

# CS1632, Lecture 19: Static Analysis, Part 3

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# State Space Reduction Techniques

- State collapsing
  - Heuristic state approximation
  - Hash compaction
  - Heap canonicalization
  - Symbolic execution
- 
- So far we have been looking at *enumerative* model checking
    - Enumerating all the states a program can be in
      - No matter how hard you try, no escaping state explosion
  - Symbolic execution can fundamentally change the equation

# Let's take a step back

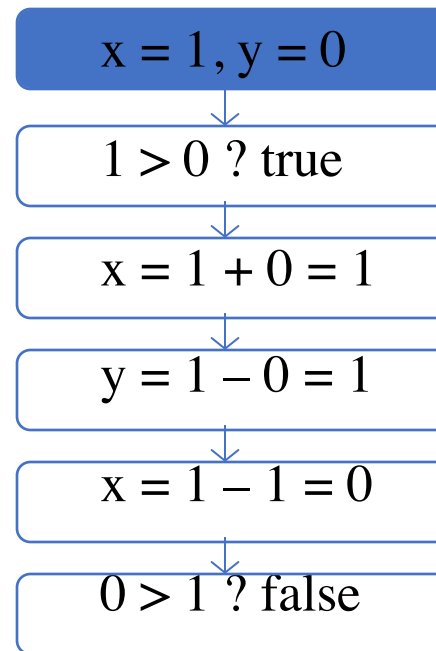
- Model checking
  - Abstract model checking: models abstraction of states
  - Concrete model checking: models actual program states
    - ☛ We are here
- Concrete model checking can be subdivided into:
  - Concrete *enumerative* model checking
  - Concrete *symbolic* model checking
    - Concrete model checking using symbolic execution
      - ☛ And we are here

# Example: Enumerative Model Checking

Code that swaps 2 integers

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```

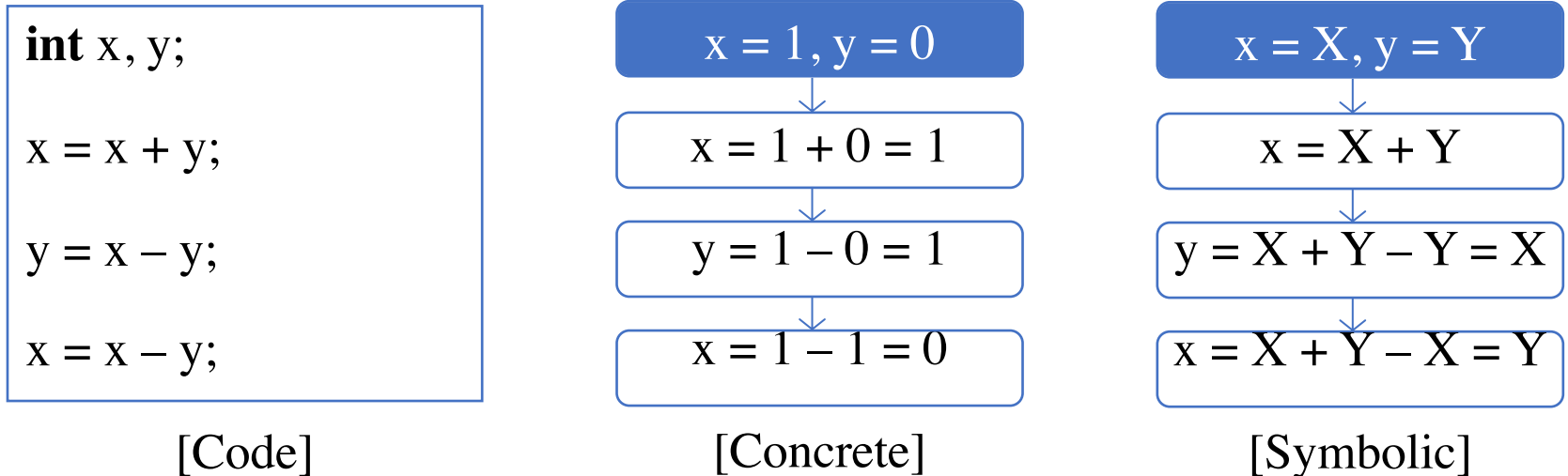
Execution Path for  $x=1, y=0$



- Must do this for all values of  $x$  and  $y$ .
- But is that how a human would do it?

# Symbolic Model Checking

- Trace through a program like a human being would
- In a symbolic execution:
  - Inputs are *symbolic values* instead of concrete data values
  - Variables are *symbolic expressions* on the *symbolic values*
- Example:



• Symbolic execution proves that the swap works for all X and Y!

# Symbolic Model Checking

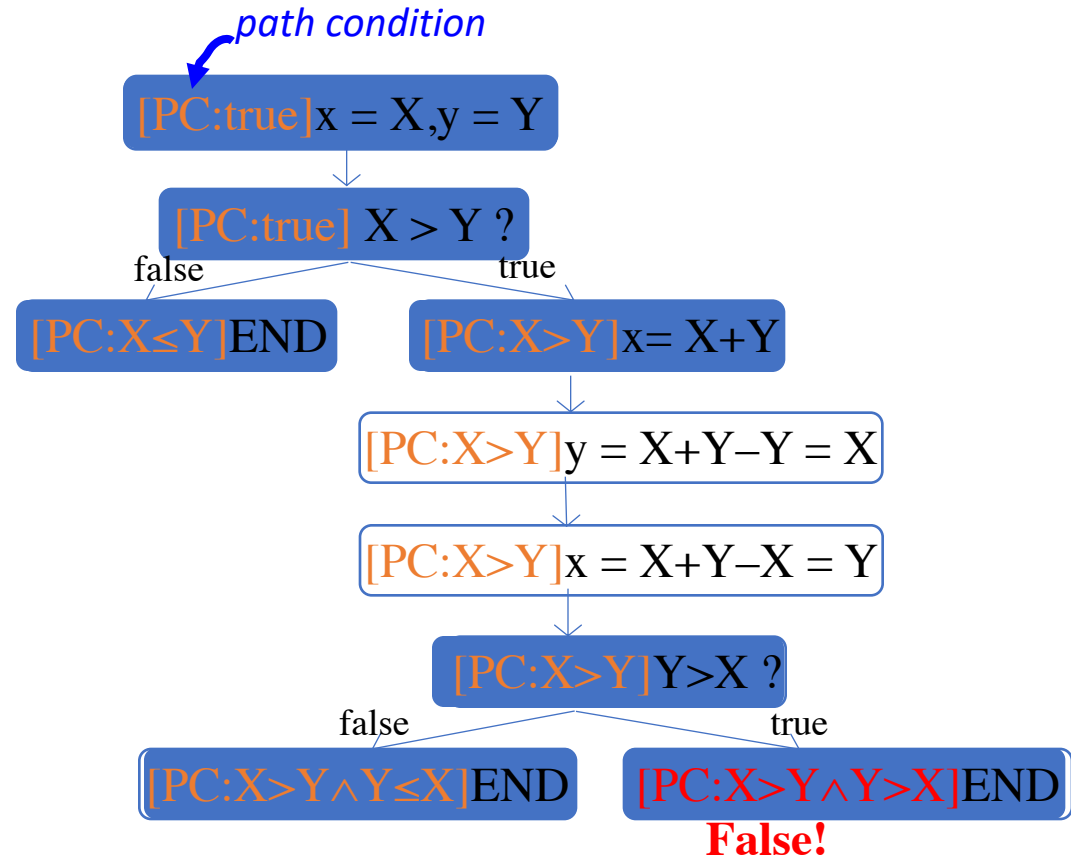
- What if there is path divergence?
  - if statement
  - for loop
  - while loop
- For each path, build a **Path Condition (PC)**
  - Condition on symbolic values (the Xs and the Ys)

# Example: Symbolic Execution

Code that swaps 2 integers:

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```

Symbolic Execution Tree:



# Is the Path Condition Feasible?

- Each path condition is checked using a constraint solver



- If path is infeasible, does not continue down that path
  - Hence, **assert false** is never reached

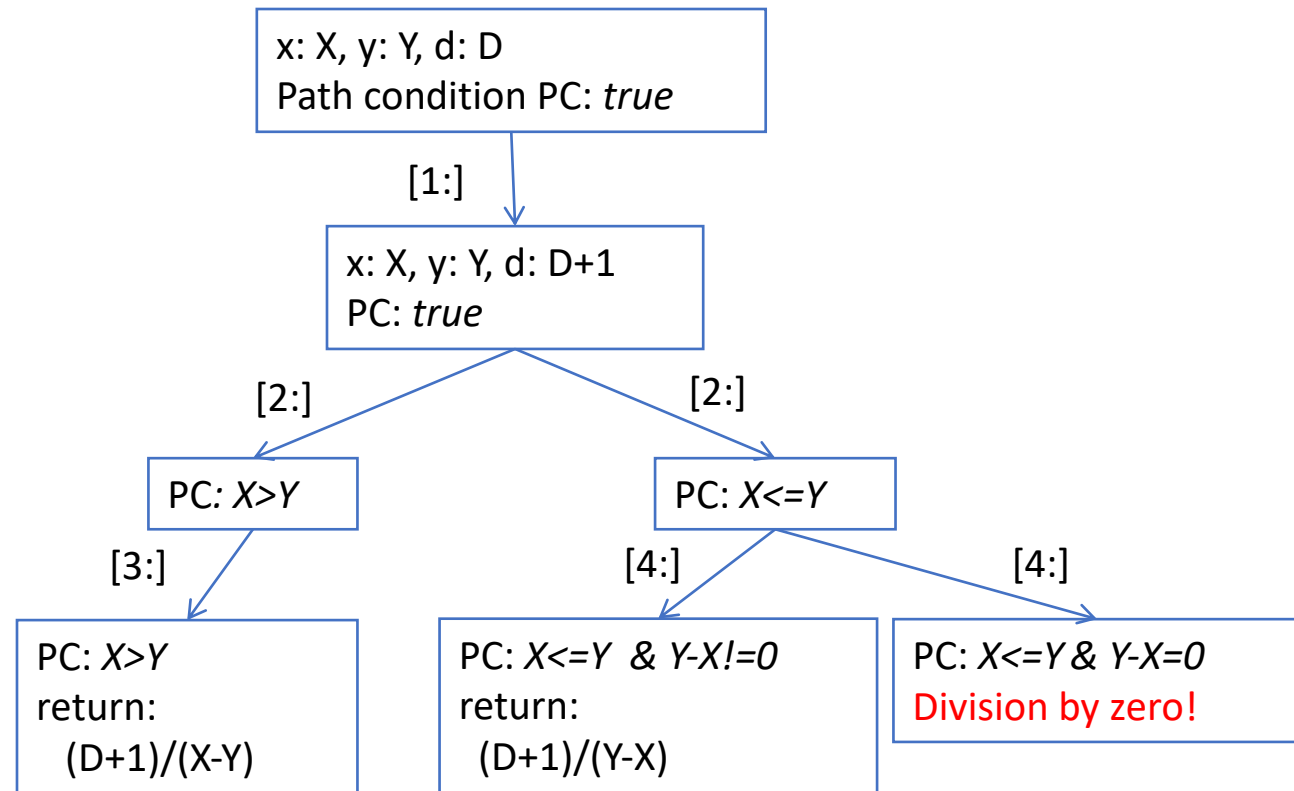


# Symbolic Model Checking Uses

- Prove a program correct
  - Much less state explosion than enumerative checking
  - Now proving correctness suddenly becomes feasible
- Generate test cases
  - Generate input values that trigger a defect
  - Input values can be generated out of path conditions
- Generate program invariants
  - Invariants enhance programmer's understanding of code
  - Invariants can also be generated out of path conditions

# Generating Test Cases out of Path Conditions

Symbolic execution tree:



*Solve path conditions → test inputs*

# Auto-generated JUnit Tests

```
@Test public void t1() {    Pass    PC:  $X > Y$   
    m(1, 0, 1);  
}  
@Test public void t2() {    Pass    PC:  $X \leq Y \ \& \ Y - X \neq 0 \Leftrightarrow X < Y$   
    m(0, 1, 1);  
}  
@Test public void t3() {    Fail X    PC:  $X \leq Y \ \& \ Y - X = 0 \Leftrightarrow X = Y$   
    m(1, 1, 1);  
}
```

☛ Achieves full path coverage

# Generating Invariants out of Path Conditions

- Pre-condition:
  - “ $x \neq y$ ”
- Post-condition:
  - “ $\text{result} = ((x > y) ? (d+1)/(x-y) : (d+1)/(y-x))$ ”
- Each method can be annotated with invariants
  - Can be checked against specifications for defects
  - Can enhance programmer’s understanding of method

# Symbolic Model Checking Challenges

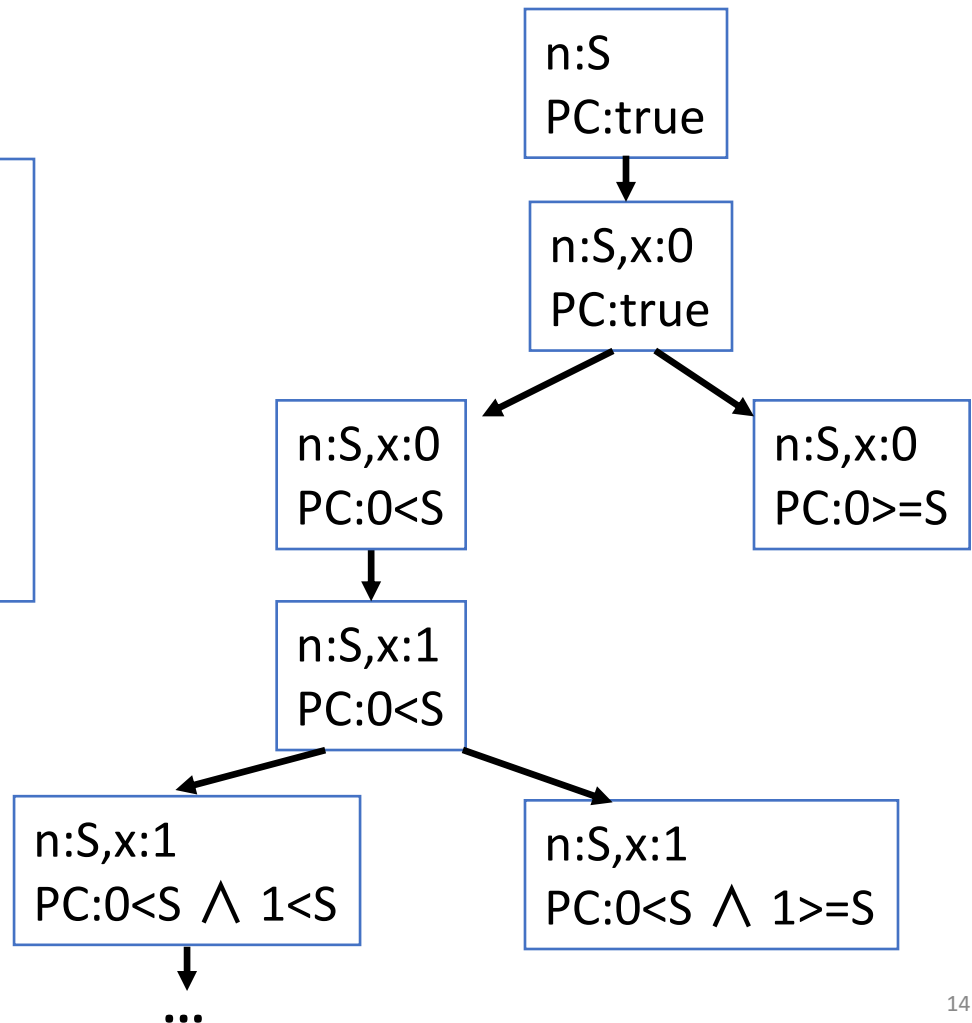
- Symbolic model checking does have challenges
- ... Or every one would be using symbolic model checking
- Some examples are:
  - Loops
  - Complex math constraints
  - Complex data structures

# Challenges: Loops

## Example Code

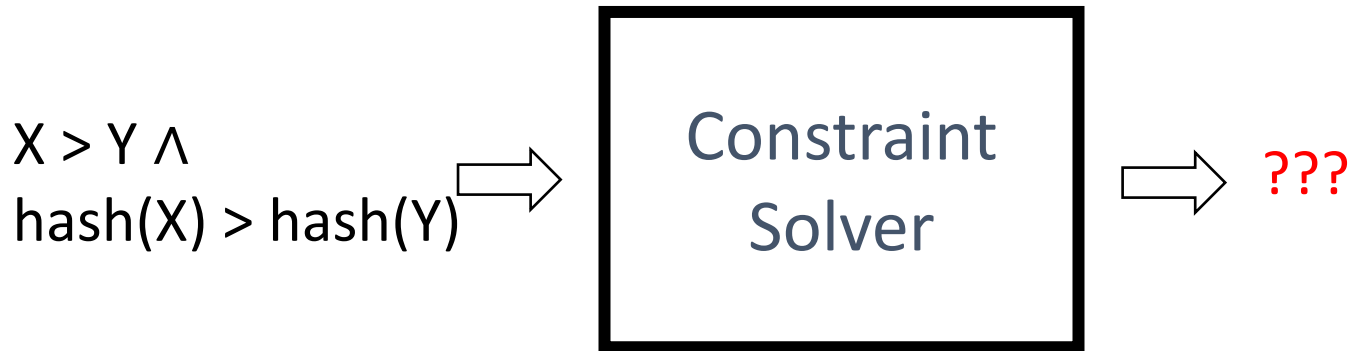
```
void test(int n) {  
  int x = 0;  
  while(x < n) {  
    x = x + 1;  
  }  
}
```

## Infinite symbolic execution tree



# Challenges: Complex Math Constraints

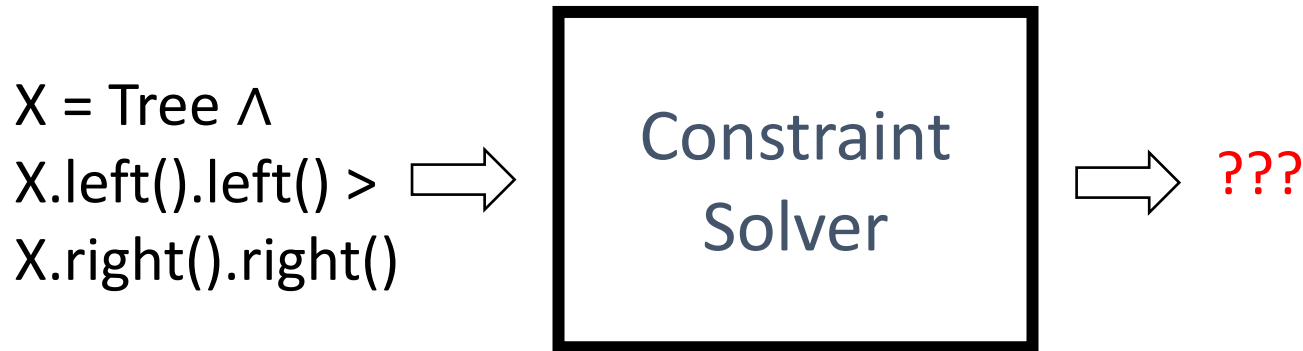
- Constraint solvers are not particularly good at math



- If above constraint was an if condition:  
if  $(X > Y \wedge \text{hash}(X) < \text{hash}(Y))$  assert false;  
    ☛ Will have a hard time checking whether assert fires

# Challenges: Complex Data Structures

- Complex data structures are confusing to solvers



- In order to solve above constraint, solver must know:
  - What a tree data structures looks like
  - What `left()` means and what `right()` means
- Solvers know some data structures, but not many



# The Best of Both Worlds

- Symbolic Model Checking (Symbolic Execution)
  - + Much less state explosion
  - Hard time dealing with loops, math, data structures
- Enumerative Model Checking (Concrete Execution)
  - Serious state explosion
  - + No problems with loops, math, data structures  
(just execute the loop, math, or data structure code)
- The best of both worlds: Concolic Execution
  - Concolic = Concrete + Symbolic
  - a.k.a. DART(Directed Automated Random Testing)

# Automated Random Testing

- Where have I heard that before? Hmm...
- Stochastic Testing is an automated random test
  - Randomly selects values to check given property
- Fuzz Testing is also an automated random test
  - Randomly fuzzes inputs in corpus to expand coverage
- Directed Automated Random Testing
  - Also uses random input for the initial run
  - But subsequently uses symbolic execution to direct search

# DART (Directed Automated Random Testing)

1. Run the program starting with some random inputs
  2. Gather symbolic constraints at conditional statements
  3. Use a constraint solver to generate new test inputs  
(New test inputs should exercise new path)
  4. Go back to 1.
- \* Repeat until all paths are covered
- So what's different from pure symbolic execution?
    - Now we have concrete values as well as symbolic values
    - Now constraint solver can do a much better job

# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



$x = 0, y = 0$

create symbolic  
variables X, Y

# Directed Search

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

create symbolic  
variables X, Y

$X \leq Y$

Solve:  $!(X \leq Y)$

Solution:  $X=1, Y=0$

$x = 0, y = 0$

# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



x = 1, y = 0

create symbolic  
variables X, Y

# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



x = 1, y = 0

create symbolic  
variables X, Y

X > Y

# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



$x = 1, y = 0$

create symbolic  
variables X, Y

$X > Y$

$x = X + Y$



# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



x = 1, y = 1

create symbolic  
variables X, Y

X > Y

x = X+Y  
y = X

# Directed Search

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```



x = 0, y = 1

create symbolic  
variables X, Y

X > Y

y = X

x = Y

# Directed Search

```
int x, y;  
if (x > y) {  
    x = x + y;  
    y = x - y;  
    x = x - y;  
    if (x > y)  
        assert false;  
}
```

Concrete  
Execution

Symbolic  
Execution

Path  
Constraint

create symbolic  
variables X, Y

$X > Y$

Solve:  $X > Y$  AND  $!(Y \leq X)$

Impossible: DONE!

$y = X$

$x = Y$

$x = 0, y = 1$

$Y \leq X$

# DART (Directed Automated Random Testing)

- Gaining popularity in industry
  - + Unlike symbolic execution, can work on complex apps
  - + Unlike stochastic testing, can achieve very high coverage
- Many tools
  - PEX, SAGE, YOGI (Microsoft)
  - KLEE: LLVM open source project
- Many applications
  - Bug finding, security, web and database applications, etc.

# State Space Reduction Techniques

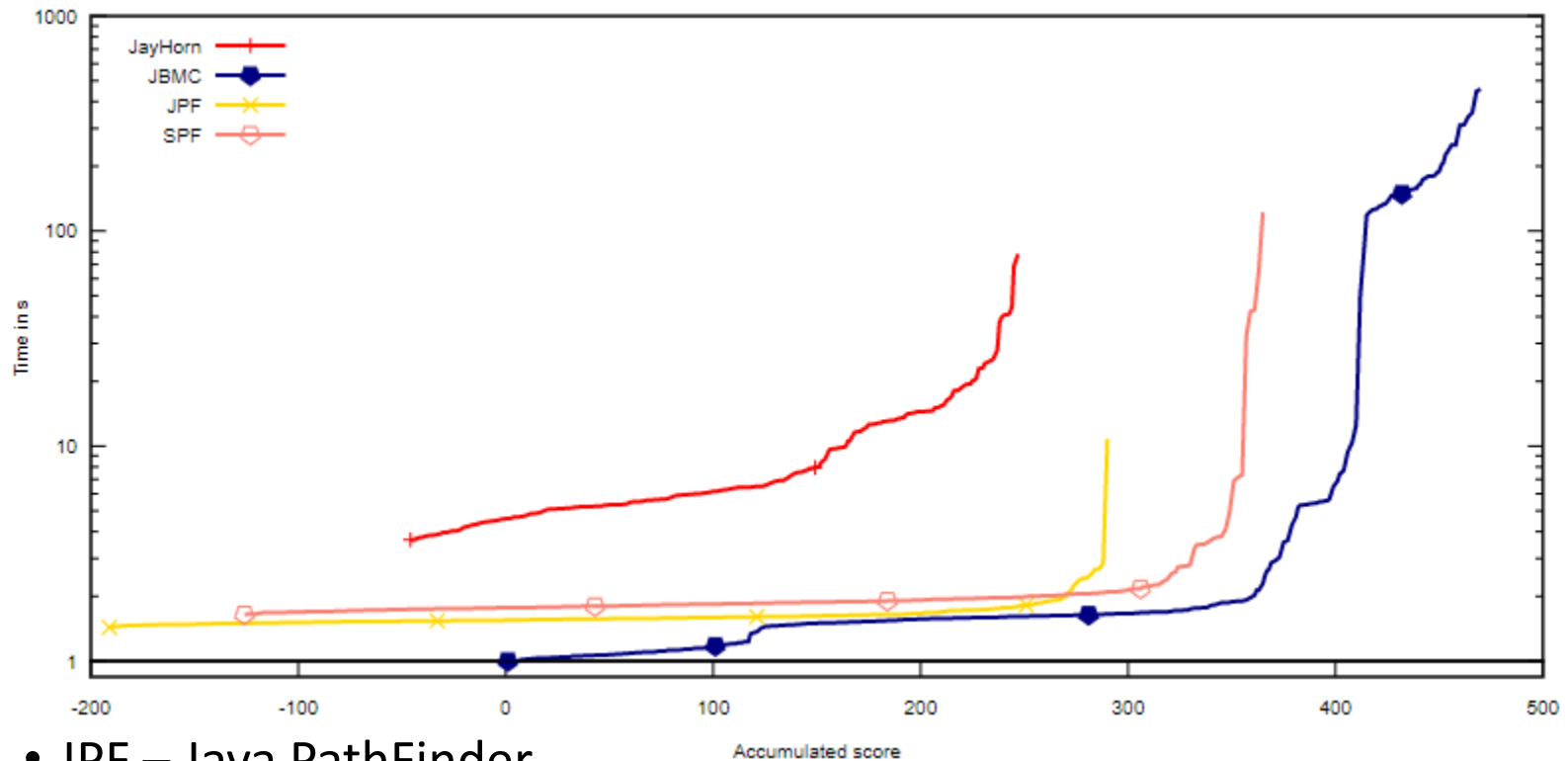
- State collapsing
  - Heuristic state approximation
  - Hash compaction
  - Heap canonicalization
  - Symbolic execution
- 
- What if the state space is *still* too large?
  - One recourse – reduce the problem size

# Reducing the problem size

- When state space explosion prevents exhaustive exploration, What are the alternatives?
  1. Put a cap on problem size and exhaustively explore
  2. Put a cap on time / space and do a partial exploration
- In most cases, putting a cap on problem size is better
  - Most corner cases exhibit with a relatively small problem size
  - Correctness with smaller sizes usually translate to larger sizes
- Examples of capping problem size
  - Instead of checking an infinite tree structure, check a limited tree
  - Instead of checking infinite number of players, check just 10
  - Etc.

# Model Checking is Getting Better Every Year

<https://sv-comp.sosy-lab.org/2019/results/results-verified/>



- JPF – Java PathFinder
- SPF – Symbolic Java PathFinder (JPF with symbolic execution)
- JBMC – Java Bounded Model Checker (2018 newcomer)

# References

- Ranjit Jhala and Rupak Majumdar. 2009. “Software model checking”. ACM Computing Surveys: <https://people.mpi-sws.org/~rupak/Papers/SoftwareModelChecking.pdf>
- Cristian Cadar and Koushik Sen. 2013. “Symbolic execution for software testing: three decades later”. Communications of the ACM: <https://people.eecs.berkeley.edu/~ksen/papers/cacm13.pdf>
- 8<sup>th</sup> Competition on Software Verification (SV-COMP), 2019: <https://sv-comp.sosy-lab.org/2019/results/results-verified/>