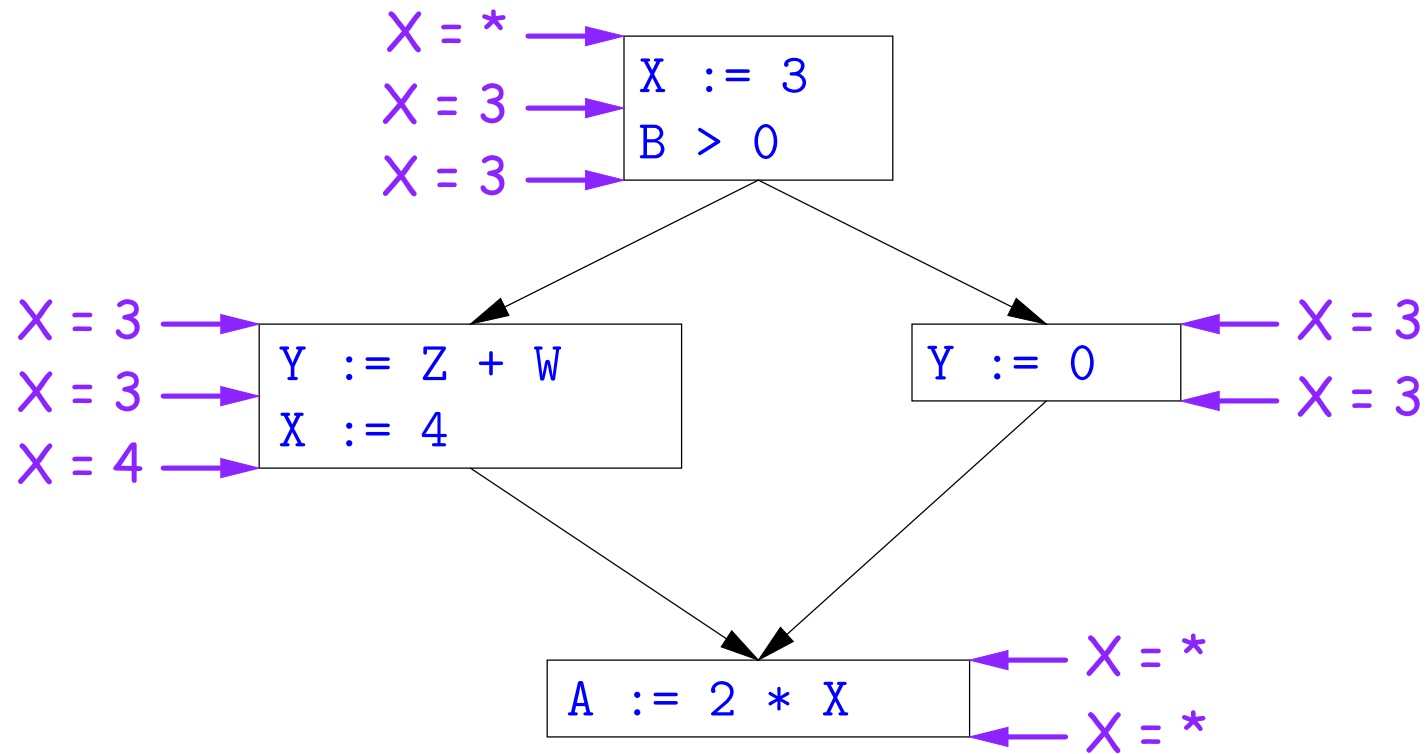
- If a certain optimization requires  $P$  to be true, then
  - If we know that  $P$  is definitely true, we can apply the optimization
  - If we don't know whether  $P$  is true, we simply don't do the optimization. Since optimizations are not supposed to change the meaning of a program, this is safe.
- In other words, in analyzing a program for properties like  $P$ , it is *always correct* (albeit non-optimal) to say "don't know."
- The trick is to say it as seldom as possible.
- *Global dataflow analysis* is a standard technique for solving problems with these characteristics.

## Example: Global Constant Propagation

- *Global constant propagation* is just the restriction of copy propagation to constants.
- In this example, we'll consider doing it for a single variable ( $X$ ).
- At every program point (i.e., before or after any instruction), we associate one of the following values with  $X$

Value	Interpretation
#	(aka <i>bottom</i> ) No value has reached here (yet)
$c$	(For $c$ a constant) $X$ definitely has the value $c$ .
*	(aka <i>top</i> ) Don't know what, if any, constant value $X$ has.

# Example of Result of Constant Propagation



# Using Analysis Results

- Given global constant information, it is easy to perform the optimization:
  - If the point immediately before a statement using  $x$  tells us that  $x = c$ , then replace  $x$  with  $c$ .
  - Otherwise, leave it alone (the conservative option).
- But how do we compute these properties  $x = \dots$ ?

# Transfer Functions

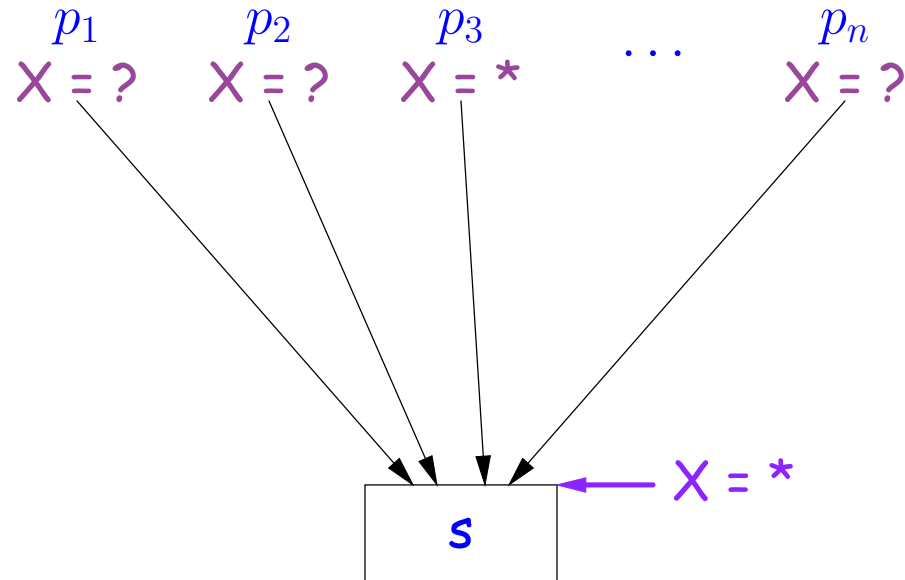
- **Basic Idea:** Express the analysis of a complicated program as a combination of simple rules relating the change in information between adjacent statements
- That is, we “*push*” or *transfer* information from one statement to the next.
- For each statement  $s$ , we end up with information about the value of  $x$  immediately before and after  $s$ :

$Cin(X,s)$  = value of  $x$  before  $s$

$Cout(X,s)$  = value of  $x$  after  $s$

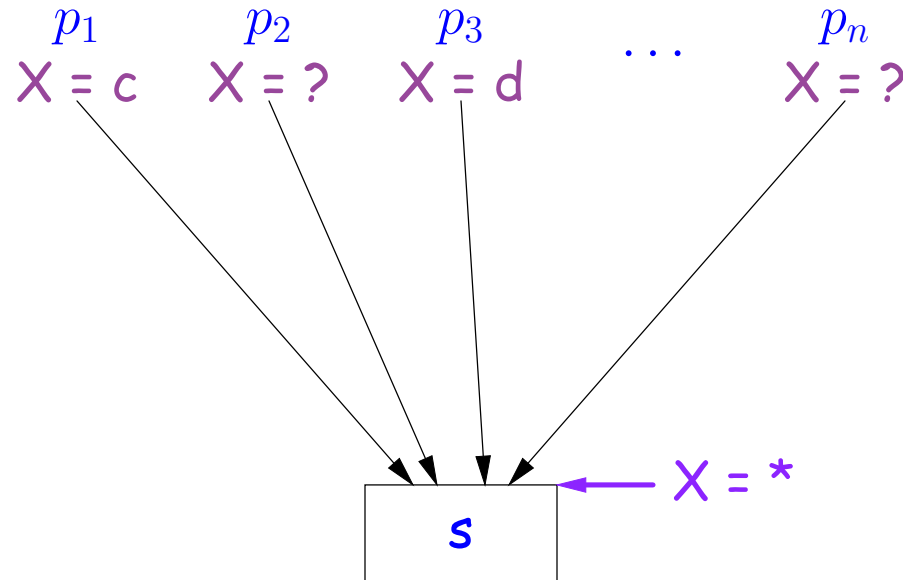
- Here, the “values of  $x$ ” we use come from an *abstract domain*, containing the values we care about— $\#, *, k$ —values computed *statically* by our analysis.
- For the constant propagation problem, we’ll compute  $Cout$  from  $Cin$ , and we’ll get  $Cin$  from the  $Couts$  of predecessor statements,  $Cout(X, p_1), \dots, Cout(X, p_n)$ .

# Constant Propagation: Rule 1



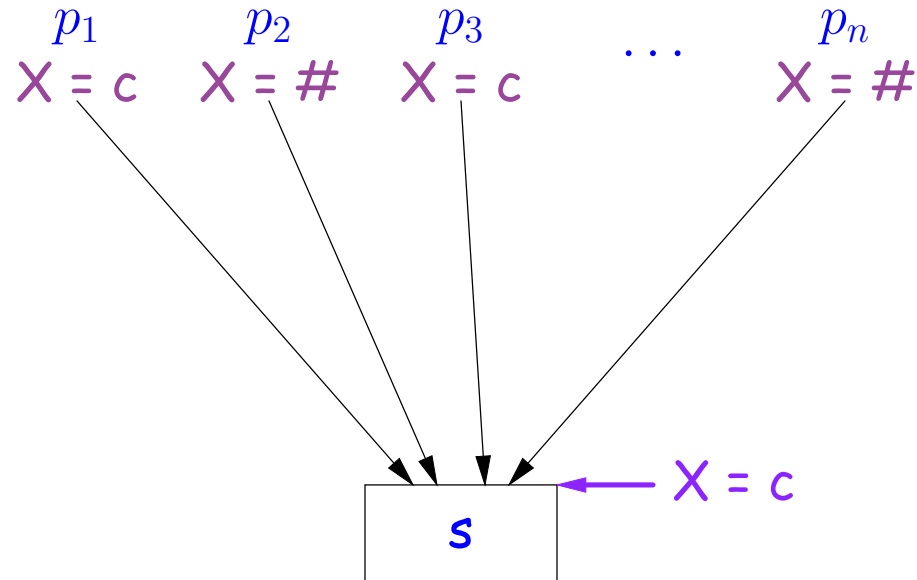
If  $\text{Cout}(X, p_i) = *$  for some  $i$ , then  $\text{Cin}(X, s) = *$

## Constant Propagation: Rule 2



If  $\text{Cout}(X, p_i) = c$  and  $\text{Cout}(X, p_j) = d$  with constants  $c \neq d$ ,  
then  $\text{Cin}(X, s) = *$

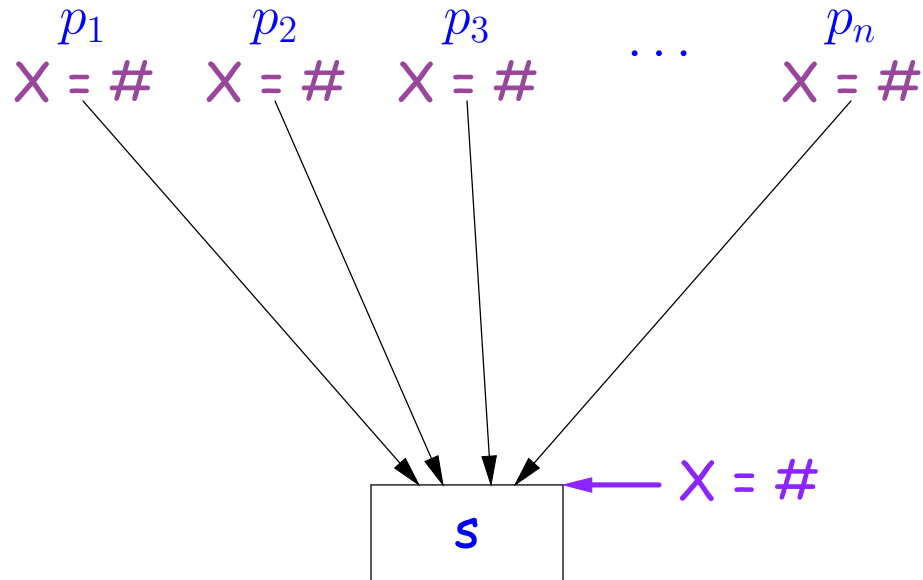
## Constant Propagation: Rule 3



If  $\text{Cout}(X, p_i) = c$  for some  $i$  and  
 $\text{Cout}(X, p_j) = c$  or  $\text{Cout}(X, p_j) = \#$  for all  $j$ ,  
then  $\text{Cin}(X, s) = c$



## Constant Propagation: Rule 4

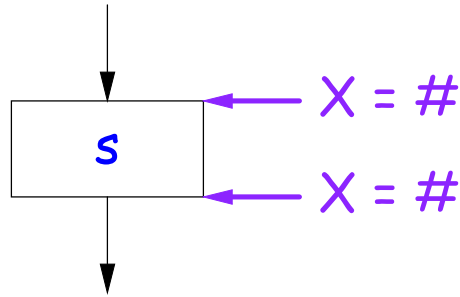


If  $\text{Cout}(X, p_j) = \#$  for all  $j$ , then  $\text{Cin}(X, s) = \#$

# Constant Propagation: Computing Cout

- Rules 1-4 relate the *out* of one statement to the *in* of the successor statements, thus propagating information *forward* across CFG edges.
- Now we need *local* rules relating the *in* and *out* of a single statement to propagate information across statements.

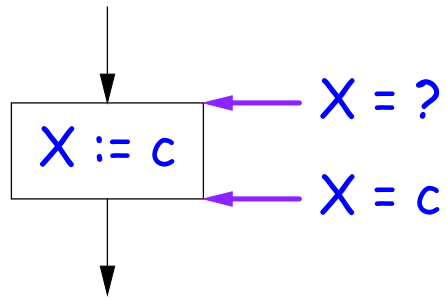
## Constant Propagation: Rule 5



$$\text{Cout}(X, s) = \# \text{ if } \text{Cin}(X, s) = \#$$

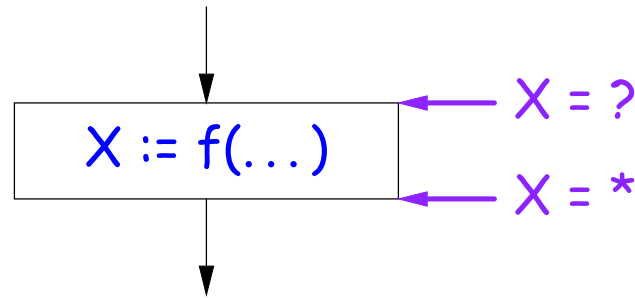
The value '#' means "so far, no value of  $X$  gets here, because we don't (yet) know that this statement ever gets executed."

## Constant Propagation: Rule 6



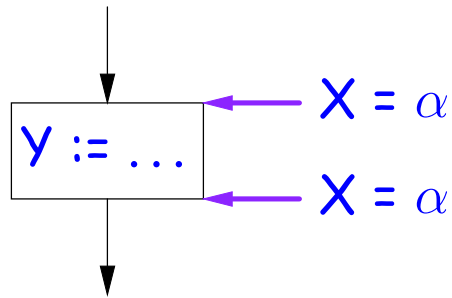
$\text{Cout}(X, X := c) = c$  if  $c$  is a constant and  $?$  is not  $\#$ .

## Constant Propagation: Rule 7



$\text{Cout}(X, X := f(\dots)) = *$  for any function call, if  $?$  is not  $\#$ .

## Constant Propagation: Rule 8

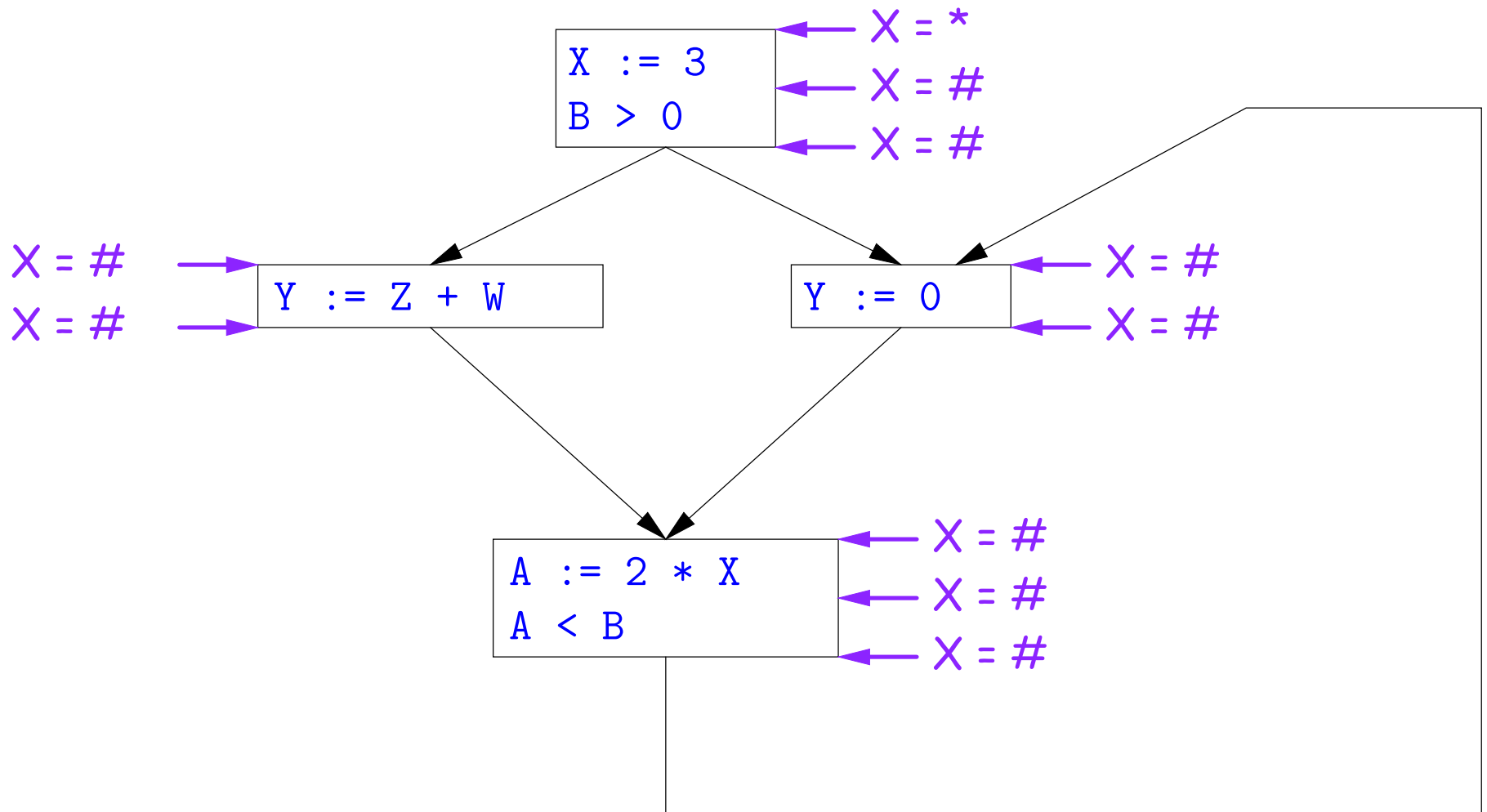


$C_{out}(X, Y := \dots) = C_{in}(X, Y := \dots)$  if  $X$  and  $Y$  are different variables.

# Propagation Algorithm

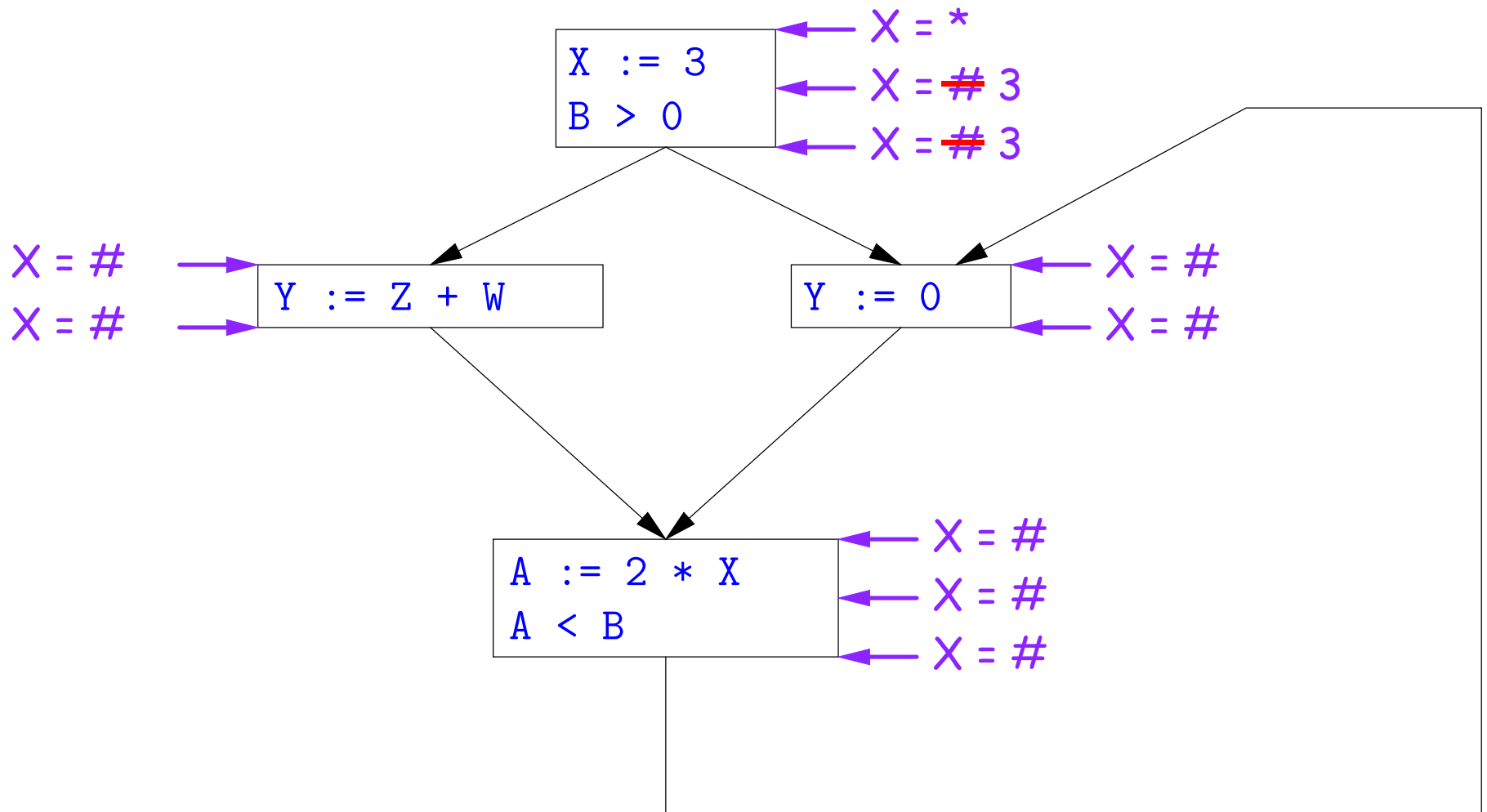
- To use these rules, we employ a standard technique: *iteration to a fixed point*:
- Mark all points in the program with *current approximations* of the variable(s) of interest ( $X$  in our examples).
- Set the initial approximations to  $X = *$  for the program entry point and  $X = \#$  everywhere else.
- Repeatedly apply rules 1-8 every place they are applicable until nothing changes—until the program is at a *fixed point* with respect to all the transfer rules.
- We can be clever about this, keeping a list of all nodes any of whose predecessors' *Cout* values have changed since the last rule application.

# An Example of the Algorithm

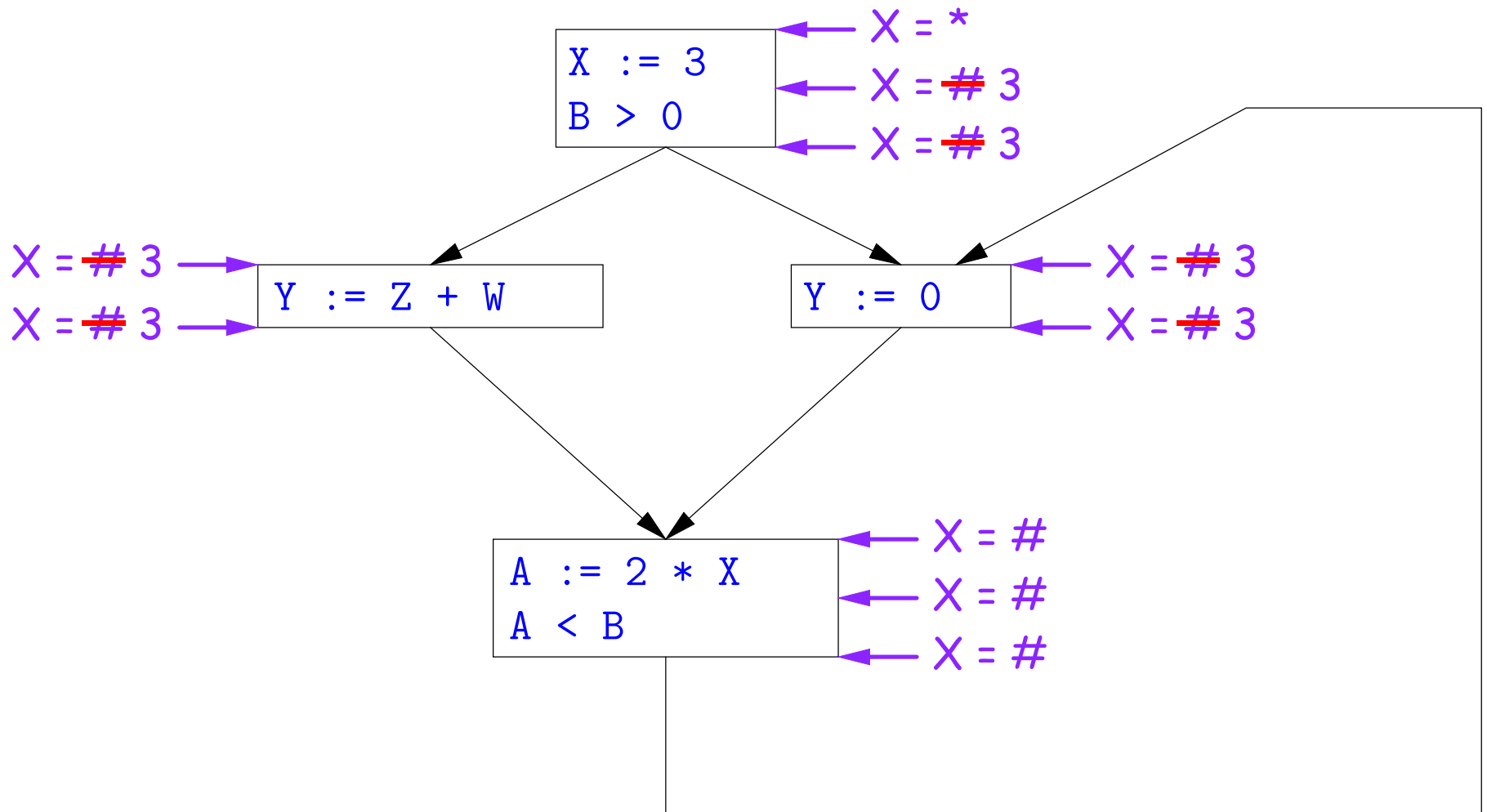




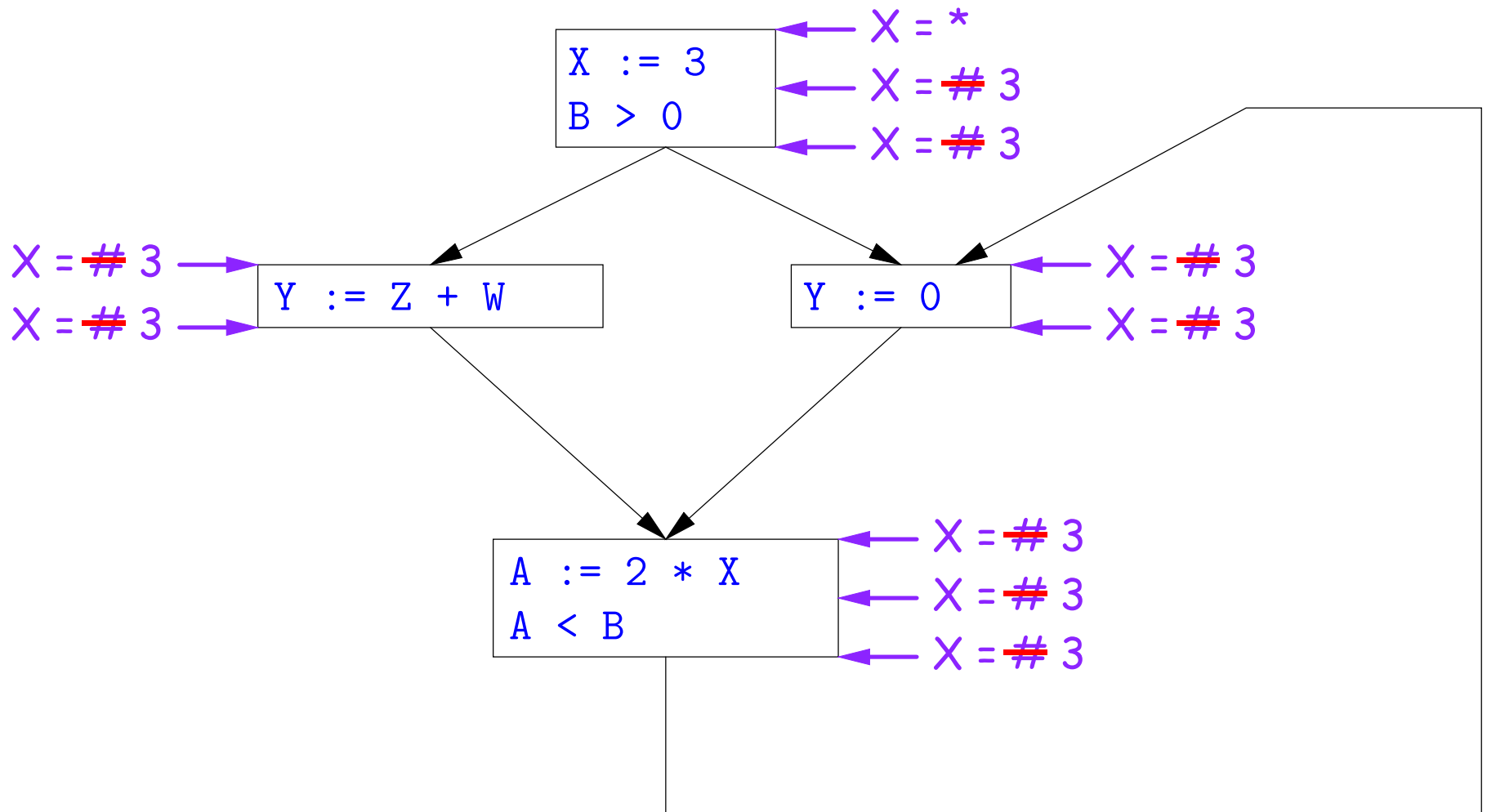
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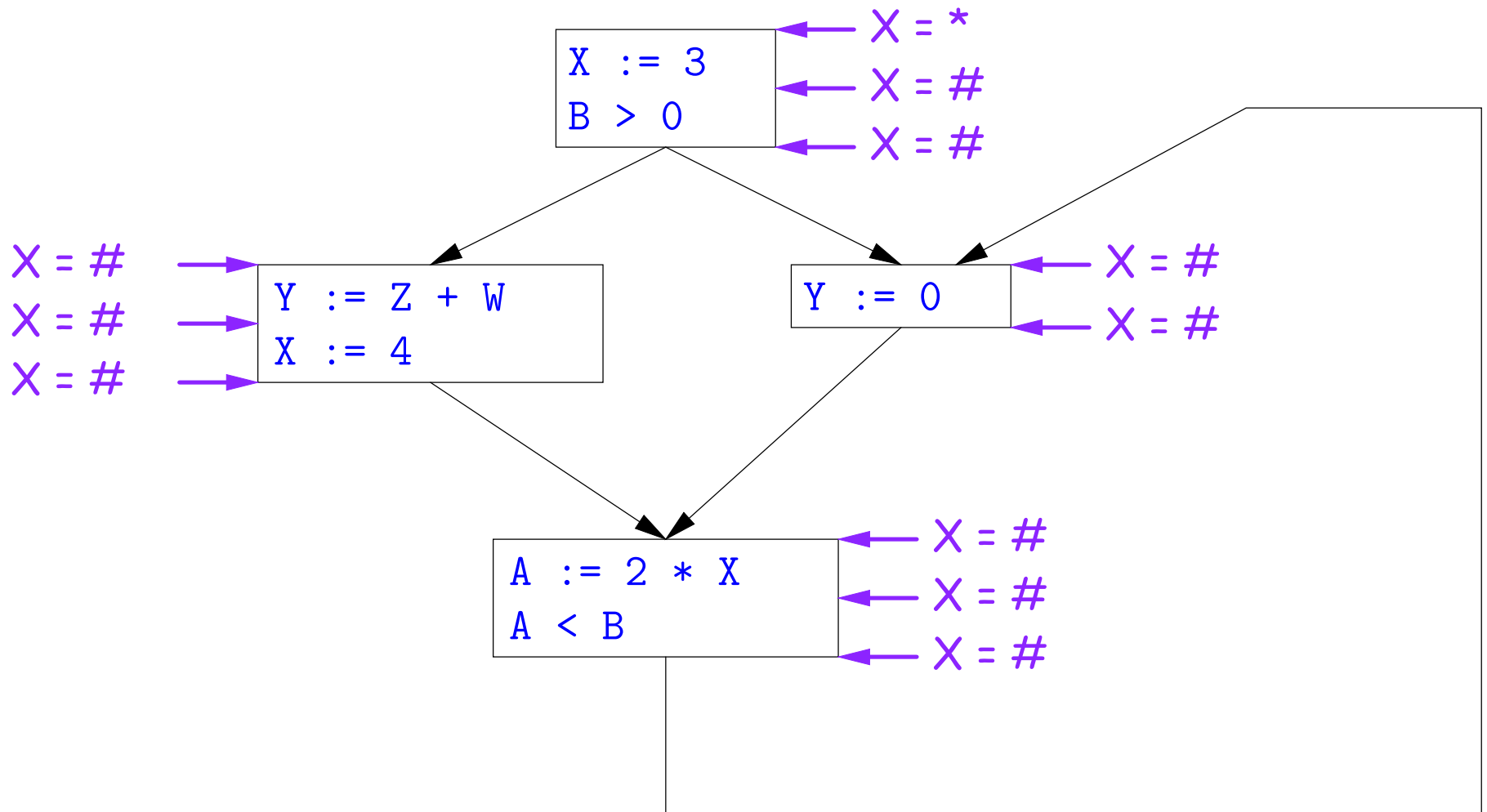


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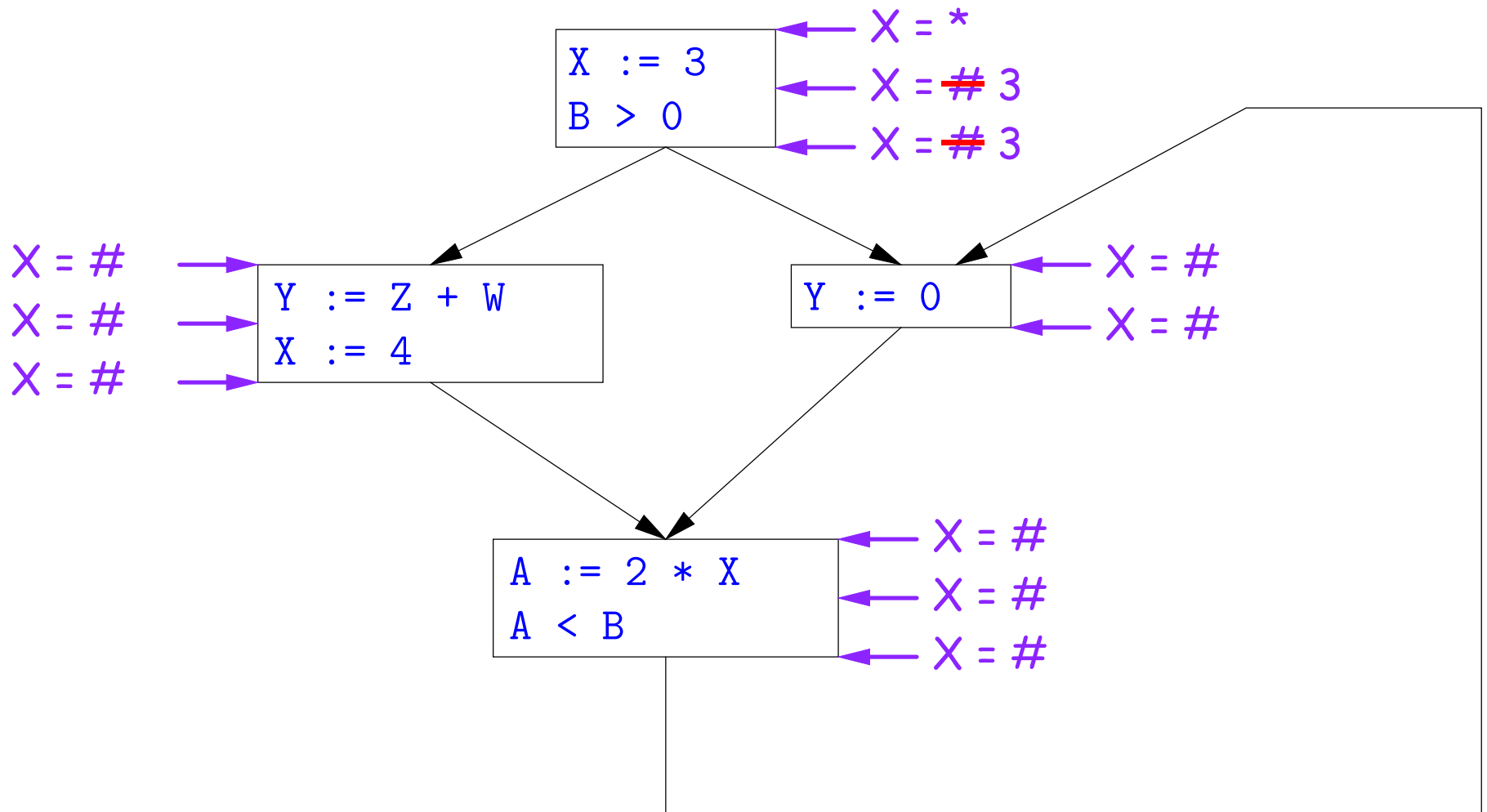


So we can replace `X` with `3` in the bottom block.

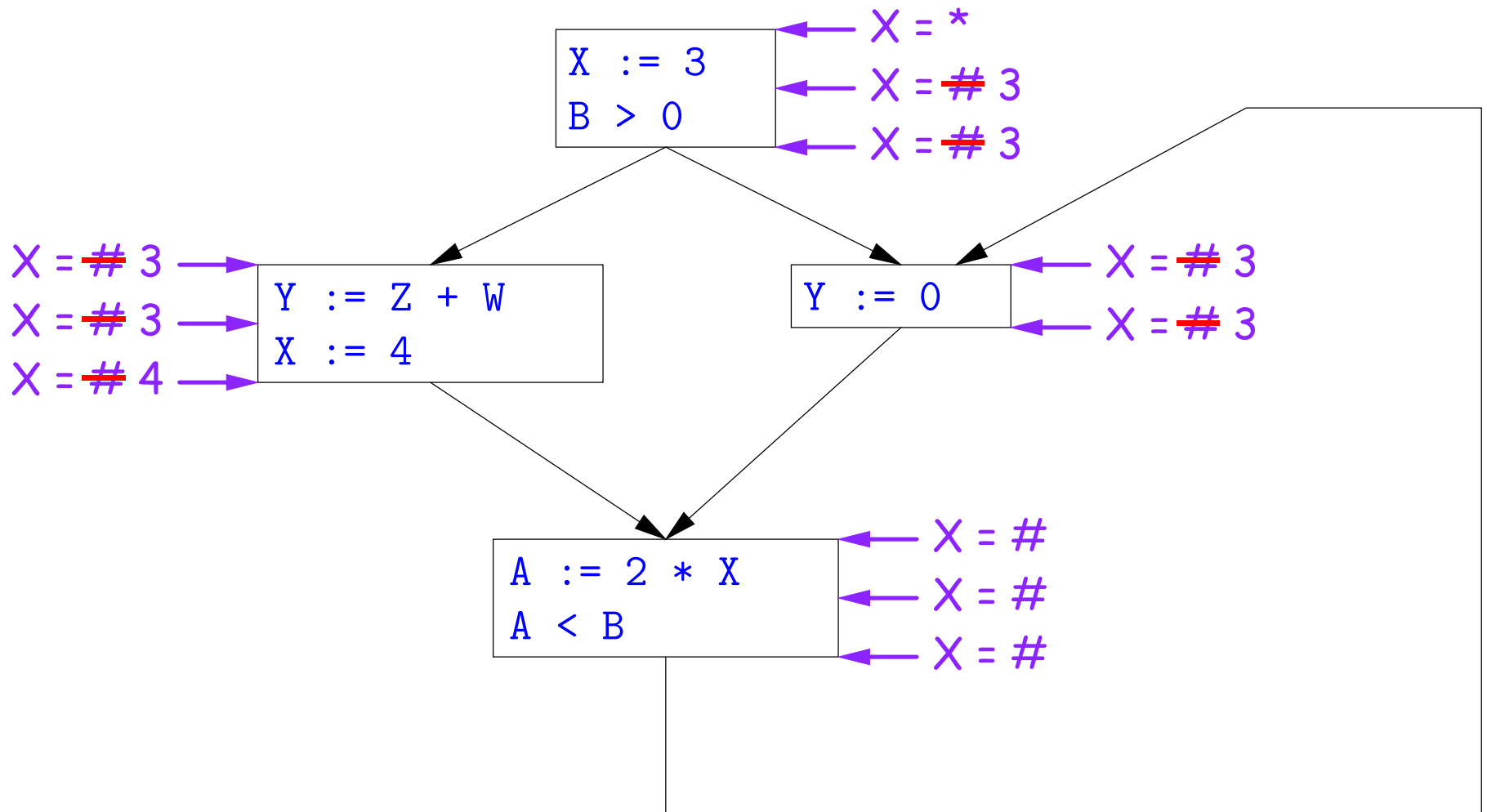
# Another Example of the Propagation Algorithm



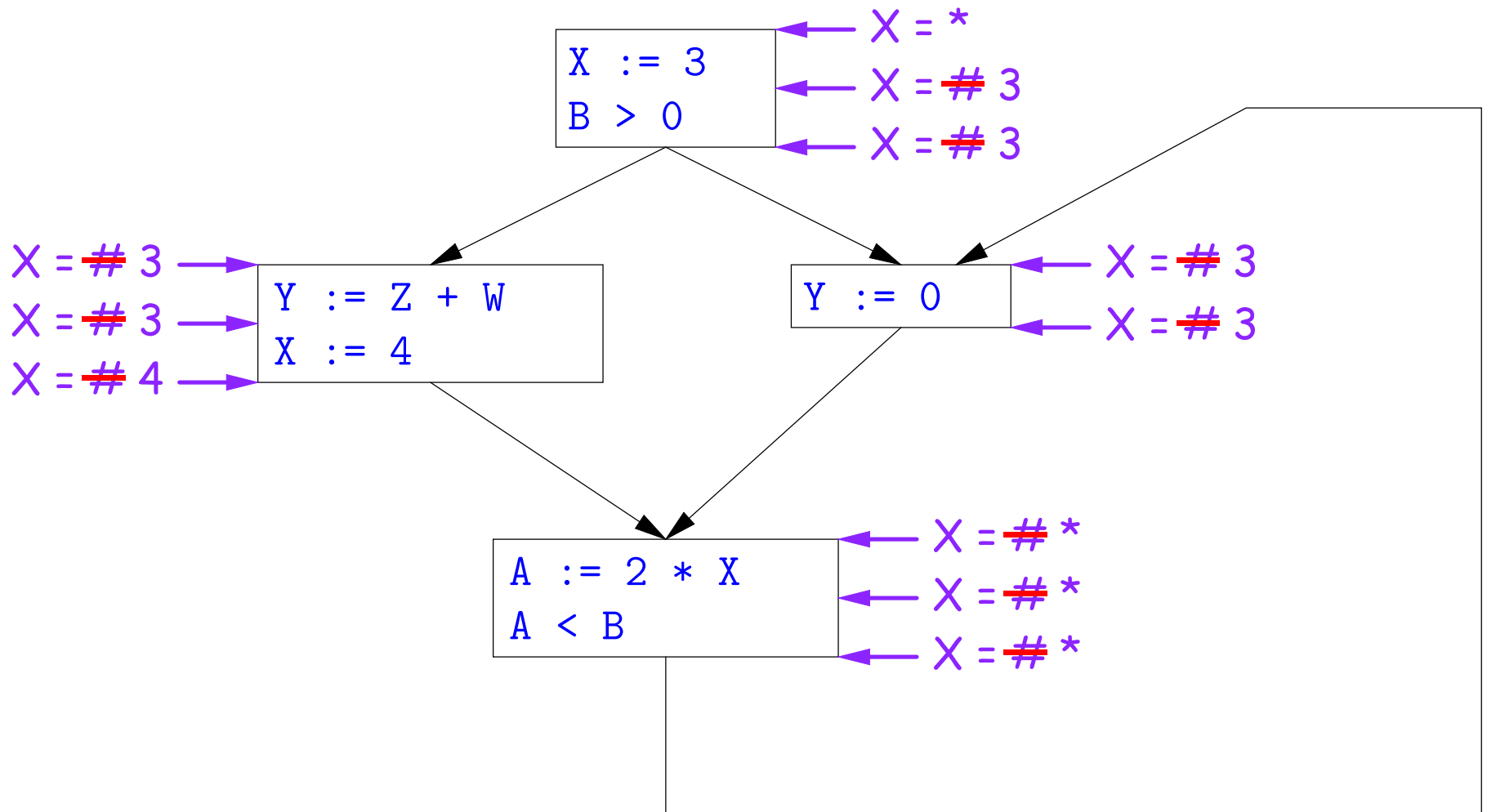
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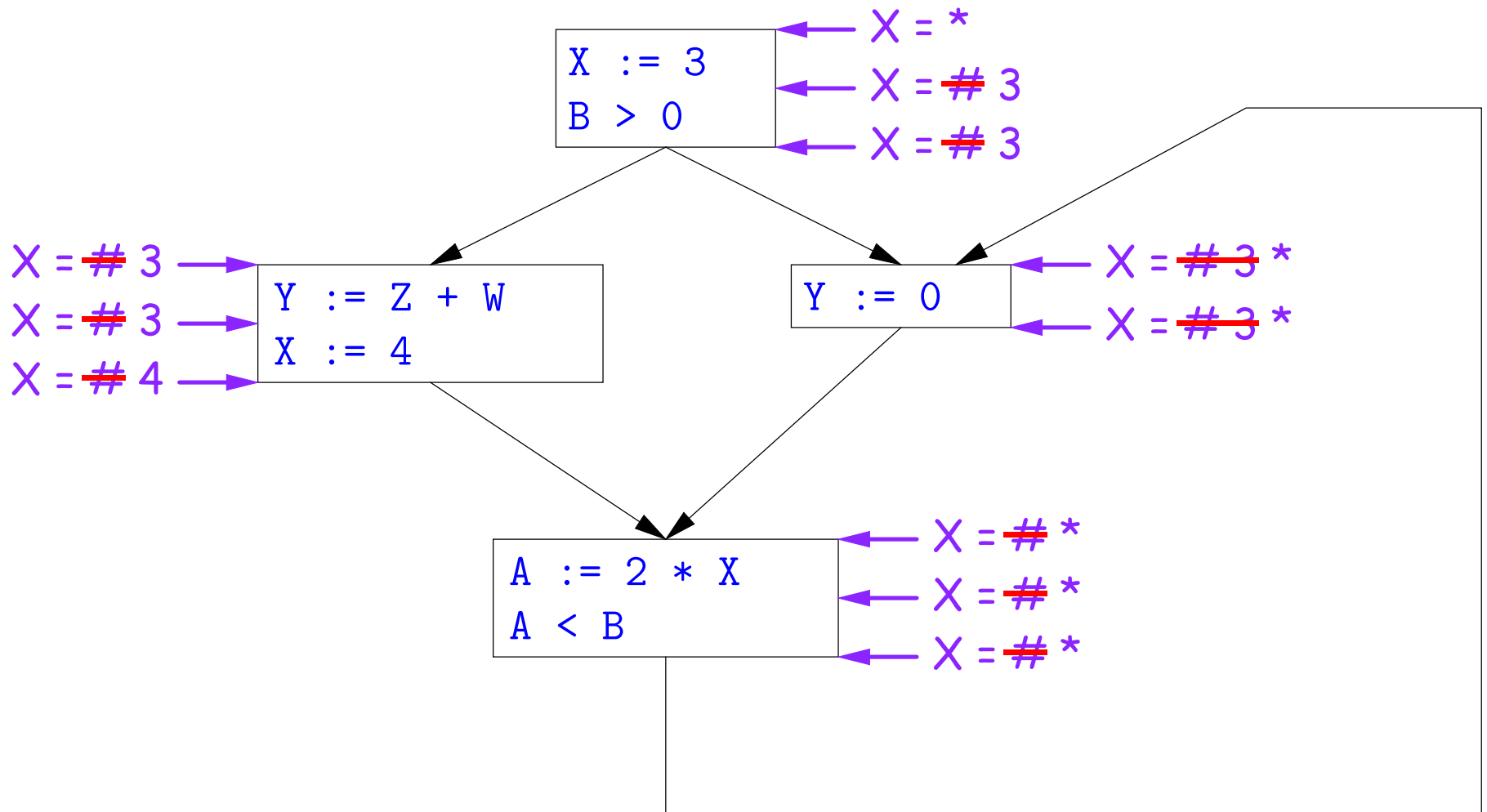
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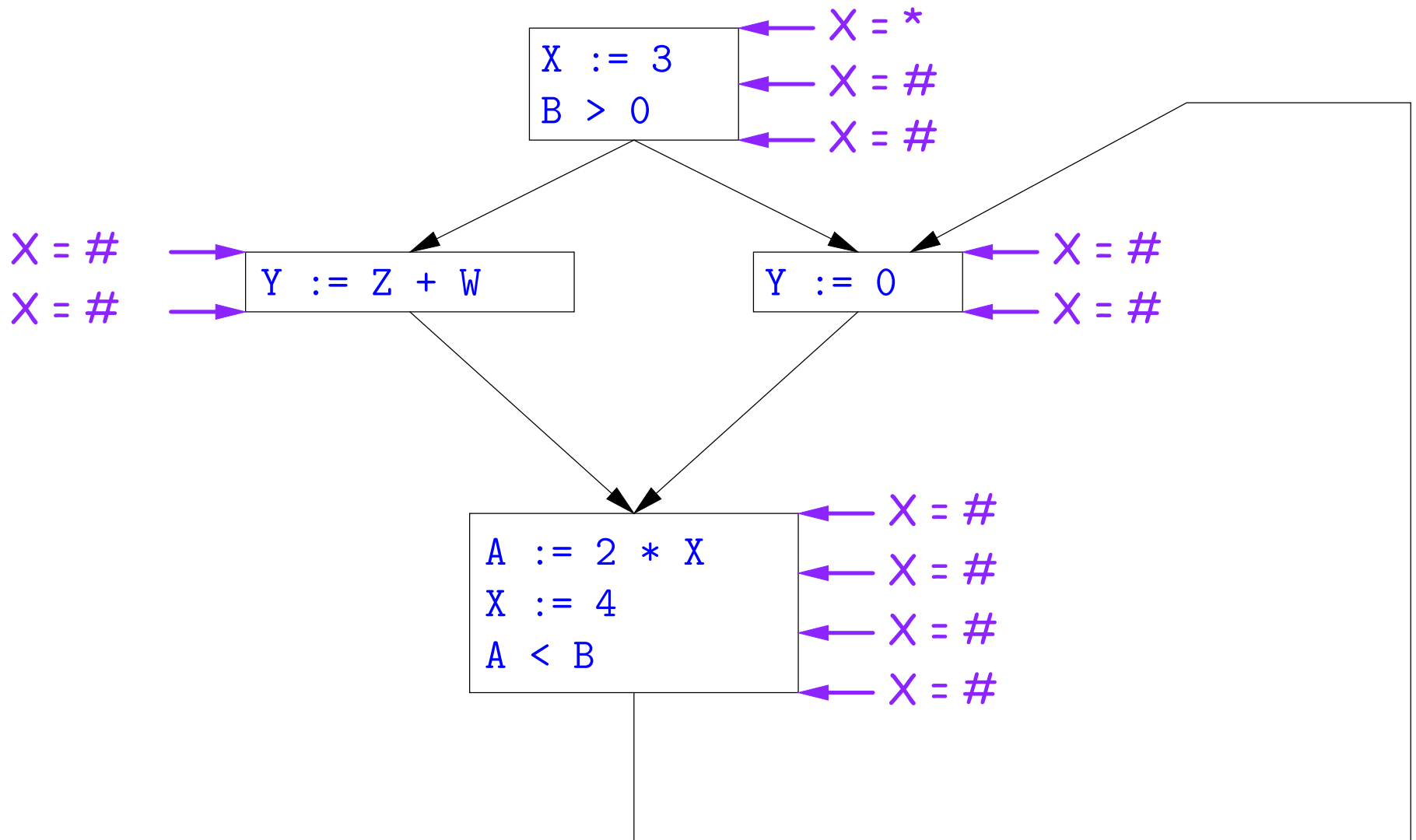
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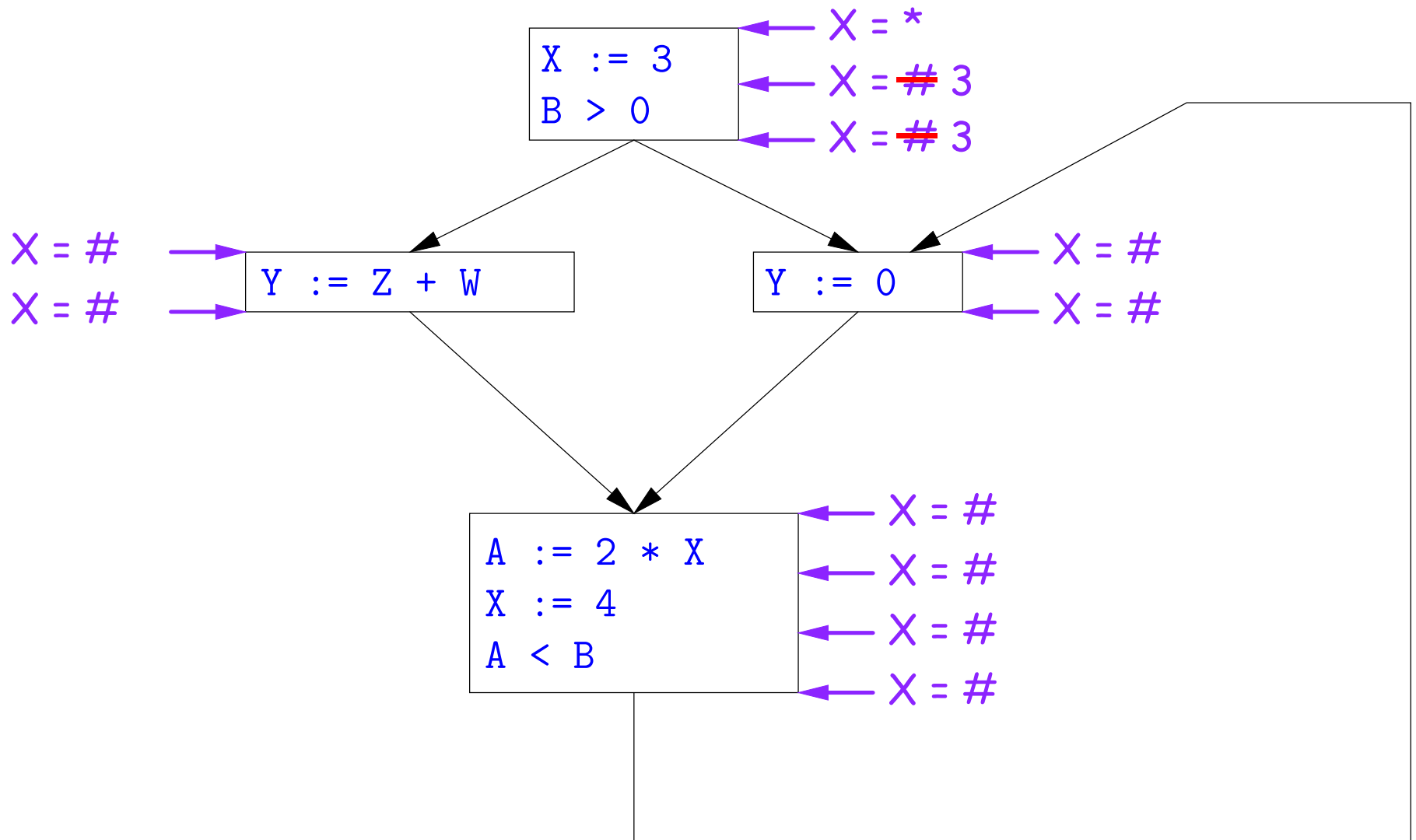
Here, we *cannot* replace  $X$  in two of the basic blocks.



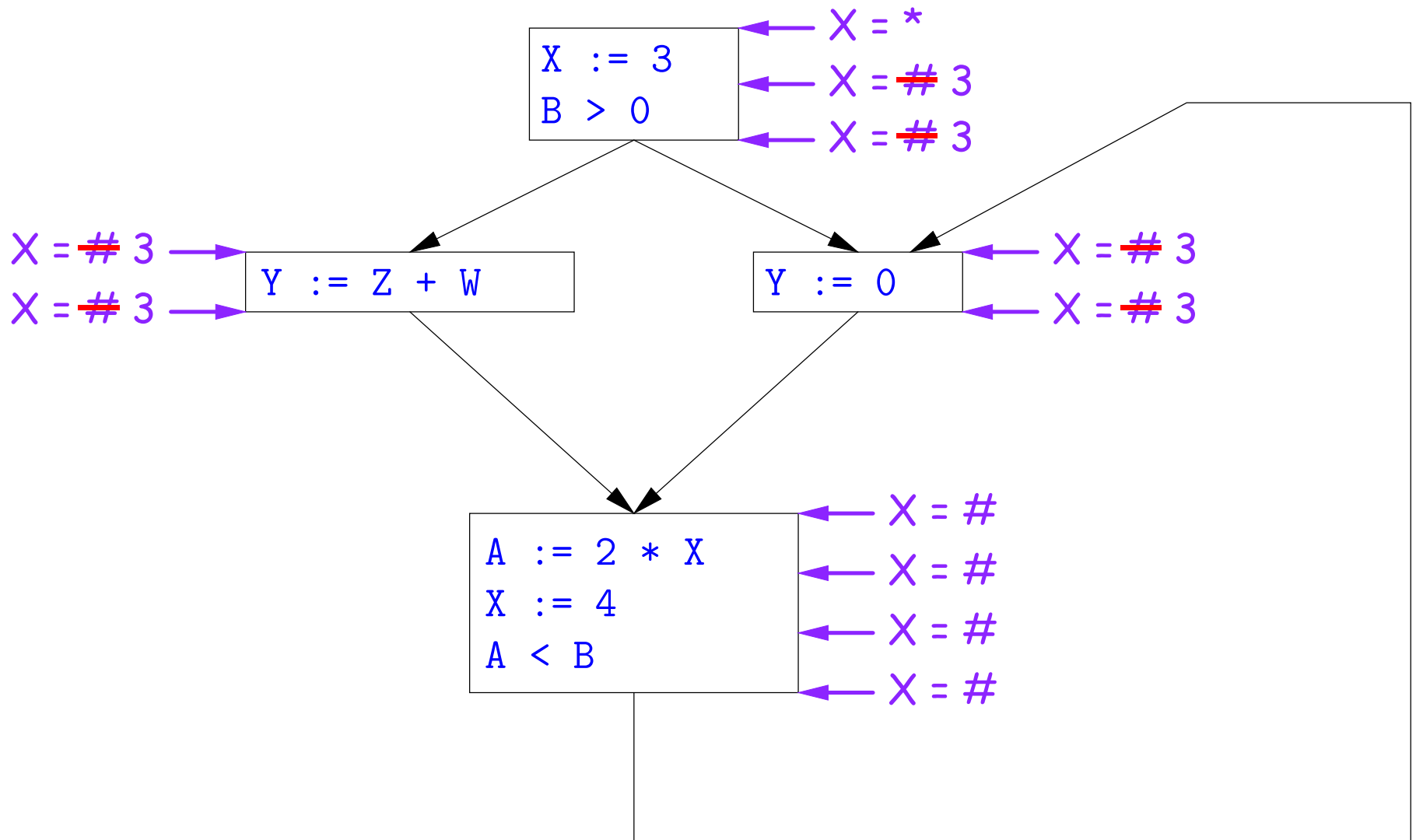
# A Third Example



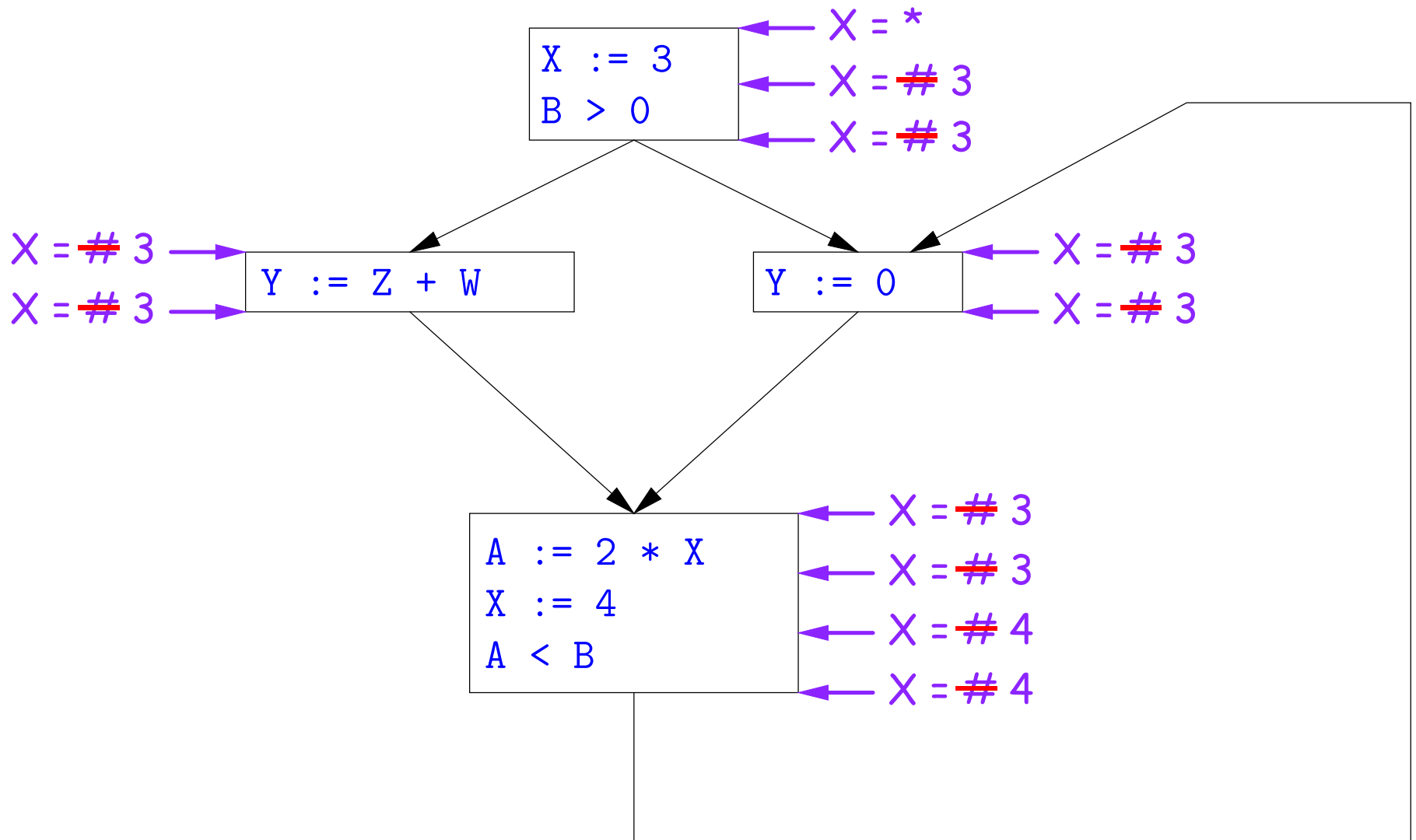
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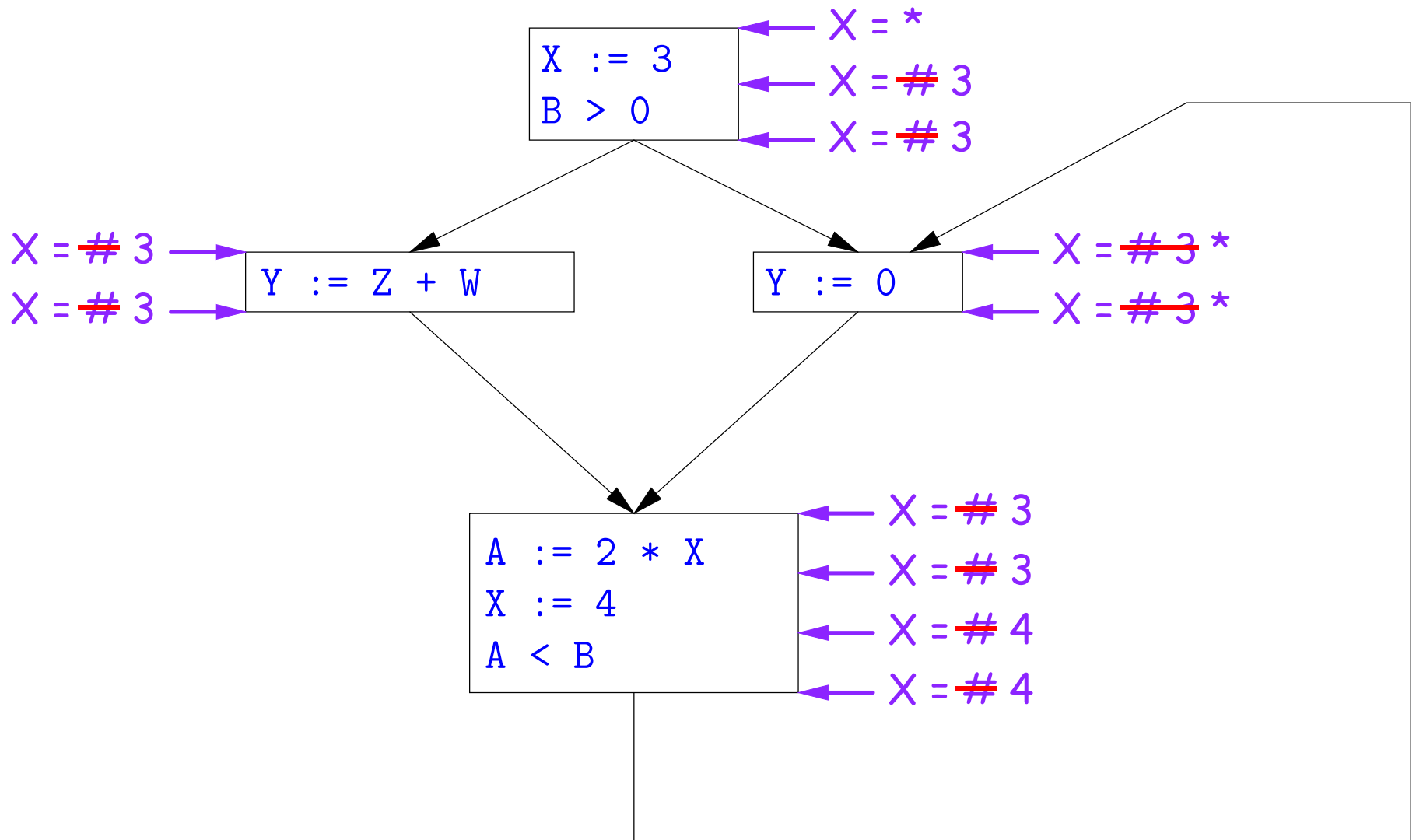
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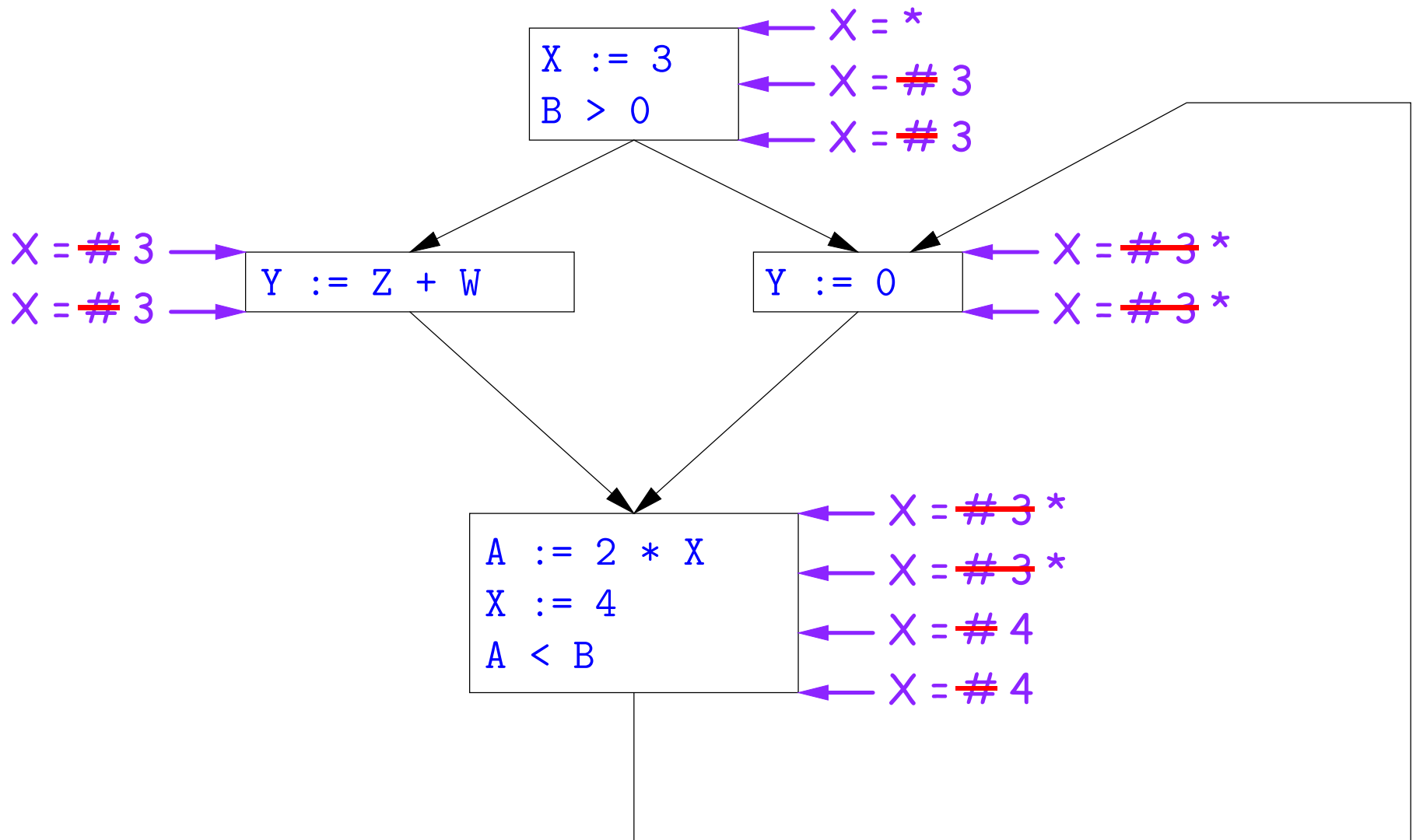
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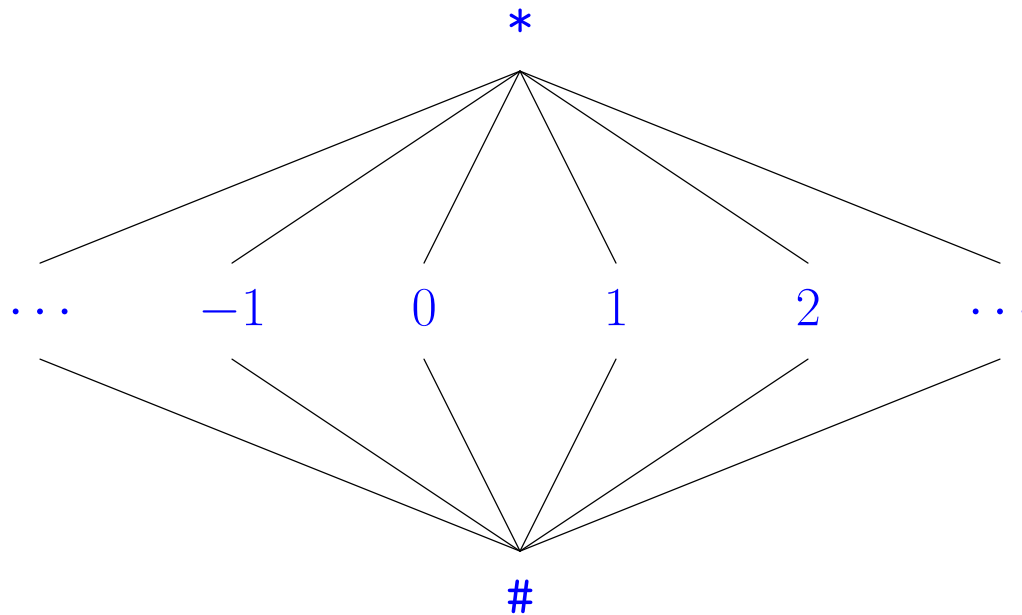
Likewise, we *cannot* replace  $X$ .

## Comments

- The examples used a breadth-first approach to considering possible places to apply the rules, starting from the entry point.
- In fact, the order in which one looks at statements is irrelevant. We could have changed the **Cout** values after the assignments to **X** first, for example.
- The **#** value is necessary to avoid deciding on a final value too soon. In effect, it allows us to tentatively propagate constant values through before finding out what happens in paths we haven't looked at yet.

# Ordering the Abstract Domain

- We can simplify the presentation of the analysis by ordering the values  $\# < c < *$ .
- Or pictorially, with lower meaning less than,



- ...a mathematical structure known as a *lattice*.
- With this, our rule for computing  $C_{in}$  is simply a *least upper bound*:

$$C_{in}(x, s) = \text{lub} \{ C_{out}(x, p) \text{ such that } p \text{ is a predecessor of } s \}.$$

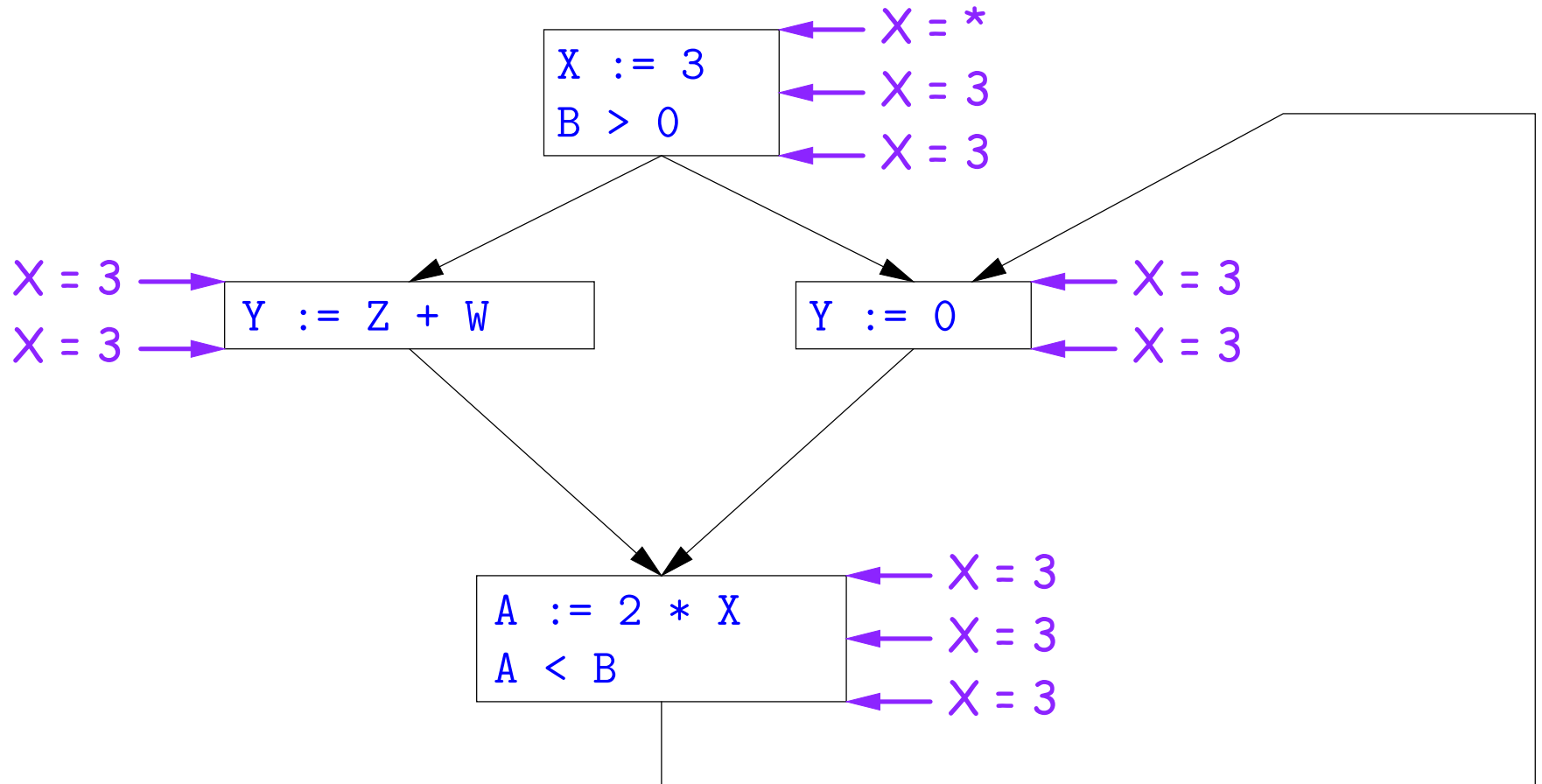


# Termination

- Simply saying “repeat until nothing changes” doesn’t guarantee that eventually nothing changes.
- But the use of lub explains why the algorithm terminates:
  - Values start as  $\#$  and only increase
  - By the structure of the lattice, therefore, each value can only change twice.
- Thus the algorithm is linear in program size. The number of steps
  - =  $2 \times$  Number of **Cin** and **Cout** values computed
  - =  $4 \times$  Number of program statements.

# Liveness Analysis

Once constants have been globally propagated, we would like to eliminate dead code



After constant propagation,  $X := 3$  is dead code (assuming this is the entire CFG)

# Terminology: Live and Dead

- In the program

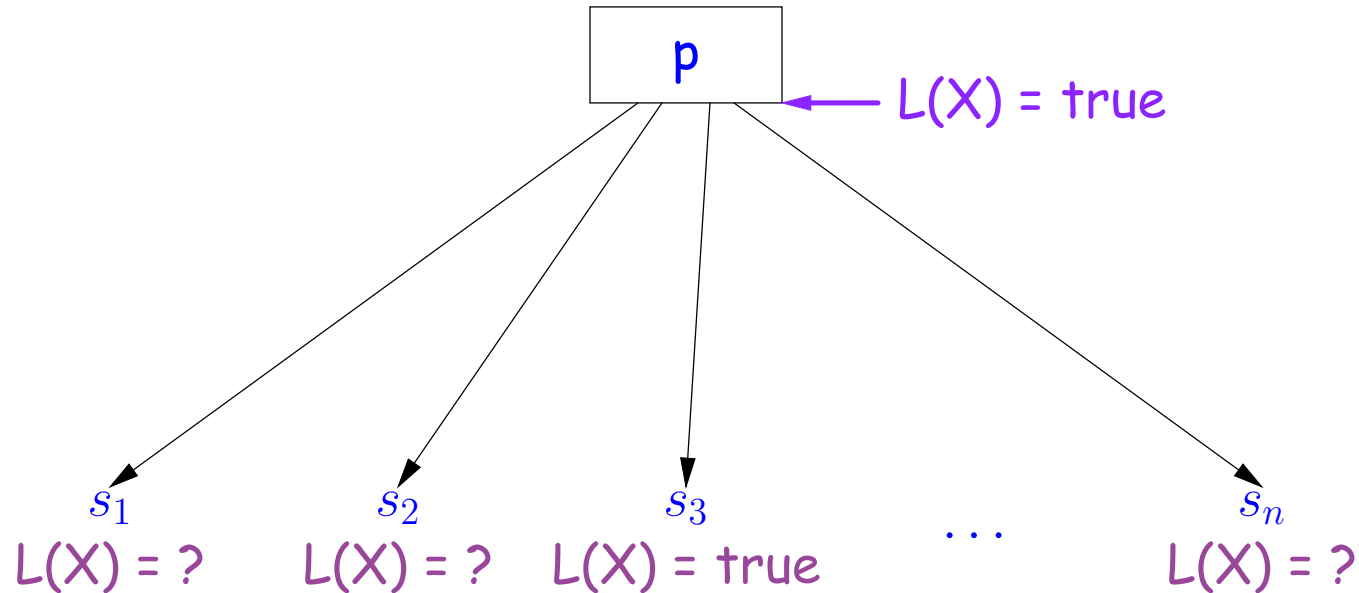
`X := 3; /*(1)*/ X = 4; /*(2)*/ Y := X /*(3)*/`

- the variable `X` is *dead* (never used) at point (1), *live* at point (2), and may or may not be live at point (3), depending on the rest of the program.
- More generally, a variable `x` is live at statement `s` if
  - There exists a statement `s'` that uses `x`;
  - There is a path from `s` to `s'`; and
  - That path has no intervening assignment to `x`
- A statement `x := ...` is dead code (and may be deleted) if `x` is dead after the assignment.

# Computing Liveness

- We can express liveness as a function of information transferred between adjacent statements, just as in copy propagation
- Liveness is simpler than constant propagation, since it is a boolean property (true or false).
- That is, the lattice has two values, with `false < true`.
- It also differs in that liveness depends on what comes *after* a statement, not before—we propagate information *backwards* through the flow graph, from `Lout` (liveness information at the end of a statement) to `Lin`.

# Liveness Rule 1

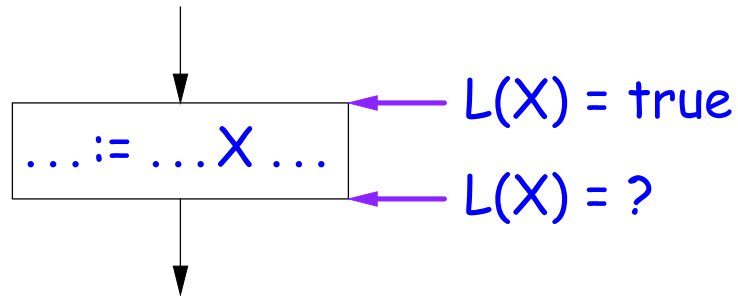


- So

$\text{Lout}(x, p) = \text{lub} \{ \text{Lin}(x, s) \text{ such that } p \text{ is a predecessor of } s \}.$

- Here, least upper bound (**lub**) is the same as "or".

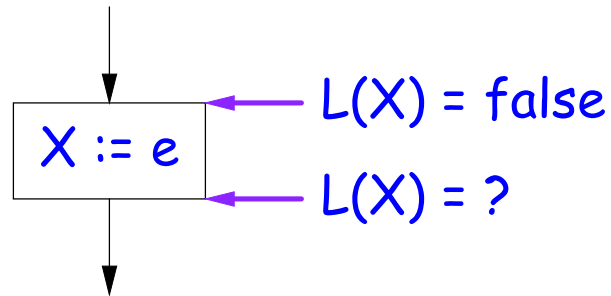
## Liveness Rule 2



$\text{Lin}(X, s) = \text{true}$  if  $s$  uses the previous value of  $X$ .

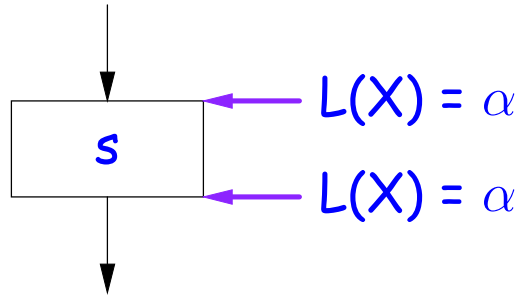
- The same rule applies to any other statement that uses the value of  $X$ , such as tests (e.g.,  $X < 0$ ).

## Liveness Rule 3



$\text{Lin}(X, X := e) = \text{false}$  if  $e$  does not use the previous value of  $X$ .

## Liveness Rule 4



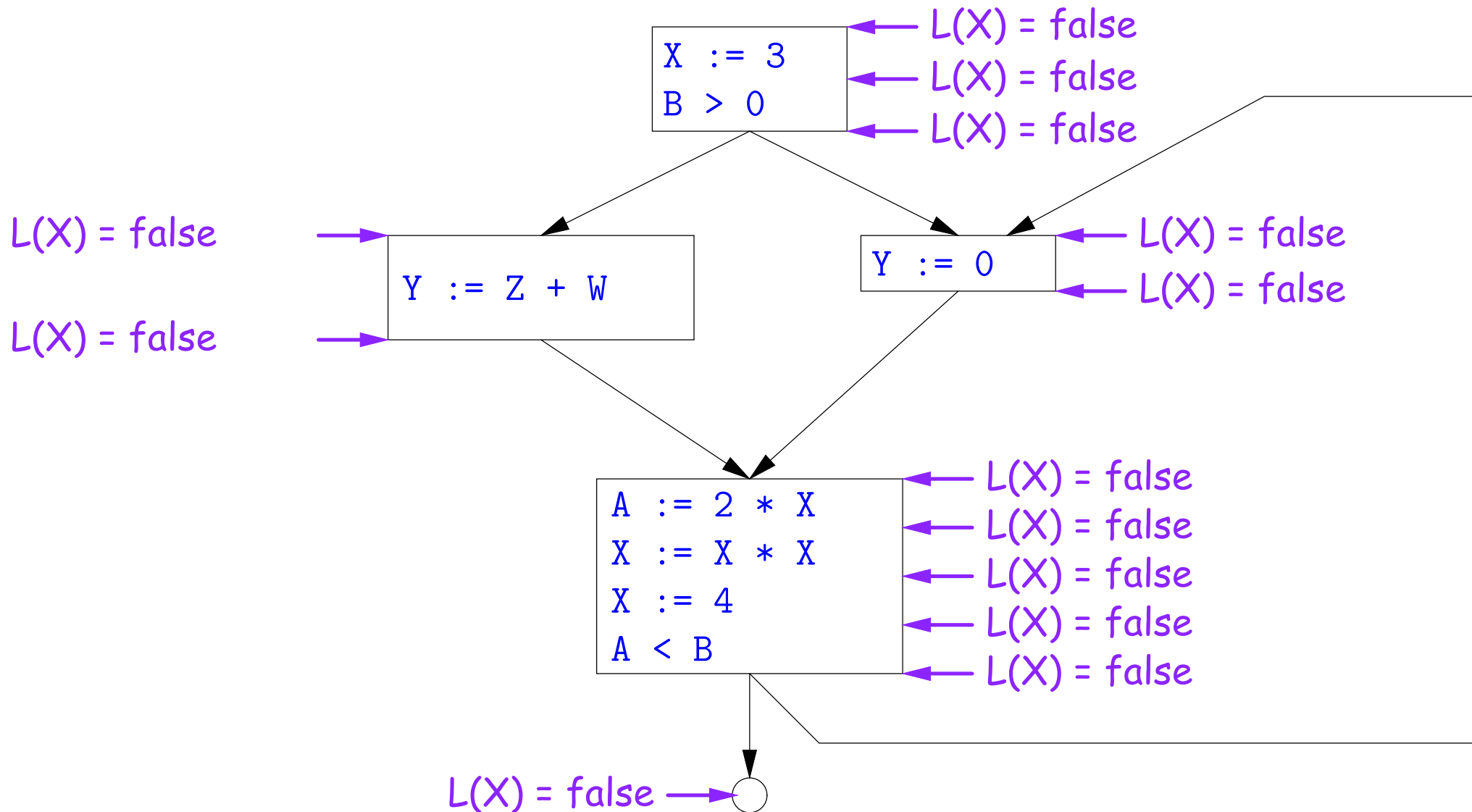
$L_{out}(X, s) = L_{in}(X, s)$  if  $s$  does not mention  $X$ .



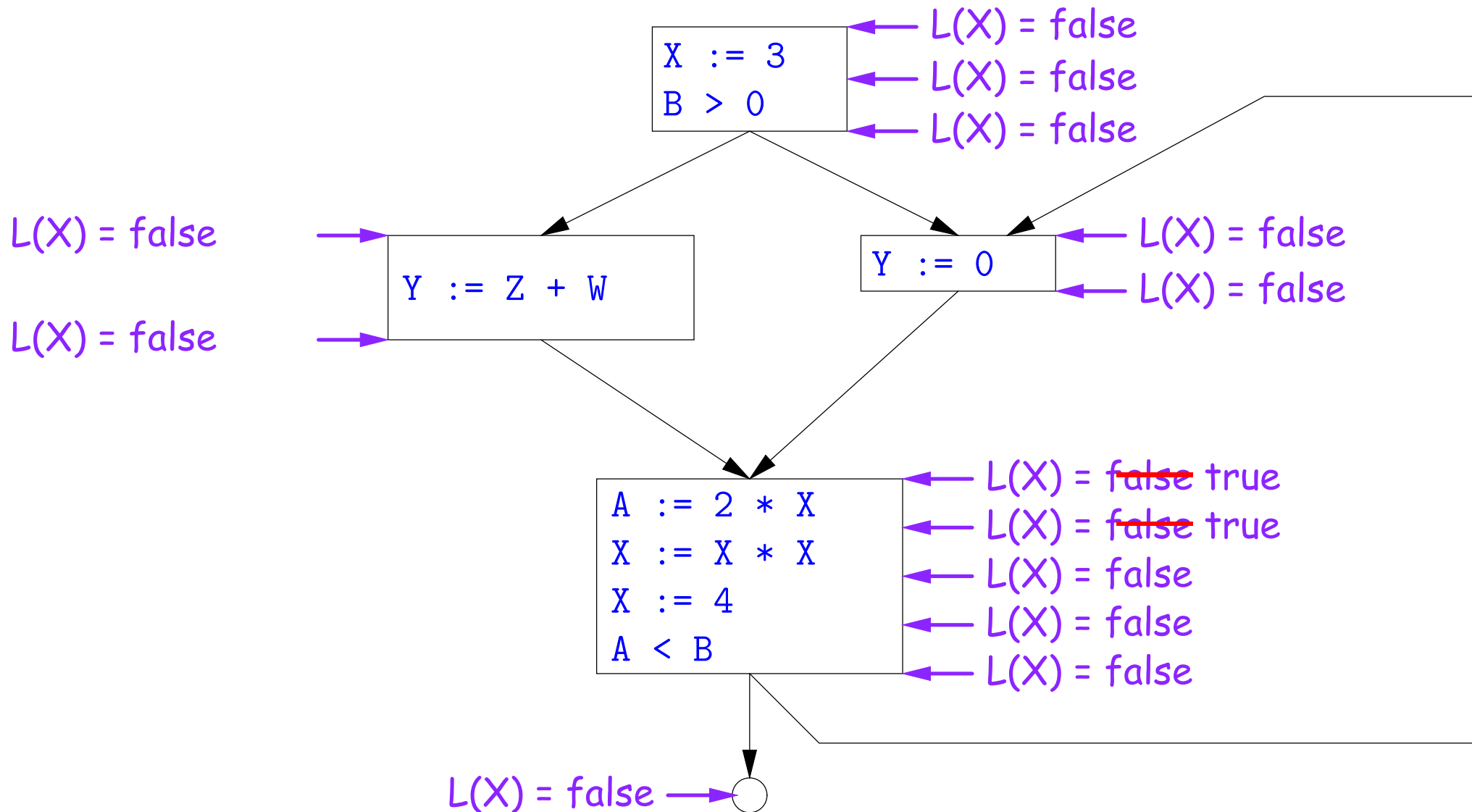
# Propagation Algorithm for Liveness

- Initially, let all **Lin** and **Lout** values be false.
- Set **Lout** value at the program exit to true iff **x** is going to be used elsewhere (e.g., if it is global and we are analyzing only one procedure).
- As before, repeatedly pick **s** where one of 1-4 does not hold and update using the appropriate rule, until there are no more violations.
- When we're done, we can eliminate assignments to **x** if **x** is dead at the point after the assignment.

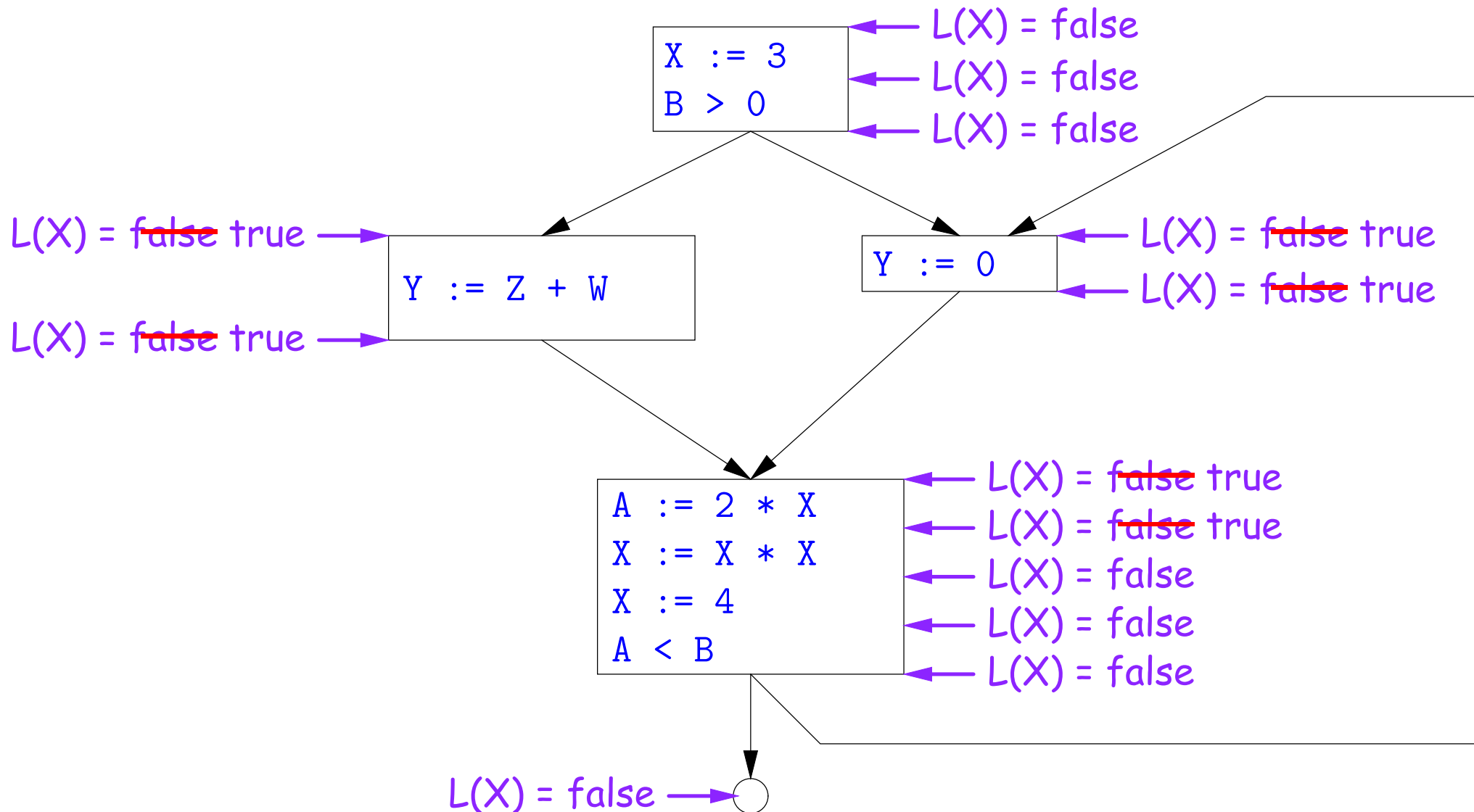
# Example of Liveness Computation



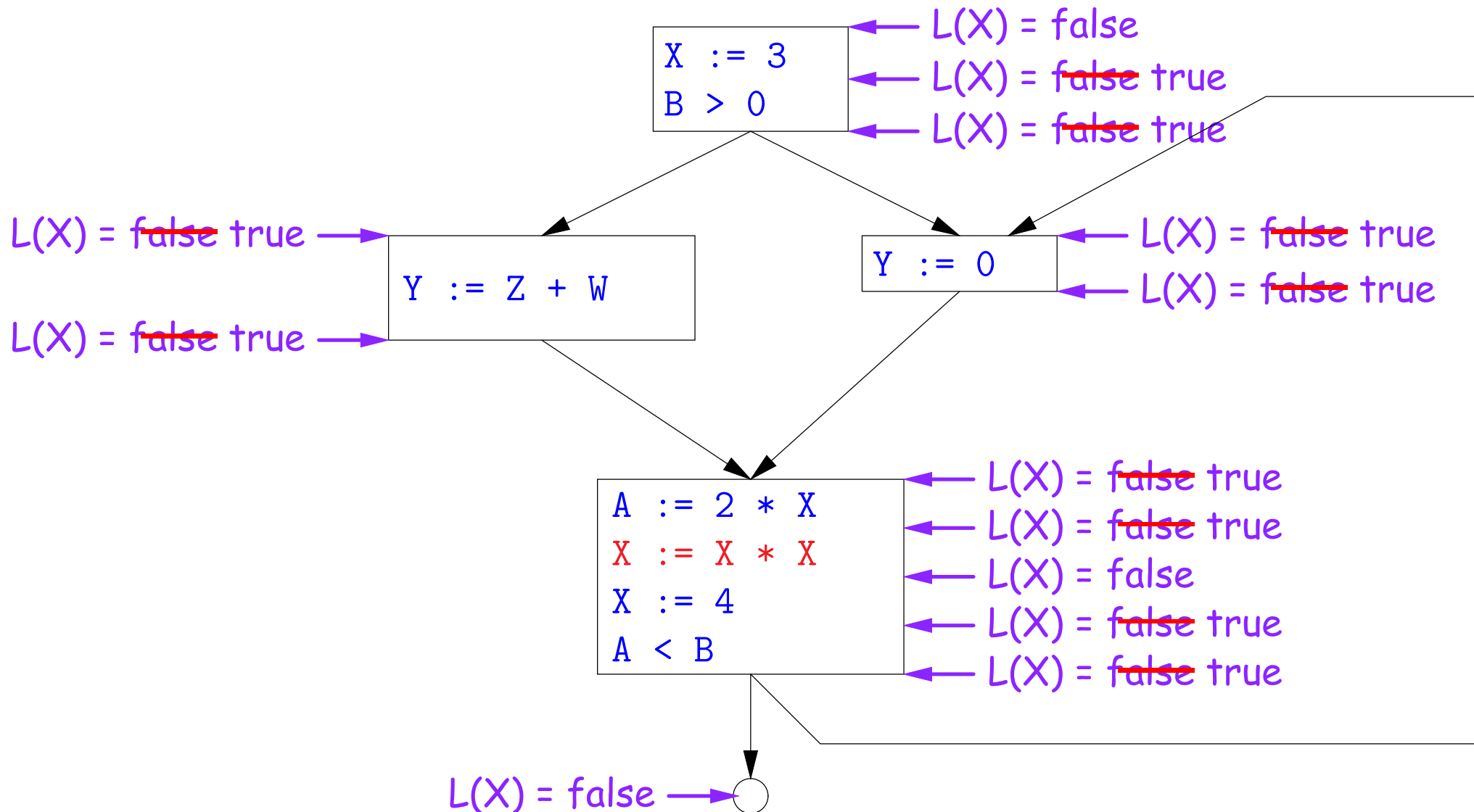
# Example of Liveness Computation



# Example of Liveness Computation



# Example of Liveness Computation



# Termination

- As before, a value can only change a bounded number of times: the bound being 1 in this case.
- Termination is guaranteed
- Once the analysis is computed, it is simple to eliminate dead code, but having done so, we must recompute the liveness information.

# SSA and Global Analysis

- For local optimizations, the single static assignment (SSA) form was useful.
- But applying it to a full CFG requires a trick.
- E.g., how do we avoid two assignments to the temporary holding  $x$  after this conditional?

```
if a > b:
    x = a
else:
    x = b
# where is x at this point?
```

- Answer: a small kludge known as  $\phi$  "functions"
- Turn the previous example into this:

```
if a > b:
    x1 = a
else:
    x2 = b
x3 =  $\phi$ (x1, x2)
```

## $\phi$ Functions

- An artificial device to allow SSA notation in CFGs.
- In a basic block, each variable is associated with one definition,
- $\phi$  functions in effect associate each variable with a set of possible definitions.
- In general, one tries to introduce them in strategic places so as to minimize the total number of  $\phi$ s.
- Although this device increases number of assignments in IL, register allocation can remove many by assigning related IL registers to the same real register.
- Their use enables us to extend such optimizations as CSE elimination in basic blocks to *Global CSE Elimination*.
- With SSA form, easy to tell (conservatively) if two IL assignments compute the same value: just see if they have the same right-hand side. The same variables indicate the same values.



# Loops

- In a CFG, a loop is simply a set of basic blocks,  $L$ , containing an entry block,  $e$ , such that
  - All paths from the entry node of the entire CFG to a block in  $L$  include  $e$ ;
  - All predecessors of a node in  $L$  are also in  $L$  (except for  $e$ , which must have a predecessor outside  $L$ ).
  - Every node in  $L$  has a path in  $L$  back to  $e$ .
- Here, for example,

```
j = i+1;  
while (j < N)  
    A[j] = A[j] / A[i]
```

The entry node contains the test  $j < n$  and the rest of the loop is the node containing the assignment to  $A[j]$ , which then loops back to the entry.

# Invariant Code Motion

- Consider the loop

```
while (i < N)
    A[i] = A[i] + j * x;
```

- Since  $j * x$  does not change in the loop, we can rewrite this as

```
tmp = j * x;
while (i < N)
    A[i] = A[i] + tmp;
```

- This is an example of *invariant code motion out of a loop*.
- What tells us that  $j*x$  does not change?
- We see that all assignments to  $j$  and  $x$  that apply at the point where the product is computed are outside the loop.
- And this we can get by observing where the assignments to the SSA-form for those variables are.

# Code Motion Caveat

- Code motion is not always appropriate.
- If the code to be moved, has side effects, or might cause an exception, could change the results.
- If the code is expensive, you will increase the time required for the program when the loop is not executed.
- Hence, you will see compilers rewrite loops like this:

```
if (i < N) {  
    /* Preheader */  
    while (i < N)  
        A[i] = A[i] + j * x;  
}
```

where `Preheader` marks a spot where the compiler can insert a new block to hold code moved out of the loop.

# Summary

- We've seen two kinds of analysis:
  - Constant propagation is a *forward analysis*: information is pushed from inputs to outputs.
  - Liveness is a *backwards analysis*: information is pushed from outputs back towards inputs.
- But both make use of essentially the same algorithm.
- Numerous other analyses fall into these categories, and allow us to use a similar formulation:
  - An abstract domain (abstract relative to actual values);
  - Local rules relating information between consecutive program points around a single statement; and
  - Lattice operations like least upper bound (or *join*) or greatest lower bound (or *meet*) to relate inputs and outputs of adjoining statements.