



Project 1: 4 days left

# Prevention-Based Cyber Security: Program Testing

CS 459/559: Science of Cyber Security  
11<sup>th</sup> Lecture

**Instructor:**

Guanhua Yan

# Agenda

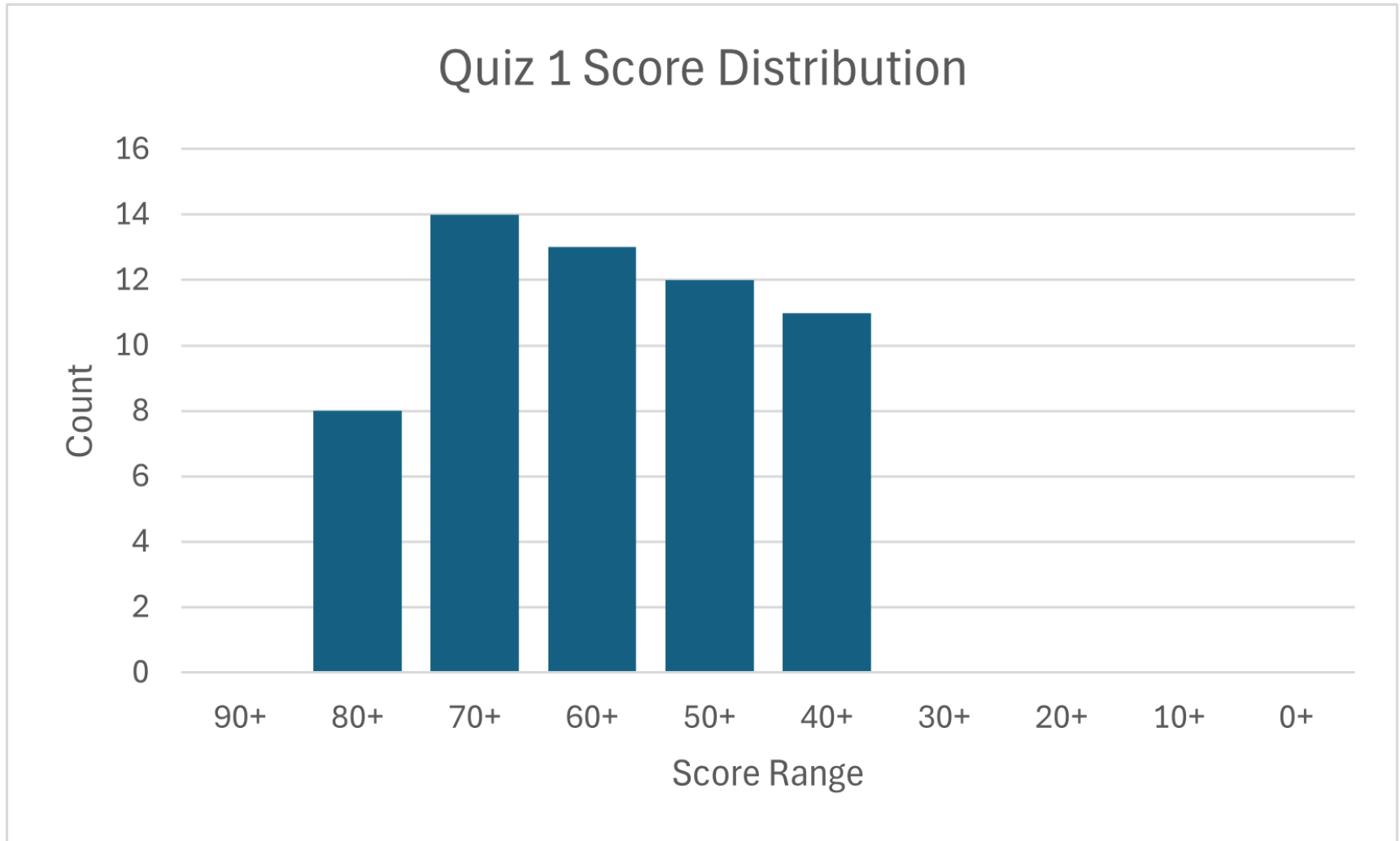
- ~~Quiz 1: September 29 (closed book)~~
- Project 1 (offense): October 10
- Project 2 (defense): December 5
- Presentations: 11/17, 11/19, 11/24, 12/1, 12/3
- Final report: December 15



# How do you think of first quiz?

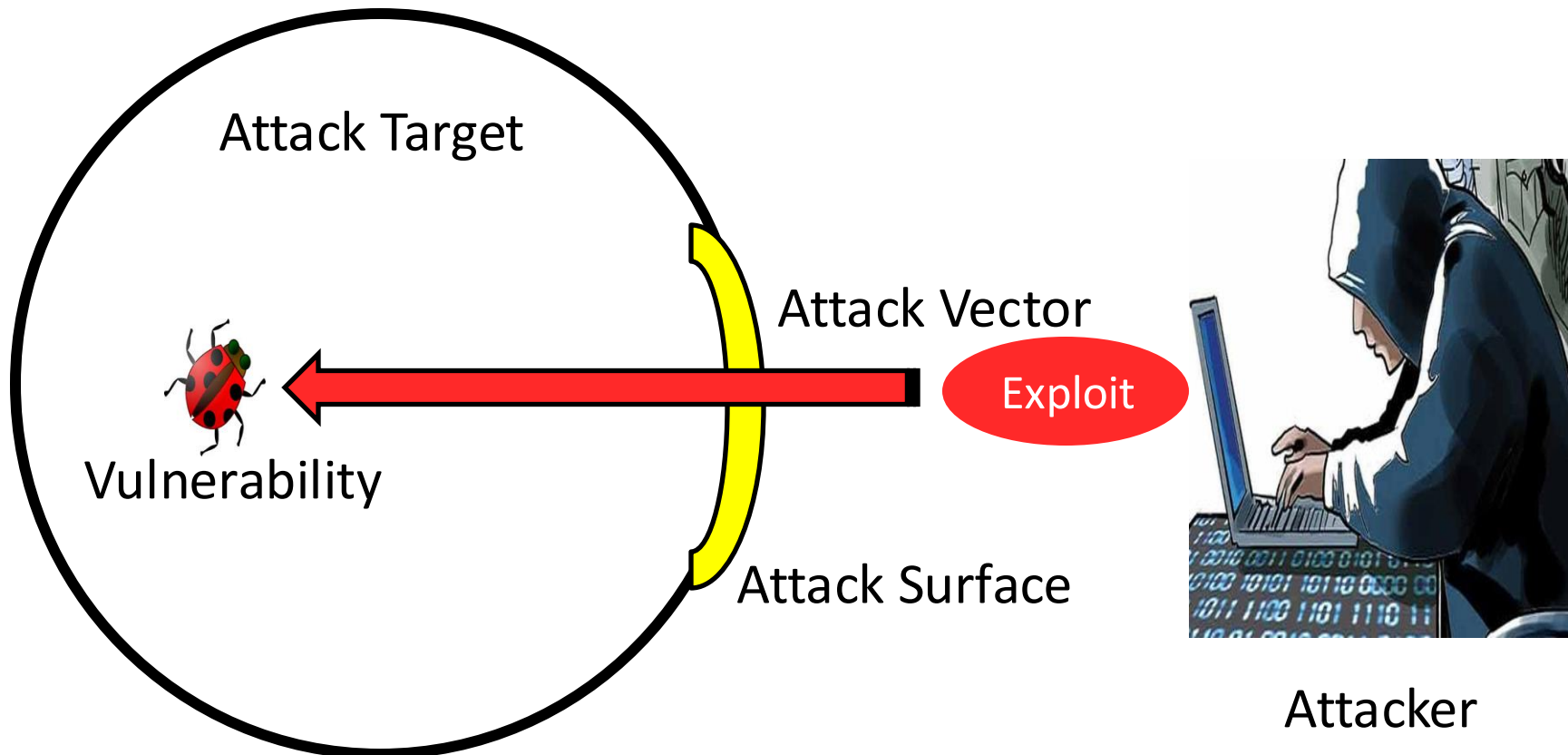
- A: Too difficult
- B: Too easy
- C: About right
- D: Refuse to comment

# Quiz 1 score distribution



# Goal of program testing

- Develop techniques to discover and patch vulnerabilities before they are exploited



# Outline

- **Fuzzing**
- **Symbolic execution**
- **Concolic execution**

# Fuzzing

# ••• | Fuzz Testing

- Run program on many **random, abnormal** inputs and look for bad behavior in the responses
  - Bad behaviors such as crashes or hangs
- What are the benefits of fuzz testing over regular testing?



## ● ● ● | Fuzz Testing (Bart Miller, U. Of Wisconsin)

- A night in 1988 with thunderstorm and heavy rain
- Connected to his office Unix system via a dial up connection
- The heavy rain introduced noise on the line
- Crashed many UNIX utilities he had been using everyday
- He realized that there was something deeper
- Asked three groups in his grad-seminar course to implement this idea of fuzz testing
  - Two groups failed to achieve any crash results!
  - The third group succeeded! Crashed 25-33% of the utility programs on the seven Unix variants that they tested

# ● ● ● | Fuzz Testing

- Approach
  - Generate random inputs
  - Run lots of programs using random inputs
  - Identify crashes of these programs
  - Correlate random inputs with crashes
- **Errors found:** Not checking returns, Array indices out of bounds, not checking null pointers, ...



# Fuzz Testing Overview

- Black-box fuzzing
  - Treating the system as a blackbox during fuzzing; not knowing details of the implementation
- Grey-box fuzzing
- White-box fuzzing
  - Design fuzzing based on internals of the system

## ● ● ● | Black Box Fuzzing

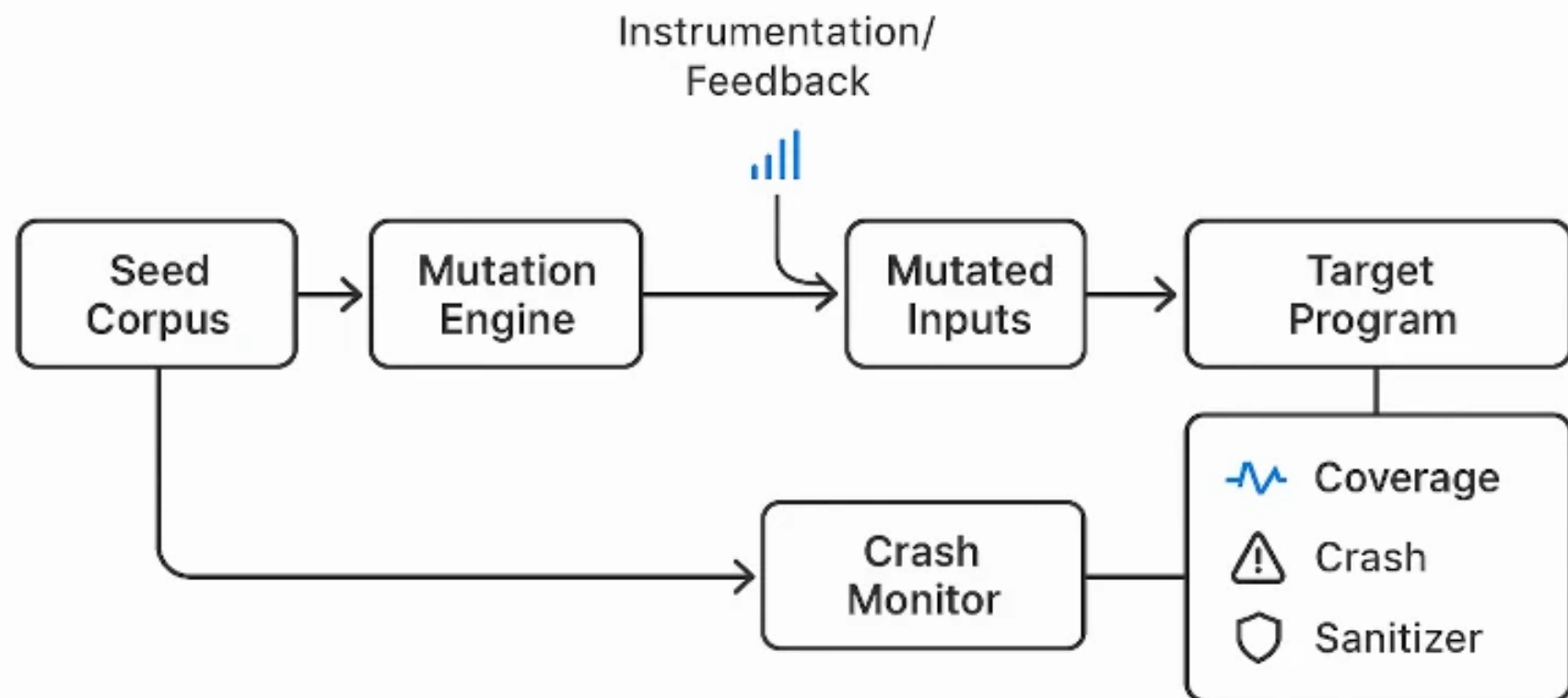
- Example that would be hard for black box fuzzing to find the error

```
function( char *name, char *passwd, char *buf )
{
    if ( authenticate_user( name, passwd ) ) {
        if ( check_format( buf ) ) {
            update( buf ); // crash here
        }
    }
}
```

## • • • | Mutation-Based Fuzzing

- User supplies a well-formed input
- Fuzzing: Generate random changes to that input
- No assumptions about input
  - Only assumes that variants of well-formed input may be problematic
- Example: zzuf
  - <http://sam.zoy.org/zzuf/>
  - Reading: The Fuzzing Project Tutorial

# MUTATION FUZZING WORKFLOW



# ••• | Mutation-Based Fuzzing

## ○ The Fuzzing Project Tutorial

- `zzuf -s 0:1000000 -c -C 0 -q -T 3 objdump -x win9x.exe`
- Fuzzes the program `objdump` using the sample input `win9x.exe`
- Try 1M seed values (-s) from command line (-c) and keep running if crashed (-C 0) with timeout (-T 3)

```
zzuf [-AcdimnqSvxX] [-s seed|-s start:stop] [-r ratio|-r min:max] [-f fuzzing] [-D delay] [-j jobs] [-C crashes] [-B bytes] [-t seconds] [-T seconds] [-U seconds] [-M mebibytes] [-b ranges] [-p ports] [-P protect] [-R refuse] [-a list] [-l list] [-I include] [-E exclude] [-O omode] [PROGRAM [ARGS]...]
zzuf -h | --help
zzuf -V | --version
```

## ●●● | Mutation-Based Fuzzing

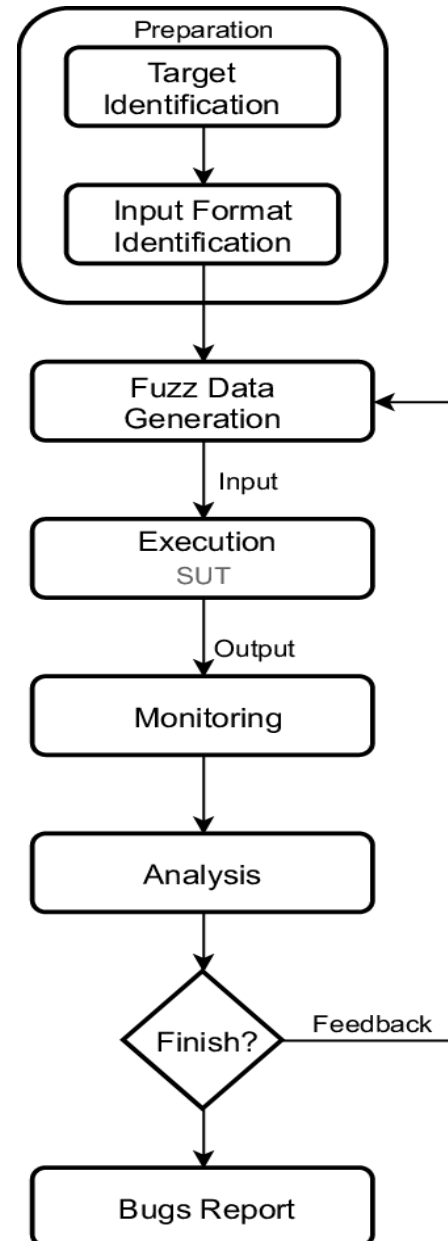
- Easy to setup, and not dependent on program details
- But may be strongly biased by the initial input
- Still prone to some problems
  - May re-run the same path over again (same test)
  - May be very hard to generate inputs for certain paths (checksums, hashes, restrictive conditions)



## ●●● | Generation-Based Fuzzing

- Generational fuzzer generate inputs “from scratch” rather than using an initial input and mutating
- However, require the user to specify a format or protocol spec to start
  - Equivalently, write a generator for generating well-formatted input
- Examples include
  - SPIKE, Peach Fuzz
- However format-aware fuzzing is cumbersome, because you'll need a fuzzer specification for every input format you are fuzzing

# Generation-Based Fuzzing Workflow

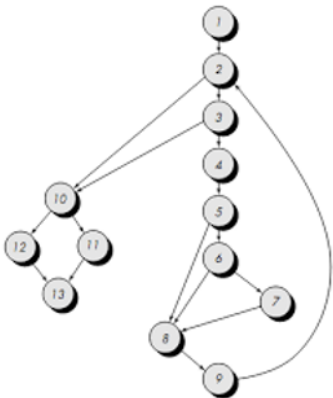


# ●●● | Generation-Based Fuzzing

- Can be more accurate, but at a cost
- **Pros:** More complete search
  - Values more specific to the program operation
  - Can account for dependencies between inputs
- **Cons:** More work
  - Get the specification
  - Write the generator – ad hoc
  - Need to do for each program

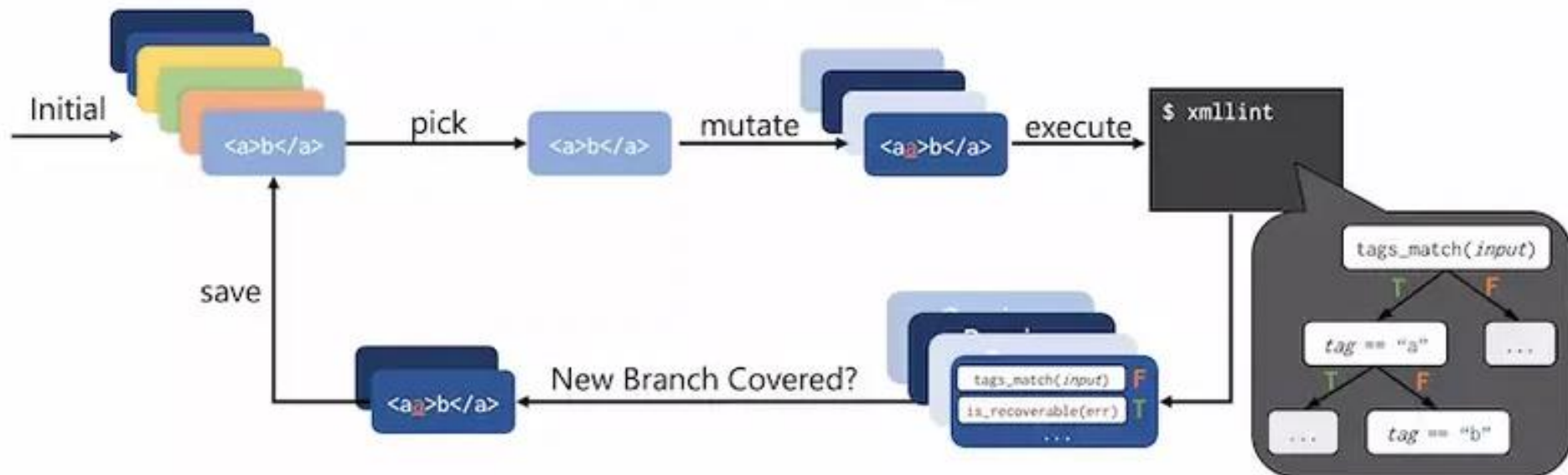
# Coverage-Based Fuzzing

- AKA grey-box fuzzing
- Rather than treating the program as a black box, instrument the program to track coverage
  - E.g., the edges covered
- Maintain a pool of high-quality tests
  - Start with some initial ones specified by users
  - Mutate tests in the pool to generate new tests
  - Run new tests
  - If a new test leads to new coverage (e.g., edges), save the new test to the pool; otherwise, discard the new test



# Coverage-Guided Fuzzing

AFL, libFuzzer, honggfuzz



# ● ● ● | AFL

- Example of coverage-based fuzzing
  - American Fuzzy Lop (AFL)
  - “State of the practice” at this time



# AFL Display

- Tracks the execution of the fuzzer

american fuzzy lop 2.51b (cmpsc497-p1)

<b>process timing</b> run time : 0 days, 2 hrs, 16 min, 32 sec last new path : 0 days, 0 hrs, 13 min, 31 sec last uniq crash : 0 days, 0 hrs, 43 min, 58 sec last uniq hang : none seen yet	<b>overall results</b> cycles done : 0 total paths : 41 uniq crashes : 11 uniq hangs : 0
<b>cycle progress</b> now processing : 3 (7.32%) paths timed out : 0 (0.00%)	<b>map coverage</b> map density : 0.11% / 0.40% count coverage : 1.62 bits/tuple
<b>stage progress</b> now trying : arith 8/8 stage execs : 12.3k/41.9k (29.31%) total execs : 243k exec speed : 30.98/sec (slow!)	<b>findings in depth</b> favored paths : 6 (14.63%) new edges on : 7 (17.07%) total crashes : 2479 (11 unique) total tmouts : 10 (5 unique)
<b>fuzzing strategy yields</b> bit flips : 7/15.4k, 32/15.4k, 0/15.4k byte flips : 0/1929, 0/1926, 0/1920 arithmetics : 8/71.7k, 4/5434, 0/0 known ints : 0/6938, 0/35.5k, 0/56.3k dictionary : 0/0, 0/0, 0/1270 havoc : 0/178, 0/0 trim : 0.00%/930, 0.00%	<b>path geometry</b> levels : 3 pending : 39 pend fav : 5 own finds : 40 imported : n/a stability : 17.69%

[cpu000: 19%]

- Key information are
  - “total paths” – number of different execution paths tried
  - “unique crashes” – number of unique crash locations

## ● ● ● | Grey Box Fuzzing

- Finds flaws, but still does not understand the program
- **Pros:** Much better than black box testing
  - Essentially no configuration
  - Lots of crashes have been identified
- **Cons:** Still a bit of a stab in the dark
  - May not be able to execute some paths
  - Searches for inputs independently from the program
- Need to improve the effectiveness further



## ••• | White Box Fuzzing

- Combines **test generation** with fuzzing
  - Test generation based on static analysis and/or symbolic execution – more later
  - Rather than generating new inputs and hoping that they enable a new path to be executed, compute inputs that will execute a desired path
    - And use them as fuzzing inputs
- Goal: Given a sequential program with a set of input parameters, generate a set of inputs that maximizes code coverage

# Symbolic Execution

# Symbolic execution

**1976:** *A system to generate test data and symbolically execute programs* (Lori Clarke)

**1976:** *Symbolic execution and program testing* (James King)

**2005-present:** practical symbolic execution

- Using SMT solvers
- Heuristics to control exponential explosion
- Heap modeling and reasoning about pointers
- Environment modeling
- Dealing with solver limitations

# Classic symbolic execution

```
def f (x, y):  
    if (x > y):  
        x = x + y  
        y = x - y  
        x = x - y  
        if (x - y > 0):  
            assert false  
    return (x, y)
```

# Classic symbolic execution

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def f (x, y):  
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Execute the program on *symbolic values*.

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```

$x \mapsto X$   
 $y \mapsto Y$

Execute the program on *symbolic values*.  
*Symbolic state* maps variables to symbolic values.

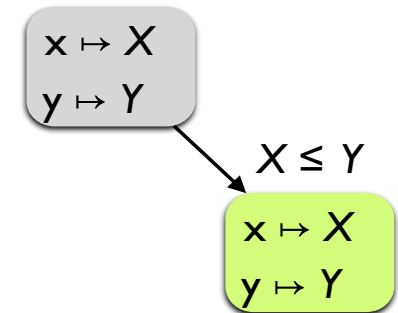
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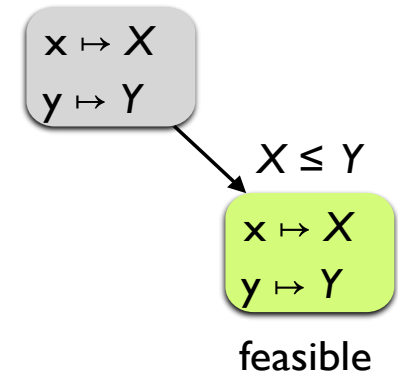
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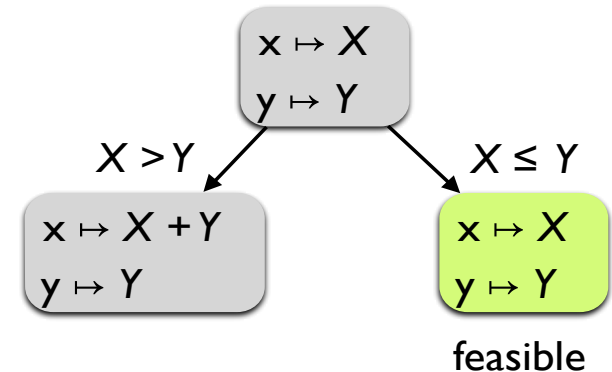
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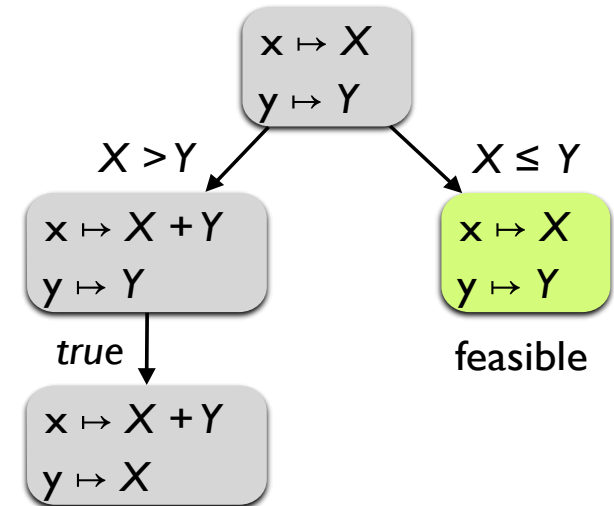
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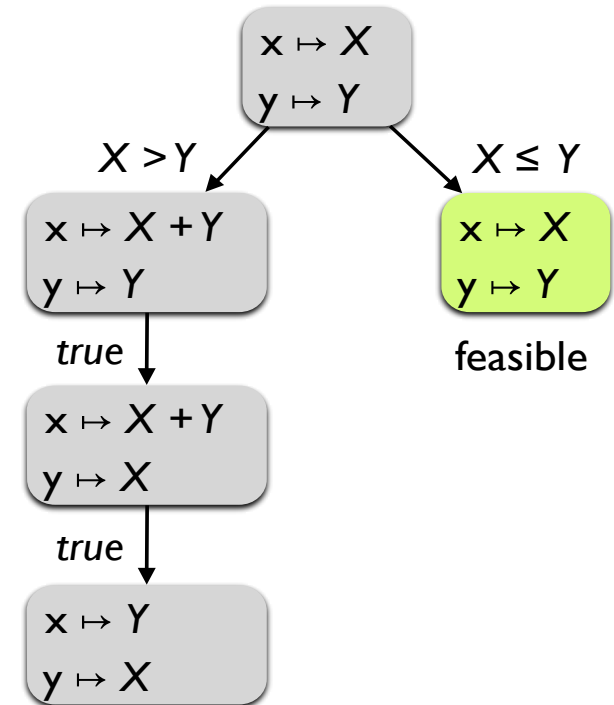
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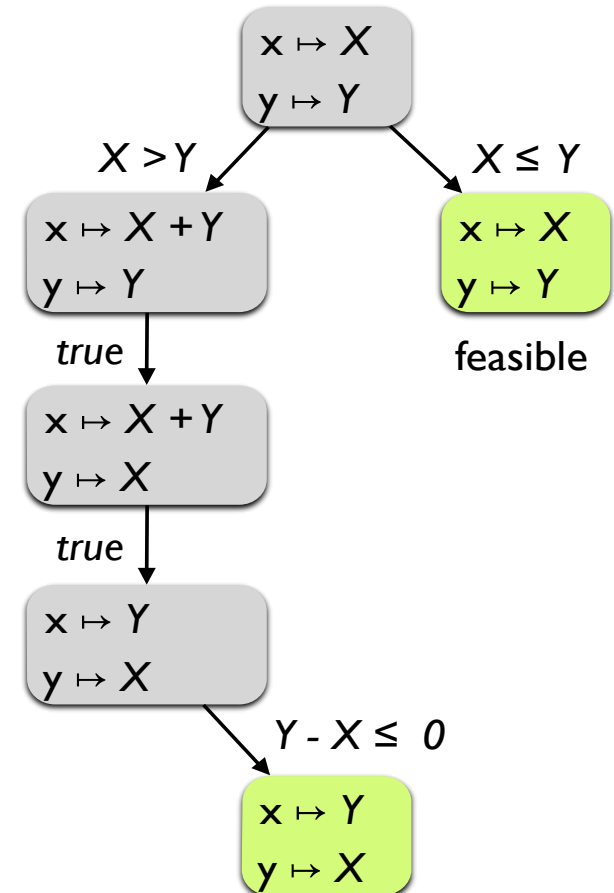
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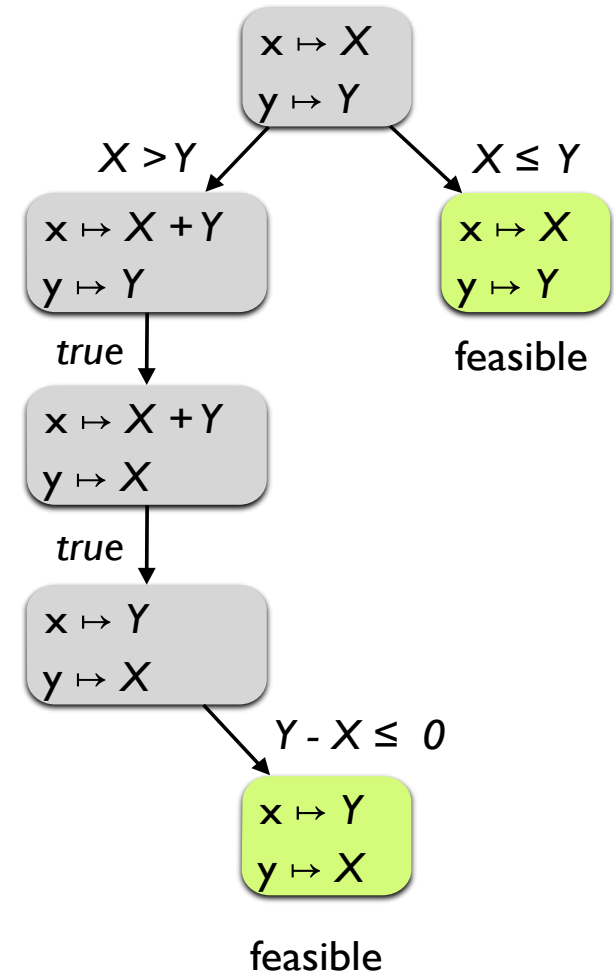
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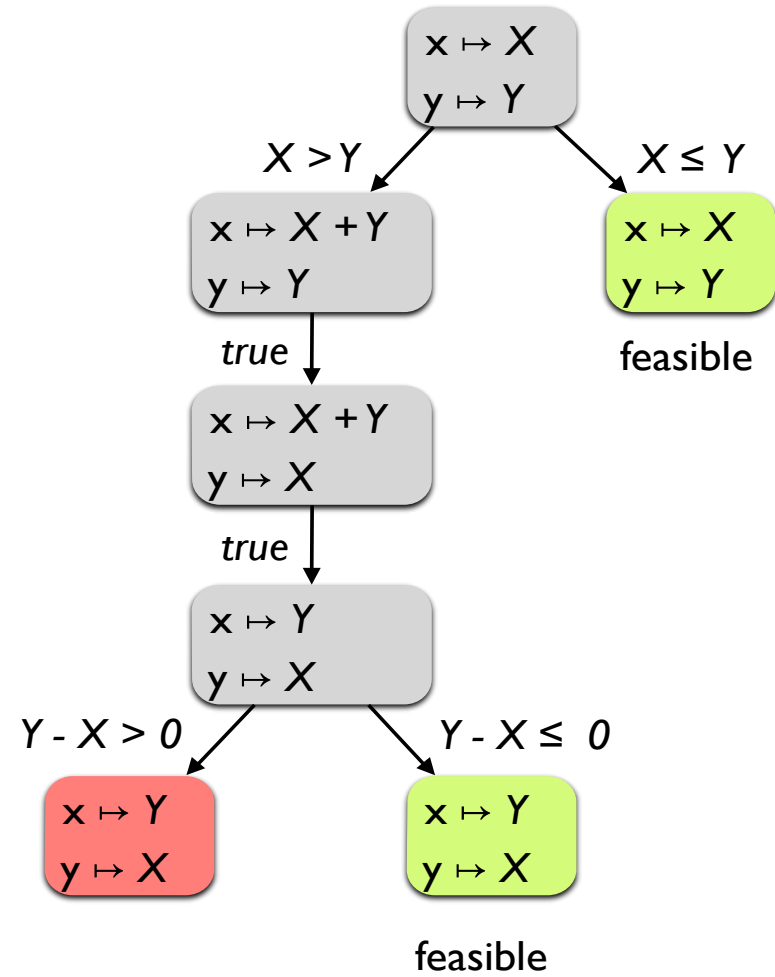
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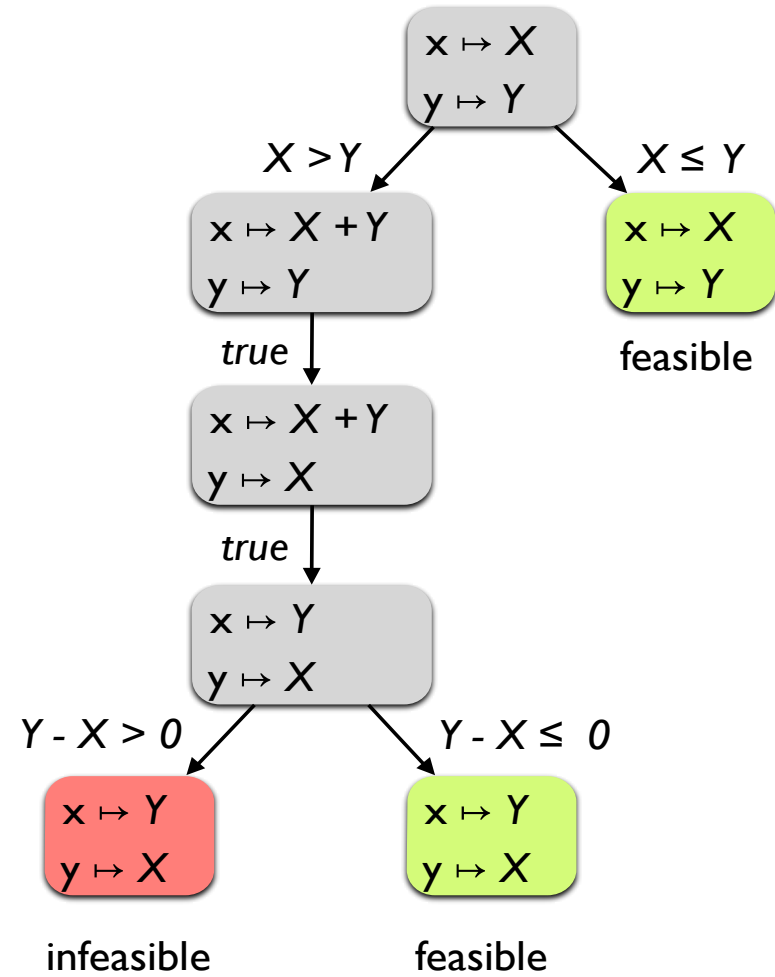
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# Another symbolic execution example

```
1. int a =  $\alpha$ , b =  $\beta$ , c =  $\gamma$ ;  
2.           // symbolic  
3. int x = 0, y = 0, z = 0;  
4. if (a) {  
5.   x = -2;  
6. }  
7. if (b < 5) {  
8.   if (!a && c) { y = 1; }  
9.   z = 2;  
10.}  
11.assert(x+y+z!=3)
```

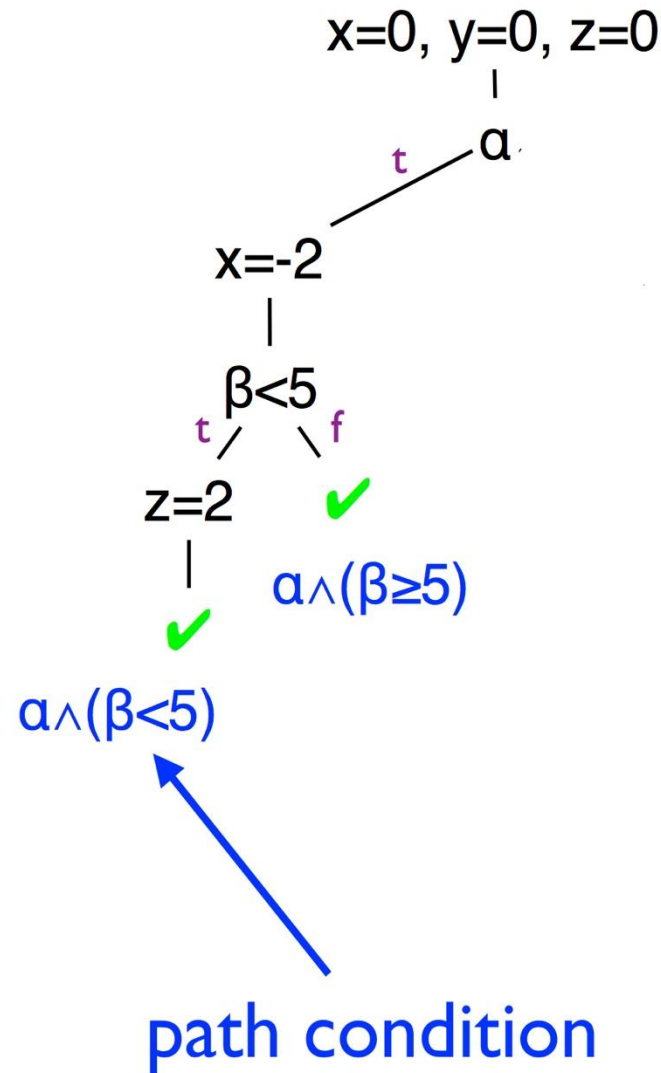


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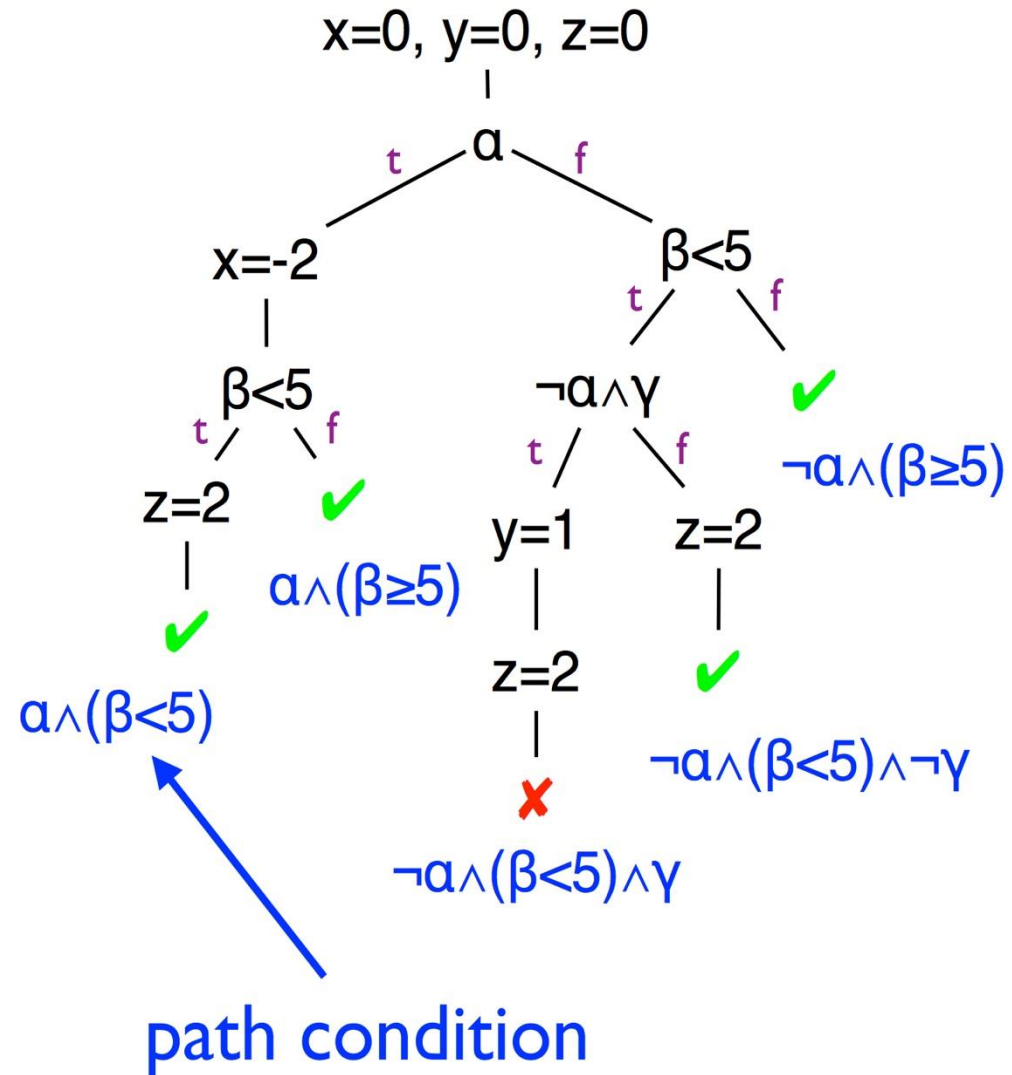
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# Another symbolic execution example

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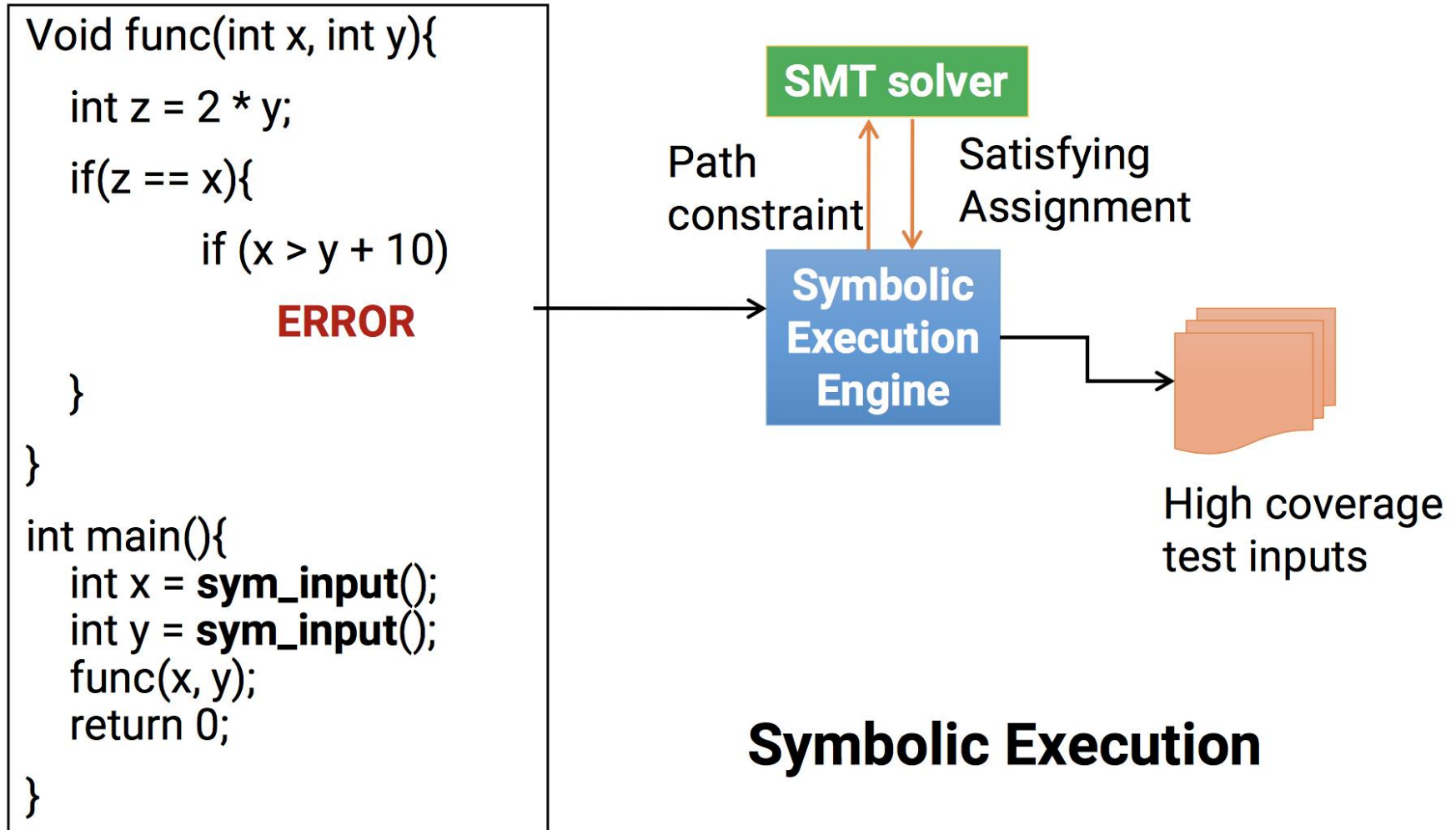
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```



# What's going on here?

- During symbolic execution, we are trying to determine if certain formulas are satisfiable
  - E.g., is a particular program point reachable?
    - Figure out if the path condition is satisfiable
  - E.g., is array access  $a[i]$  out of bounds?
    - Figure out if **conjunction of path condition and  $i < 0 \vee i > a.length$**  is satisfiable
  - E.g., generate concrete inputs that execute the same paths
- This is enabled by powerful SMT/SAT solvers
  - SAT = Satisfiability
  - SMT = Satisfiability modulo theory = SAT++
  - E.g. Z3, Yices, STP

# Symbolic execution for software testing



# How does symbolic execution work?

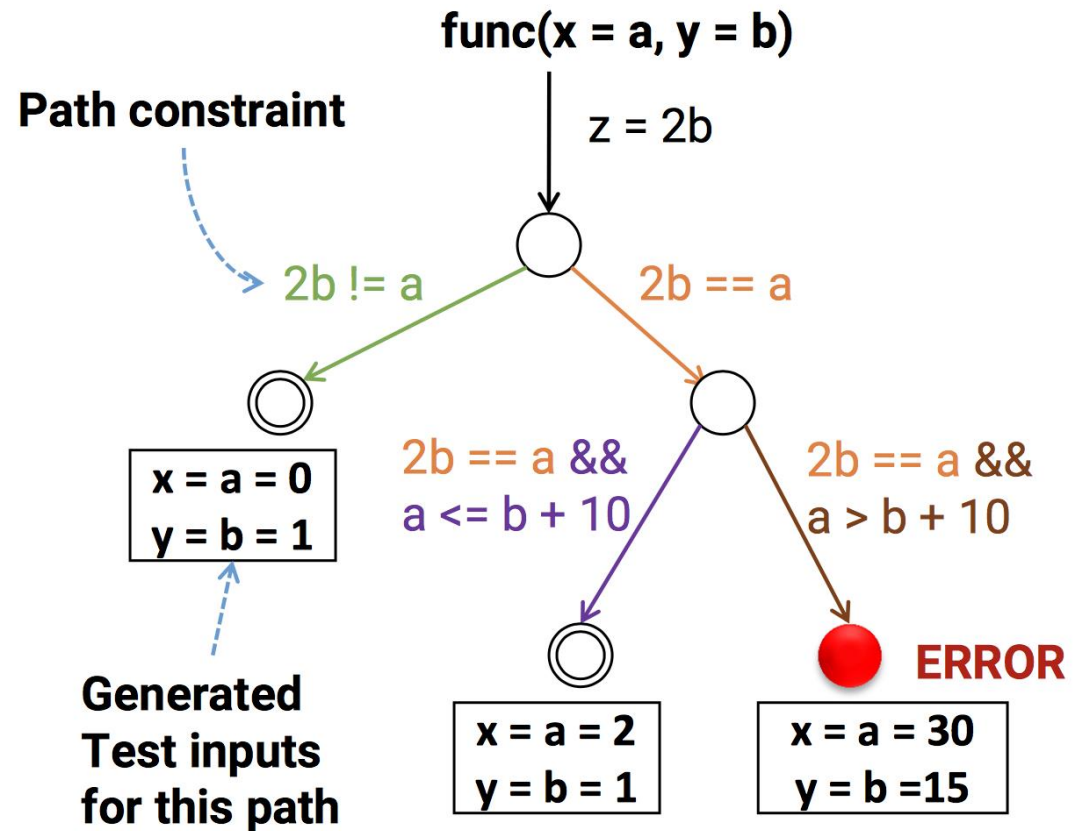
```

Void func(int x, int y){
    int z = 2 * y;
    if(z == x){
        if (x > y + 10)
            ERROR
    }
}

int main(){
    int x = sym_input();
    int y = sym_input();
    func(x, y);
    return 0;
}

```

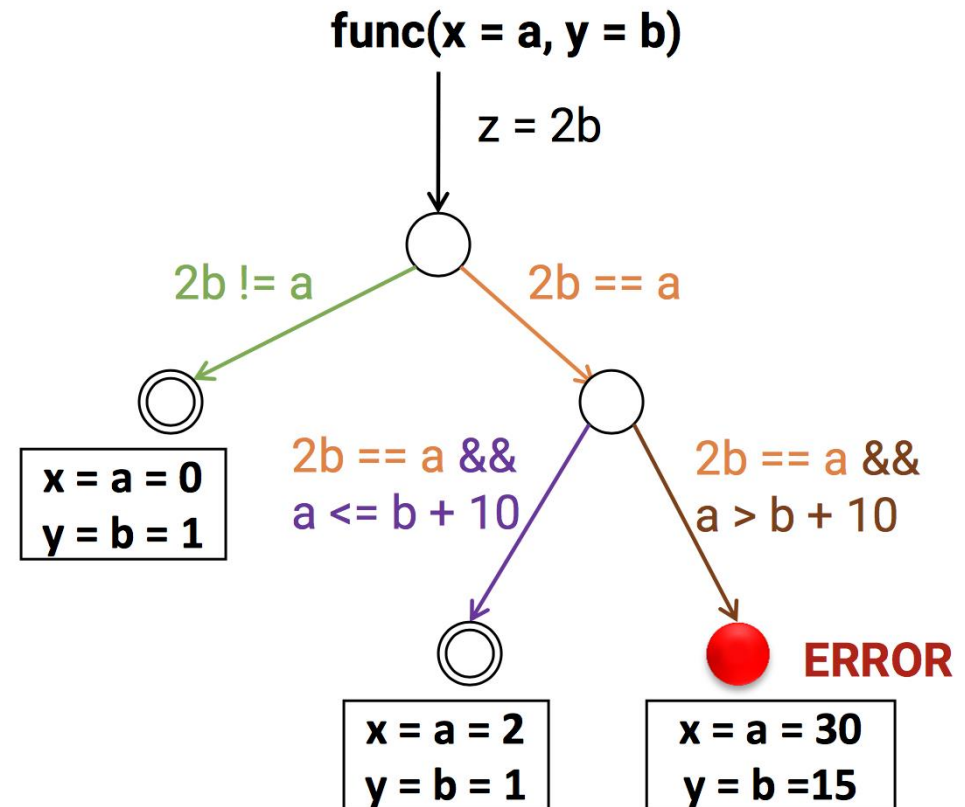
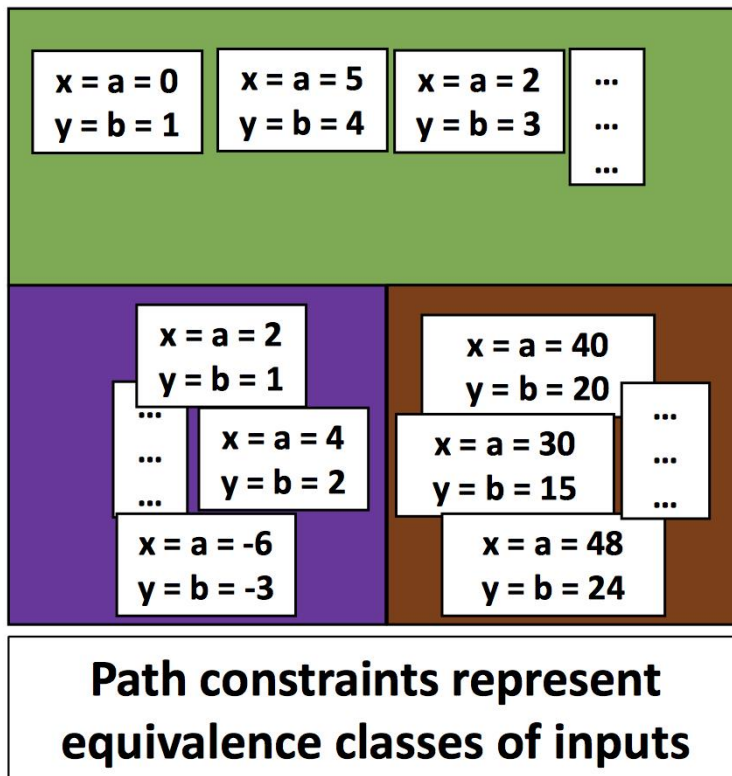
## How does symbolic execution work?



**Note: Require inputs to be marked as symbolic**

# Equivalence classes of inputs

## How does symbolic execution work?



# Concolic (Concrete + Symbolic) Execution

# Fuzz (Random) Testing

- Very low probability of reaching an error
- Problematic for complex data structures

```
Example ( ) {  
    s = readString();  
    if (s[0]=='I' && s[1]=='C' &&  
        s[2]=='S' && s[3]=='E' &&  
        s[4]=='2' && s[5]=='0' &&  
        s[6]=='0' && s[7]=='7') {  
        printf("Am I here?");  
    }  
}
```

Input domain = {'0', '2', '7', 'C', 'E', 'I', 'S'}  
Probability of reaching printf =  $7^{-8} \gg 10^{-7}$

Fast and Inexpensive



# Concolic Testing

- Combine **concrete testing** (concrete execution) and **symbolic testing** (symbolic execution)

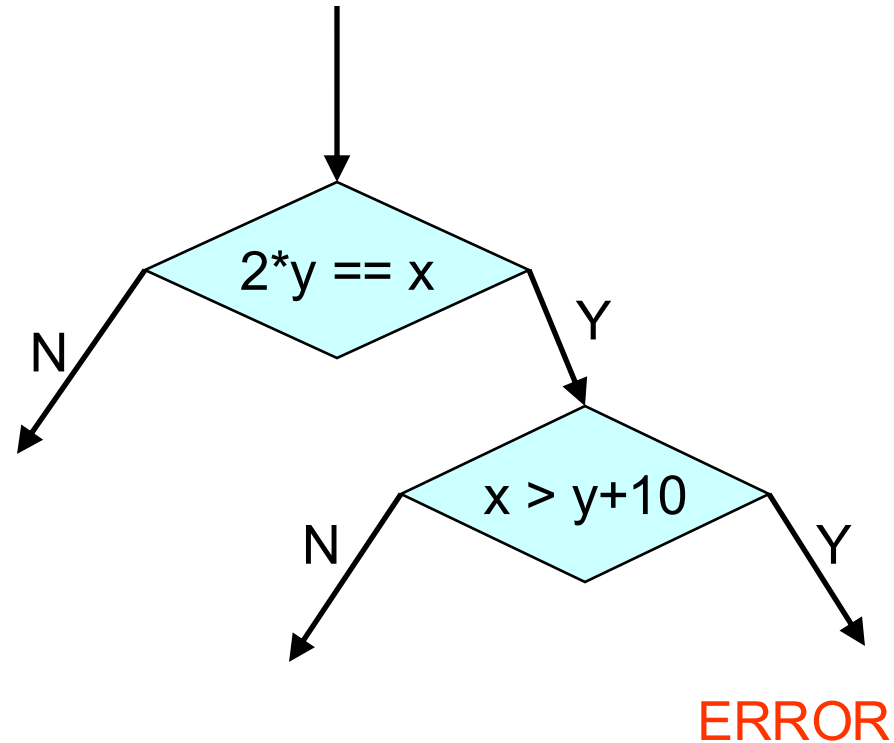
**Concrete** + **Symb**olic**** = **Concolic**

# Example

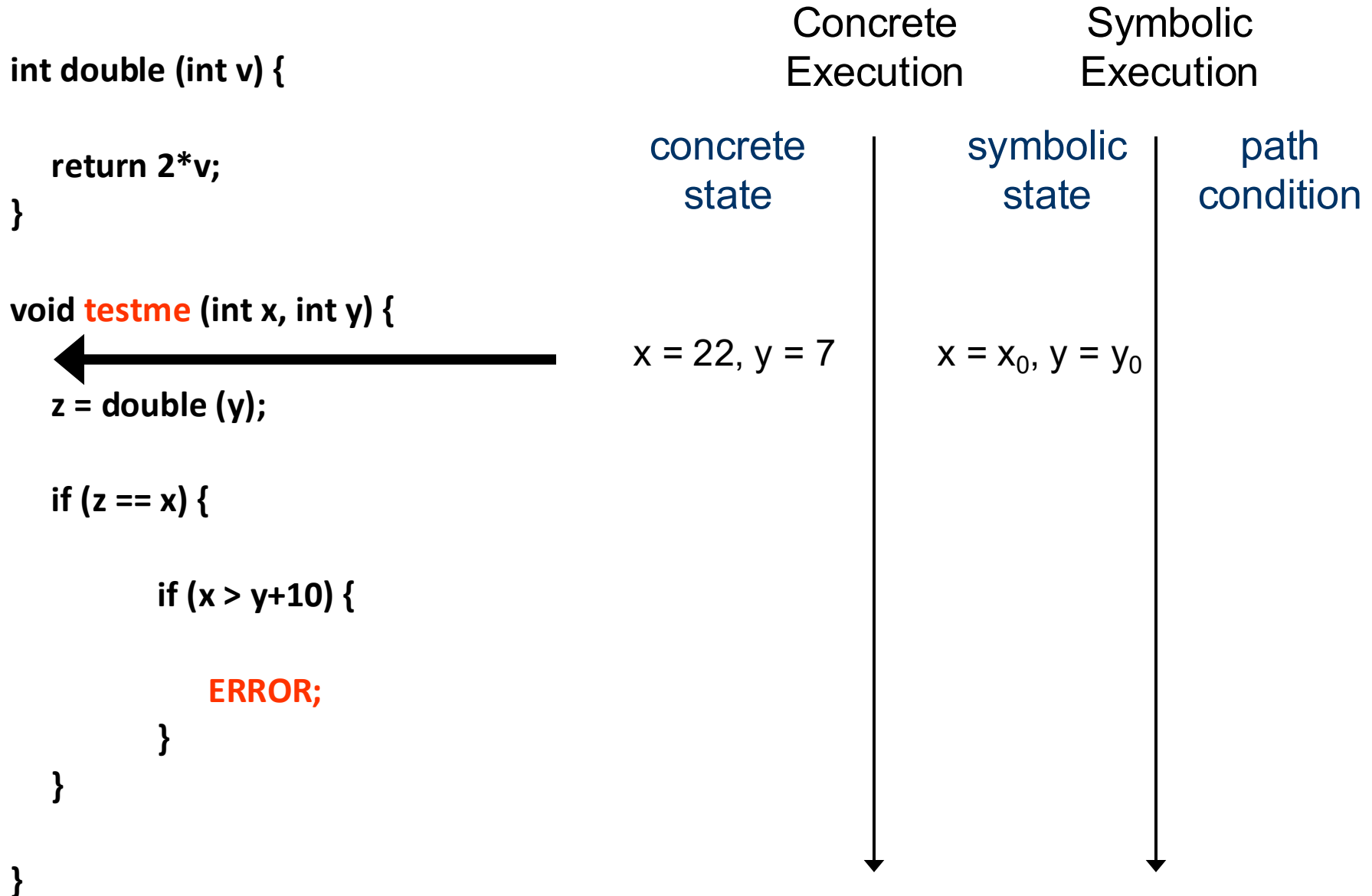
```
int double (int v) {  
  
    return 2*v;  
}  
  
void testme (int x, int y) {  
  
    z = double (y);  
  
    if (z == x) {  
  
        if (x > y+10) {  
  
            ERROR;  
        }  
    }  
}
```

# Example

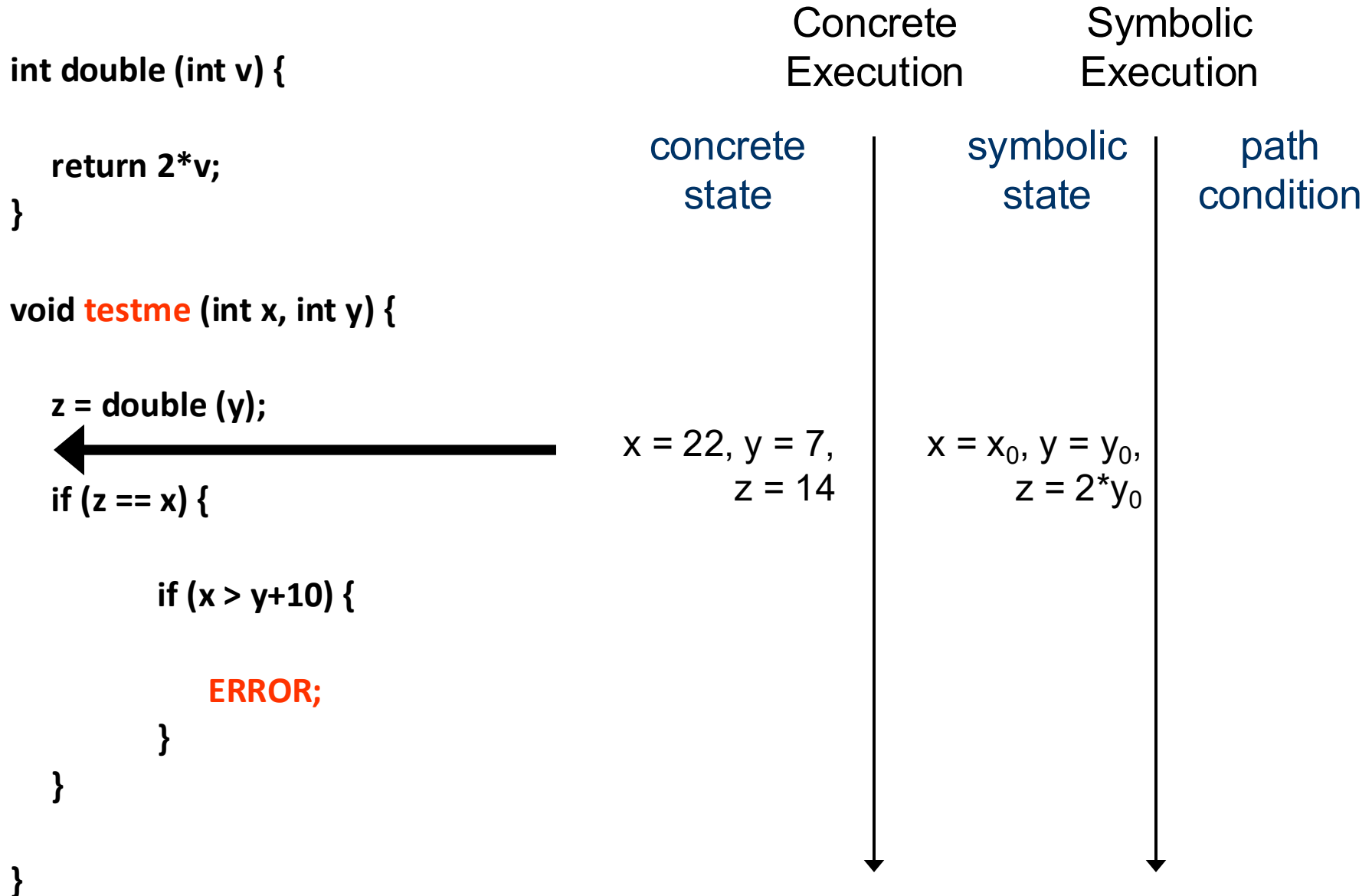
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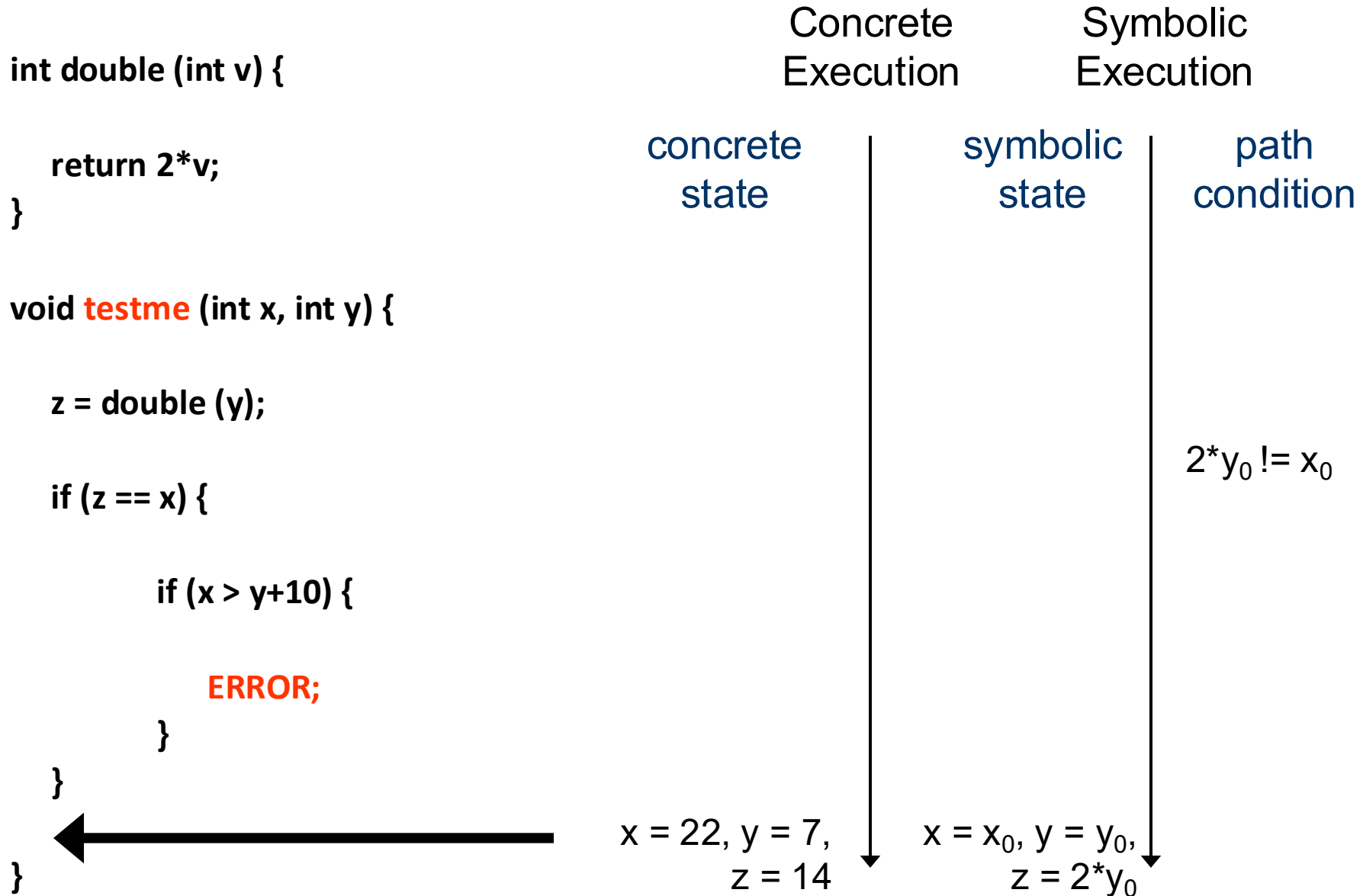
# Concolic Testing Approach



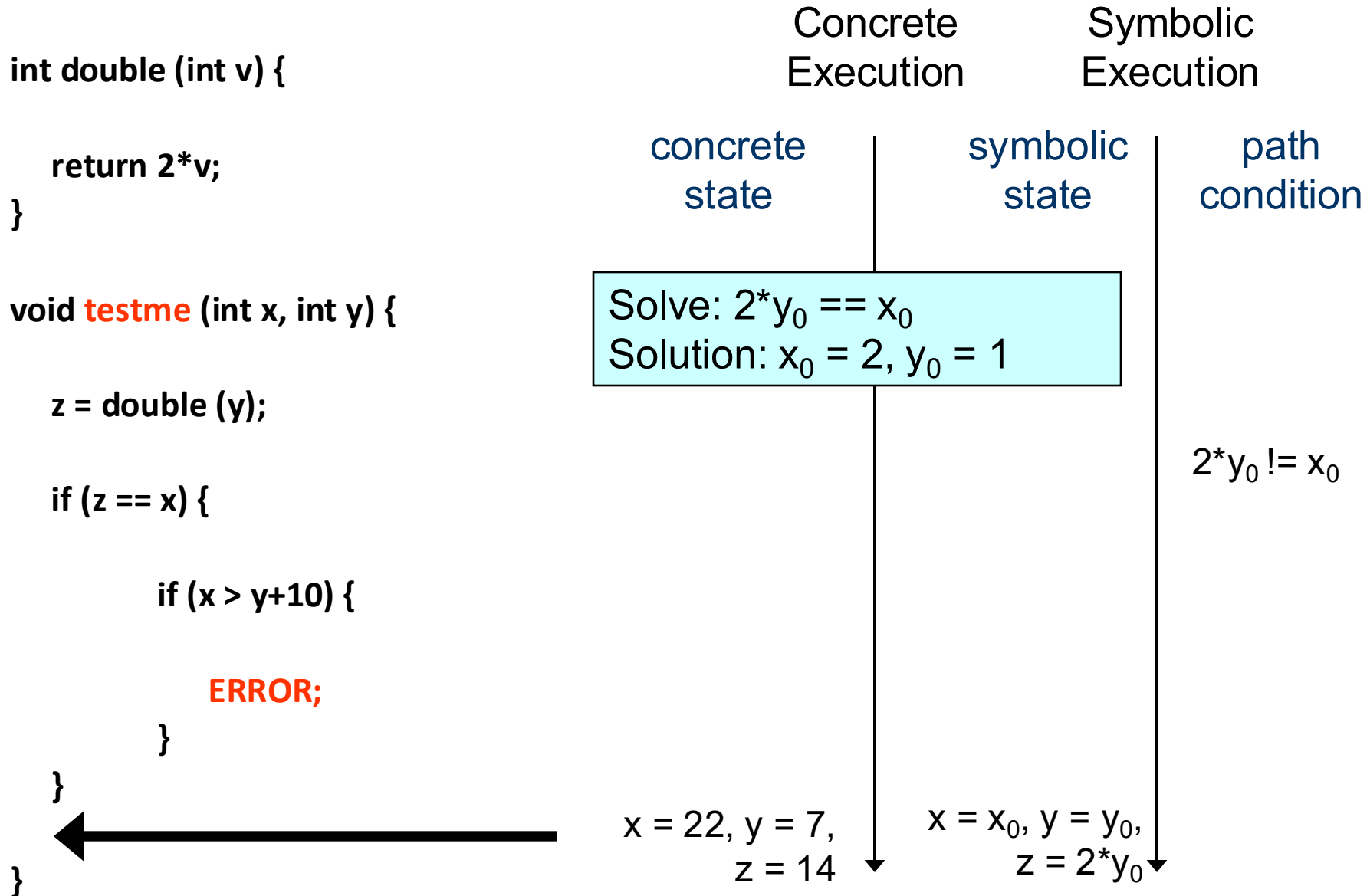
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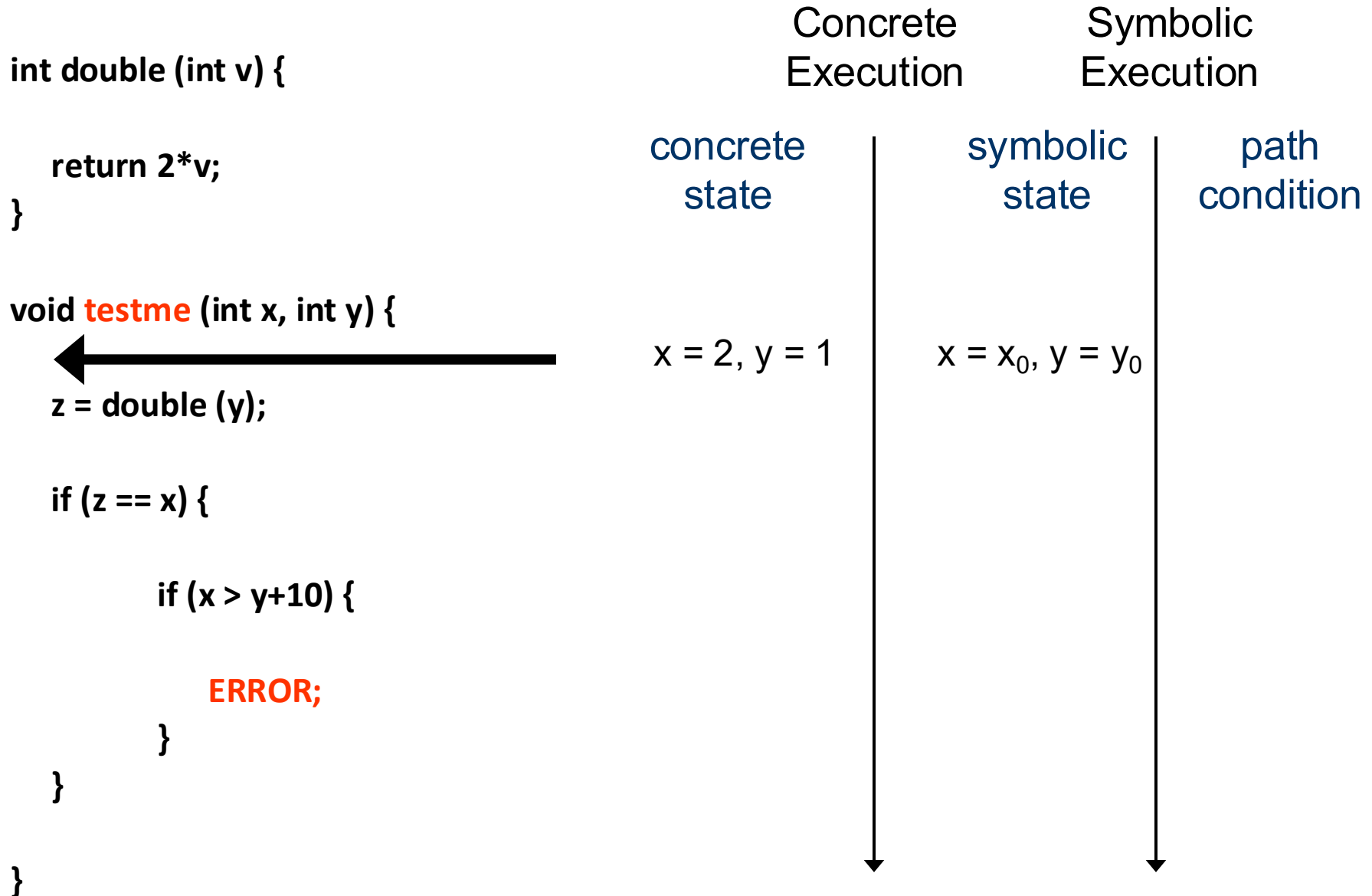
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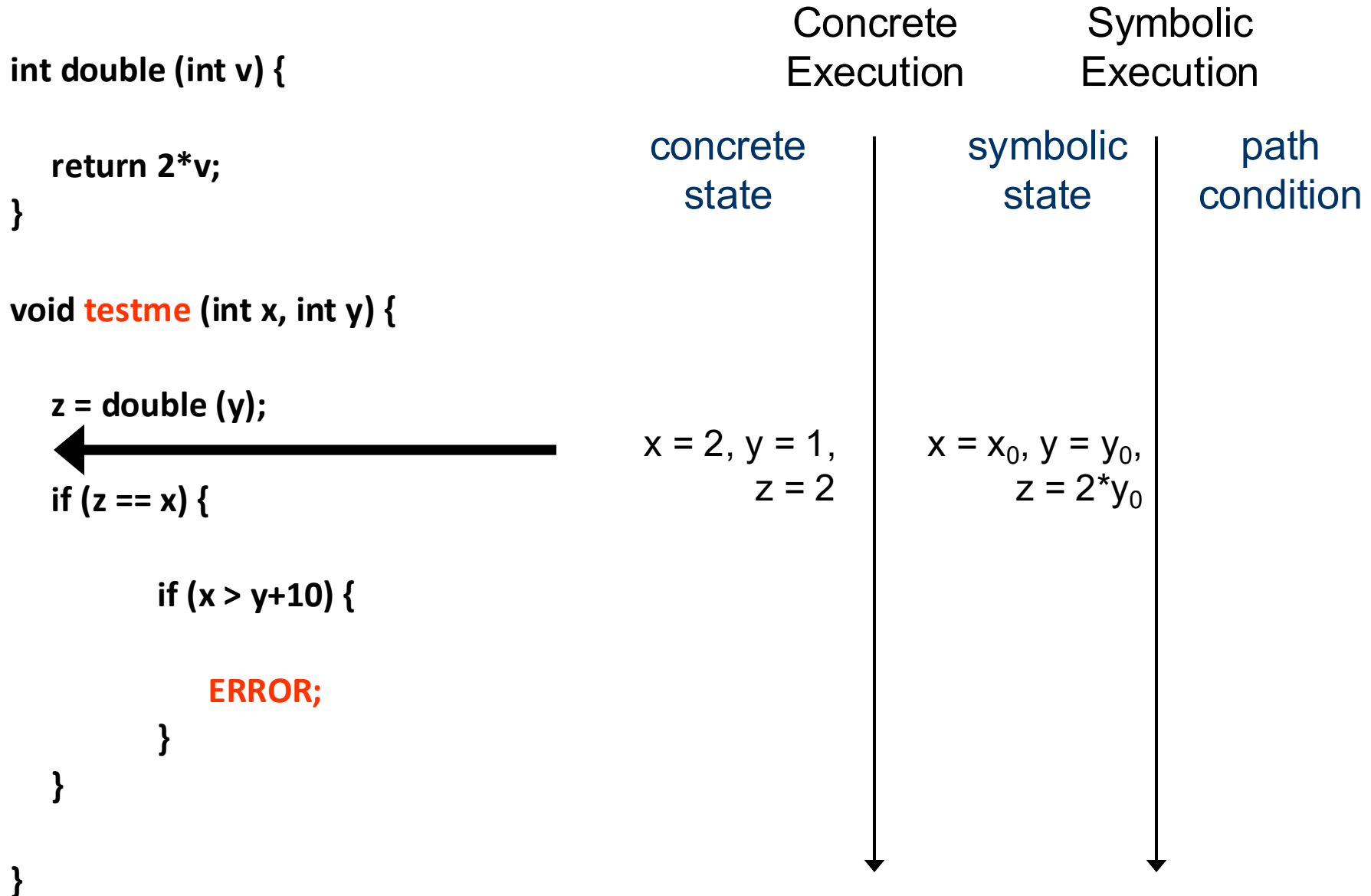


# Concolic Testing Approach





# Concolic Testing Approach



# Symbolic Execution

## path condition

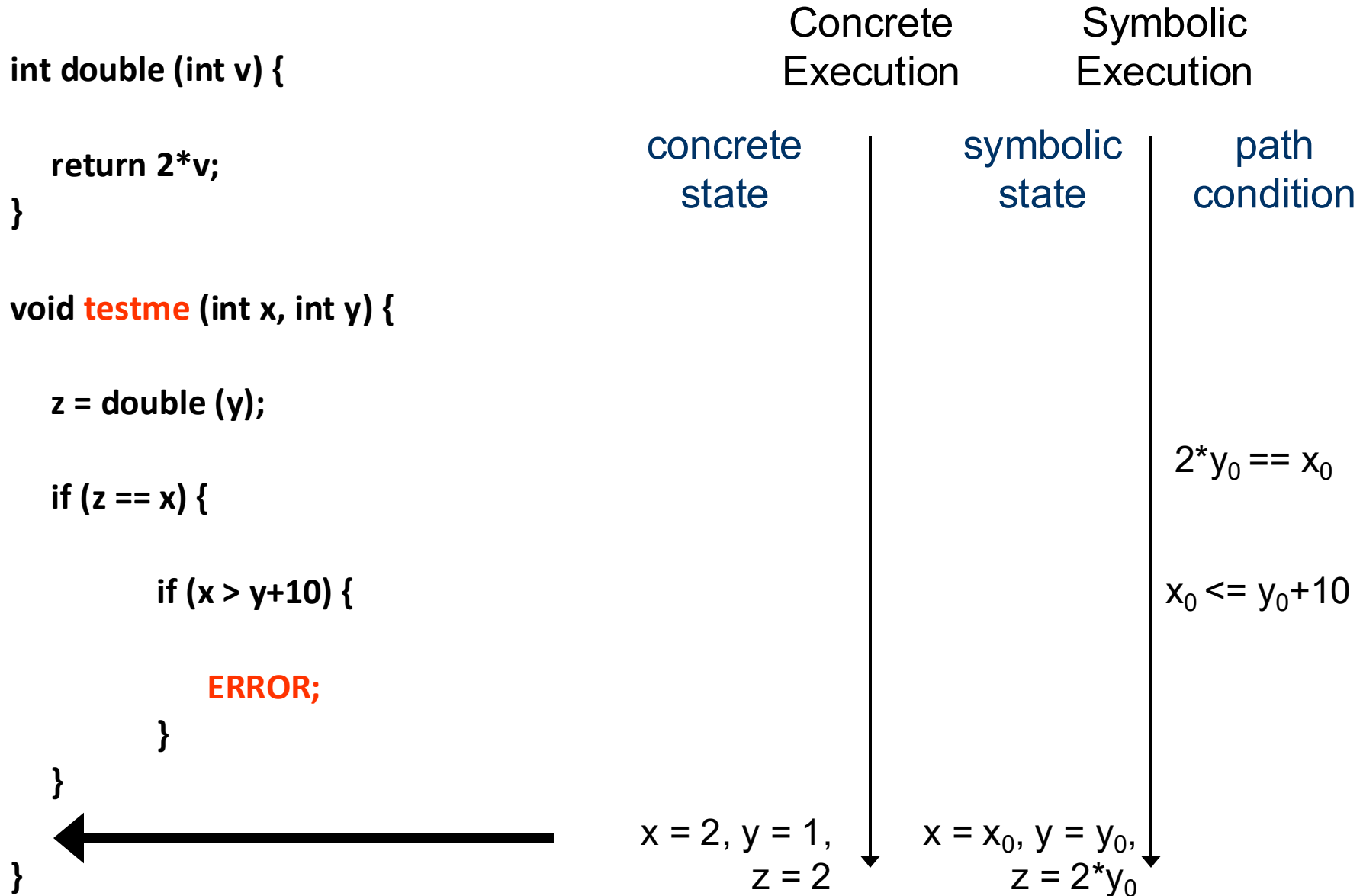
$$\begin{aligned}x &= x_0, \quad y = y_0, \\z &= 2^*y_0\end{aligned}$$
$$2^*y_0 == x_0$$

```
if (x > y+10) {
```

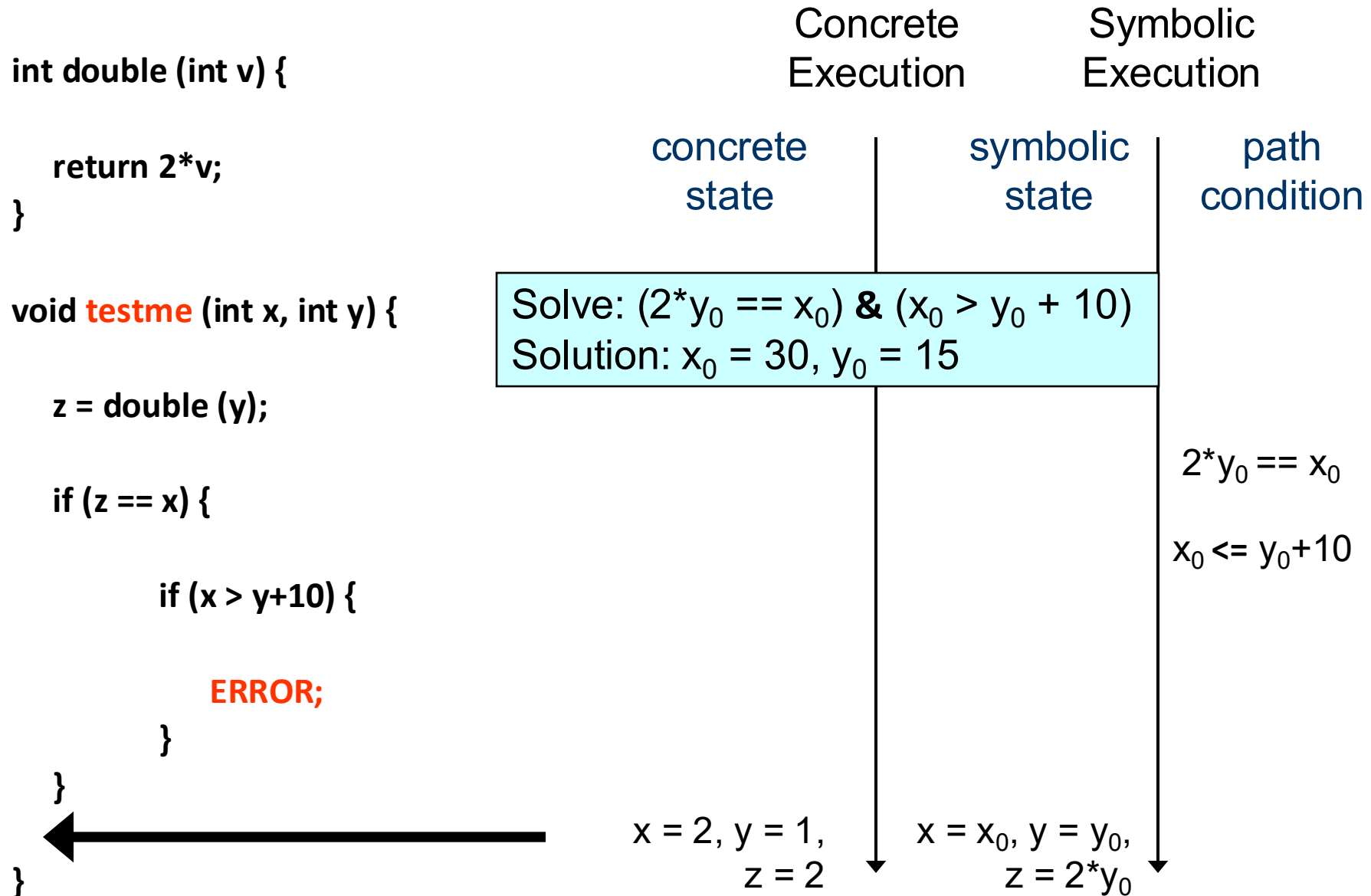
# ERROR;

}

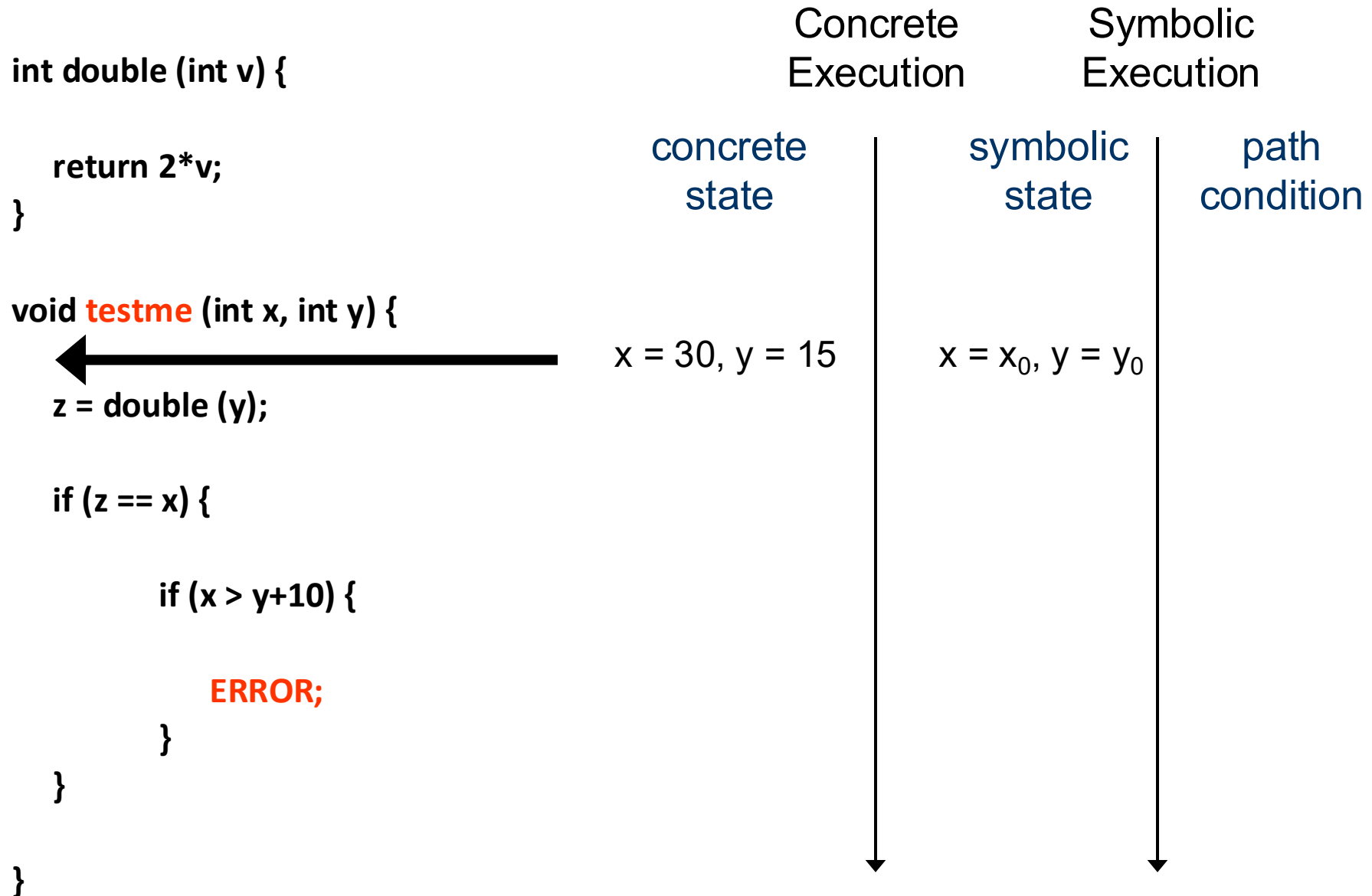
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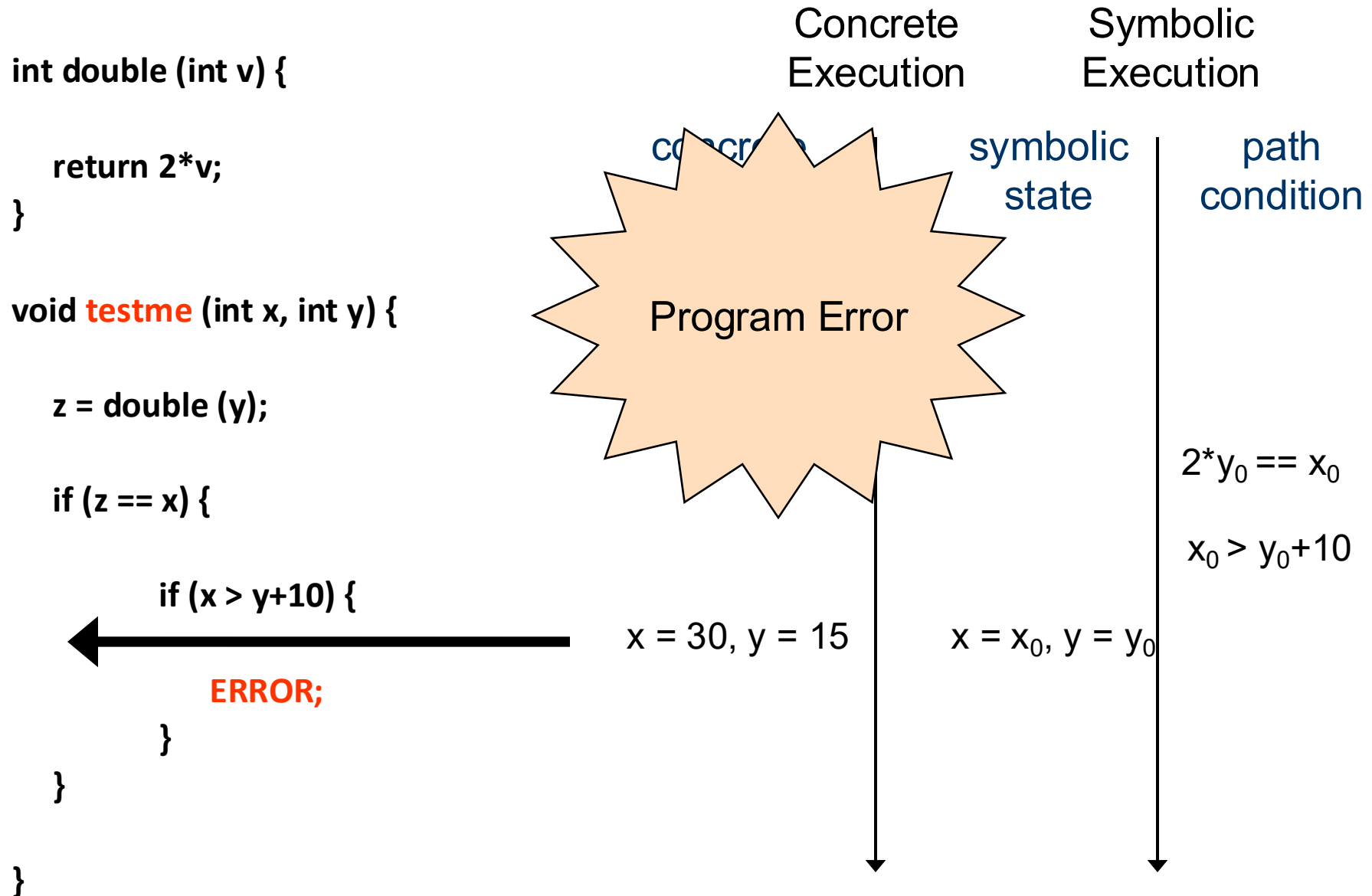
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# Concolic Testing Approach



*End of Lecture 11*