CS1632, Lecture 19: Static Analysis, Part 3

Wonsun Ahn

State Space Reduction Techniques

- State collapsing
- Heuristic state approximation
- Hash compaction
- Heap canonicalization
- Symbolic execution

Symbolic Execution

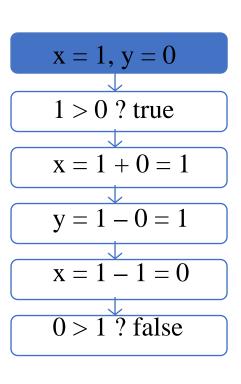
- Analysis of programs with unspecified inputs
 - Execute a program on symbolic inputs
- Symbolic states represent sets of concrete states
- For each path, build a path condition
 - Condition on inputs for the execution to follow that path
 - Check path condition satisfiability -- explore only feasible paths
- Symbolic state
 - Symbolic values/expressions for variables
 - Path condition
 - Program counter

Example: Standard 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;

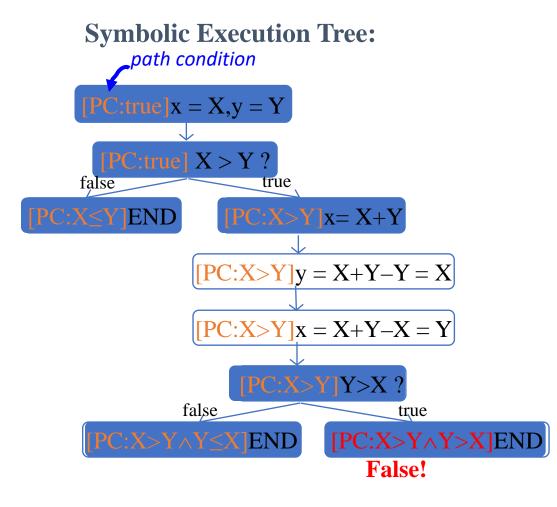
Concrete Execution Path



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;
```



Solve path conditions \rightarrow test inputs

Questions

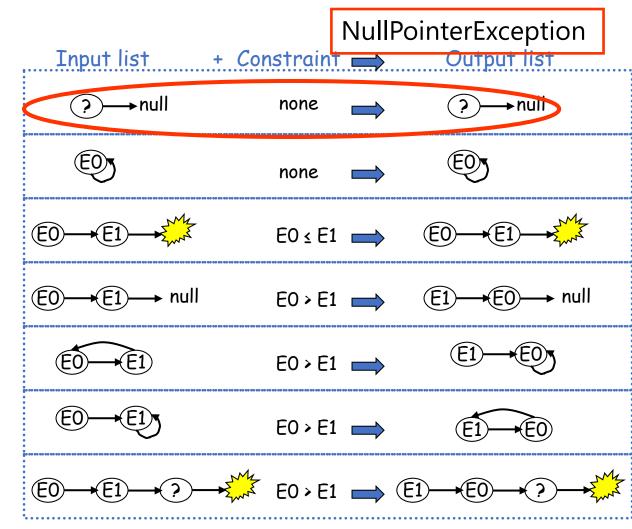
- What about loops?
- What about overflow?
- What about multi-threading?
- What about data structures?

Generalized Symbolic Execution [TACAS'03]

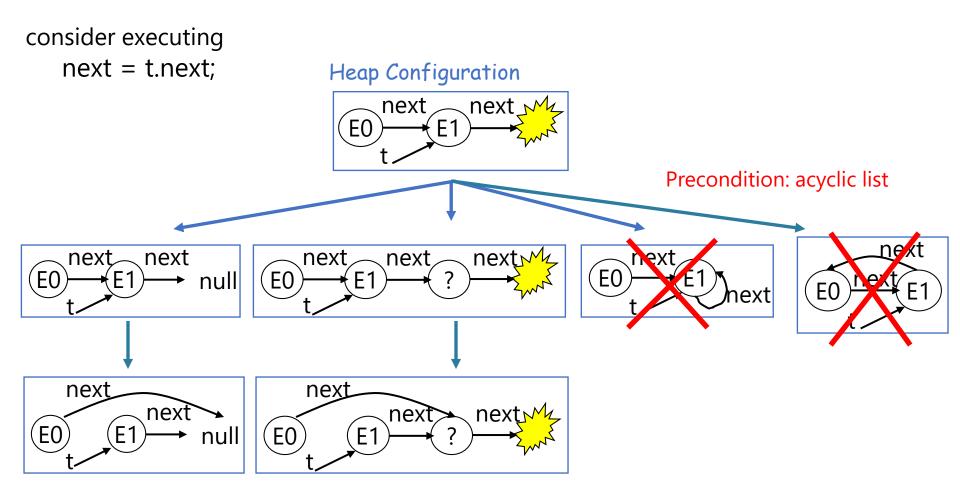
- Handles dynamically allocated data structures and multi-threading
- Key elements:
 - Lazy initialization for input data structures
 - Standard model checker (Java PathFinder) for multi-threading
- Model Checker
 - Analyzes thread inter-leavings
 - Leverages optimizations
 - Symmetry and partial order reductions, abstraction etc.
 - Generates and explores the symbolic execution tree
 - Explores different heap configurations explicitly -- non-determinism handles aliasing

Example Analysis

```
class Node {
   int elem;
   Node next;
   Node swapNode() {
     if (payt I - pull)
        if (elem > next.elem) {
           Node t = next;
           next = t.next;
           t.next = this;
           return t;
      return this;
```



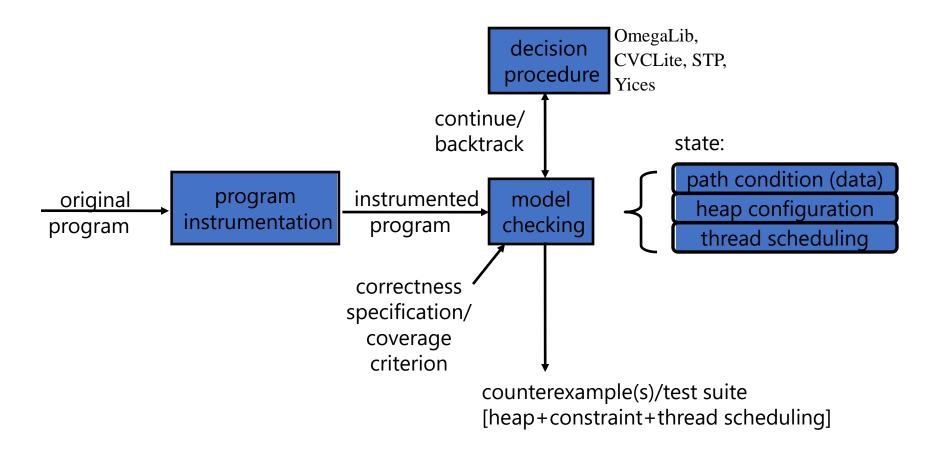
Lazy Initialization (illustration)



Implementation

- Symbolic execution of Java programs
- Code instrumentation
 - Programs instrumented to enable JPF to perform symbolic execution
 - Replace concrete operations with calls to methods that implement symbolic operations
 - General: could use/leverage any model checker
- Decision procedures used to check satisfiability of path conditions
 - Omega library for integer linear constraints
 - CVCLite, STP (Stanford), Yices (SRI)

Implementation via Instrumentation



Symbolic PathFinder

- No longer uses code instrumentation
- Implements a non-standard interpreter of byte-codes
 - Enables JPF to perform symbolic analysis
 - Replaces standard byte-code execution with non-standard symbolic execution
- During execution checks for assert violations, run-time errors, etc.
- Symbolic information:
 - Stored in attributes associated with the program data
 - Propagated dynamically during symbolic execution
- Choice generators and listeners:
 - Non-deterministic choices handle branching conditions
 - Listeners print results: path conditions, test vectors/sequences
- Native peers model native libraries:
 - Capture Math calls and send them to the constraint solver
- Generic interface for multiple decision procedures
 - Choco, IASolver, CVC3, Yices, HAMPI, CORAL [NFM11], etc.

Example: IADD

Concrete execution of IADD byte-code:

```
public class IADD extends
 Instruction { ...
 public Instruction execute(...
 ThreadInfo th){
 int v1 = th.pop();
 int v2 = th.pop();
 th.push(v1+v2,...);
 return getNext(th);
```

Symbolic execution of IADD byte-code:

```
public class IADD extends
   ....bytecode.IADD { ...
public Instruction execute(...
   ThreadInfo th){
   Expression sym v1 = ....getOperandAttr(0);
   Expression sym_v2 = ....getOperandAttr(1);
   if (sym_v1 == null && sym_v2 == null)
     // both values are concrete
     return super.execute(... th);
   else {
     int v1 = th.pop();
     int v2 = th.pop();
     th.push(0,...); // don't care
     ....setOperandAttr(Expression._plus(
    sym_v1,sym_v2));
     return getNext(th);
```

Complex mathematical constraints

Model-level interpretation of calls to math functions

$$$x + 1 \longrightarrow Math.sin \longrightarrow sin($x + 1)$$

Symbolic expression (un-interpreted function) denoting the result value of the call

CORAL solver [NFM'11]

- Target applications:
 - Symbolic execution of programs that manipulate floating-p
 - Use floating-point arithmetic
 - Call specific math functions (from java.lang.Math)

Common in software from NASA

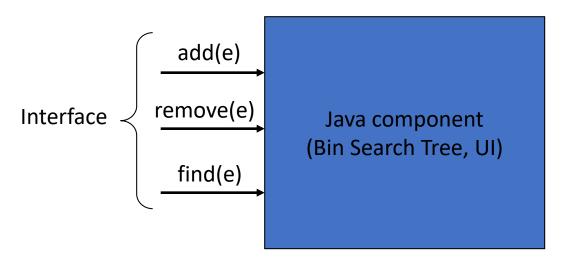
sqrt(exp(x+z))) < pow(z,
x)
$$\land$$
 x>0 \land y>1 \land z>1 \land y\land
w=x+2



> {x=4.31, y=6.08, z=9.51, w=6.31}

- Meta-heuristic solver
 - Distance-based fitness function
- Particle swarm optimization (PSO)
 - Search simulates movements in a group of animals
 - Used opt4j library (see <u>opt4j.sourceforge.net</u>)

Applications: Test Input and Sequence Generation



```
Generated test sequence:
    BinTree t = new BinTree();
    t.add(1);
    t.add(2);
    t.remove(1);
```

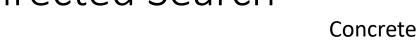
- SymbolicSequenceListener generates JUnit tests
- JUnit tests can be run directly by the developers
- Measure coverage (e.g. MC/DC)

Dynamic Techniques

- Classic symbolic execution is a static technique
- Dynamic techniques
 - Collect symbolic constraints during concrete executions
 - DART = Directed Automated Random Testing
 - Concolic (Concrete Symbolic) testing

DART = Directed Automated Random Testing

- Dynamic test generation
 - Run the program starting with some random inputs
 - Gather symbolic constraints on inputs at conditional statements
 - Use a constraint solver to generate new test inputs
 - Repeat the process until a specific program path or statement is reached (classic dynamic test generation [Korel90])
 - Or repeat the process to attempt to cover ALL feasible program paths (DART [Godefroid et al PLDI'05])
- Detect crashes, assert violations, runtime errors etc.



Concrete Execution

x = 0, y = 0

Path Constraint

if
$$(x > y)$$
 {

$$x = x + y$$
;

$$y = x - y$$
;

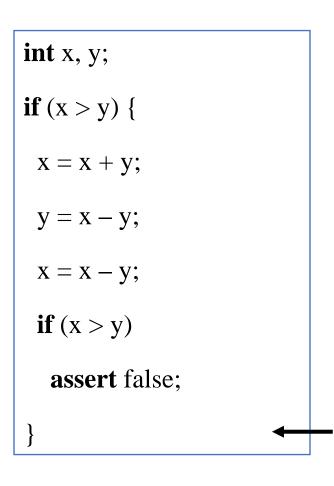
$$x = x - y$$
;

if
$$(x > y)$$

assert false;

}

create symbolic variables x, y





x = 0, y = 0

Symbolic Path Execution Constraint create symbolic variables x, y $x \le y$ Solve: $!(x \le y)$ Solution: x=1, y=0

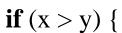


Concrete Execution

x = 1, y = 0

Symbolic Execution

Path Constraint



$$x = x + y$$
;

$$y = x - y$$
;

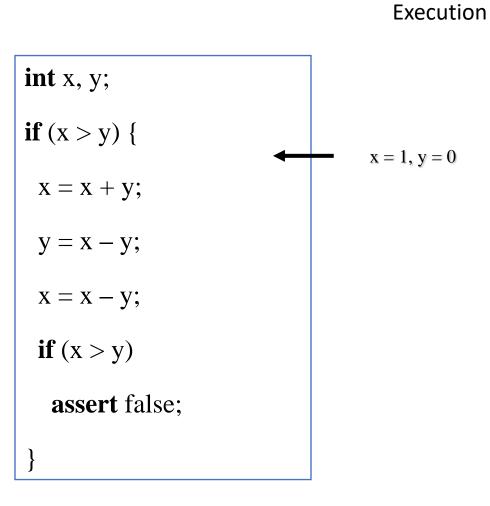
$$x = x - y$$
;

if (x > y)

assert false;

}

create symbolic variables x, y



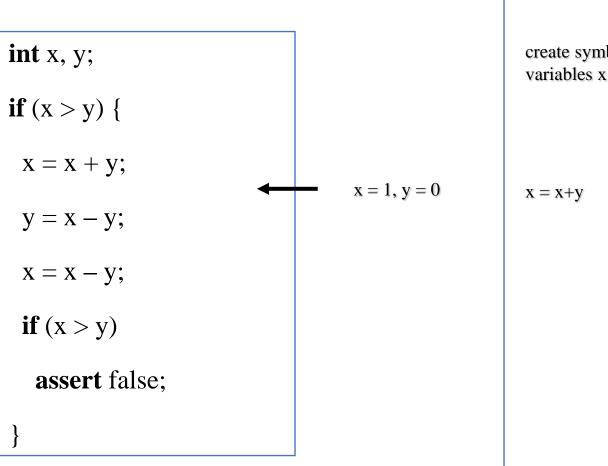
Symbolic Execution

create symbolic variables x, y

Concrete

Path Constraint

x > y



Concrete

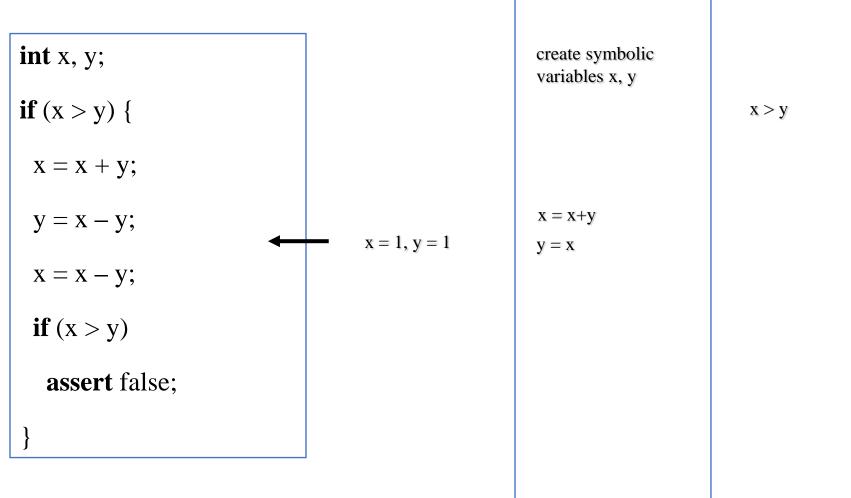
Execution

Symbolic Execution

create symbolic variables x, y

Path Constraint

x > y



Concrete

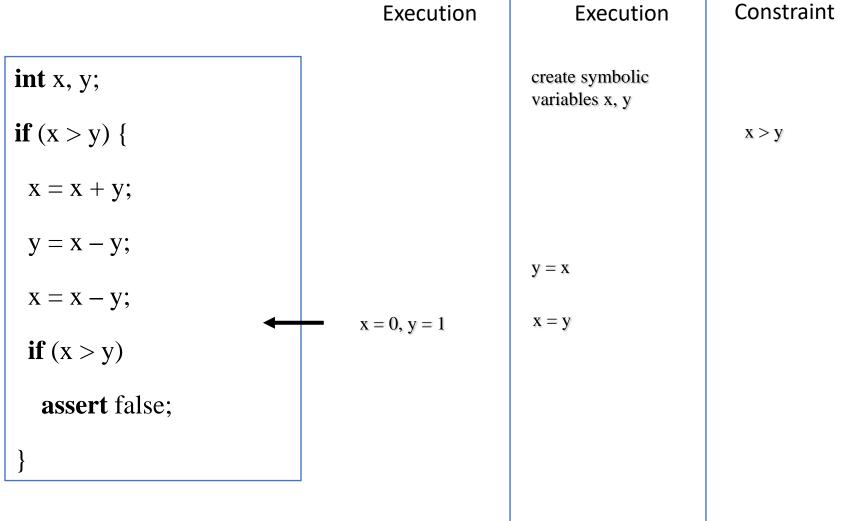
Execution

Symbolic

Execution

Path

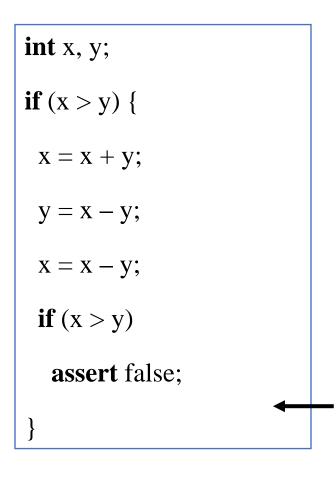
Constraint

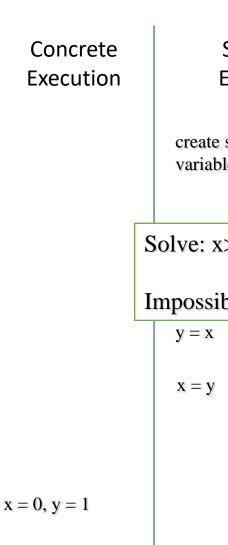


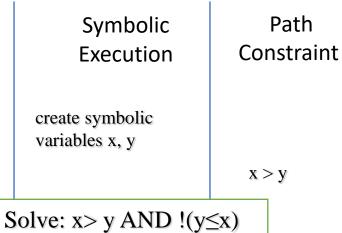
Concrete

Symbolic

Path







$$y \le x$$

Dynamic Test Generation

- Very popular
- Implemented and extended in many interesting ways
- Many tools
 - PEX, SAGE, CUTE, jCUTE, CREST, SPLAT, etc
- Many applications
 - Bug finding, security, web and database applications, etc.
- EXE (Stanford Univ. [Cadar et al TISSEC 2008])
 - Related dynamic approach to symbolic execution

Mixed Concrete-Symbolic Solving [ISSTA'11]

- Use un-interpreted functions for external library calls
- Split path condition PC into:
 - simplePC solvable constraints
 - complexPC non-linear constraints with un-interpreted functions
- Solve simplePC
 - Use obtained solutions to simplify complexPC
 - Check the result again for satisfiability

Example (assume hash(x) = 10 *x):

Solve simplePC; use solution X=4 to compute h(4)=40

Simplify complexPC: Y=40

Solve again: $X>3 \land Y>10 \land Y=40$ Satisfiable!

Challenge

Path explosion

Symbolic execution of a program may result in a very large, possibly infinite number of paths

Problem: loops and recursion

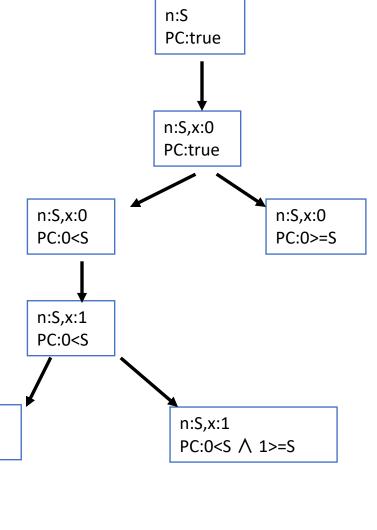
n:S,x:1

PC:0<S ∧ 1<S

Infinite symbolic execution tree

Example Code

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



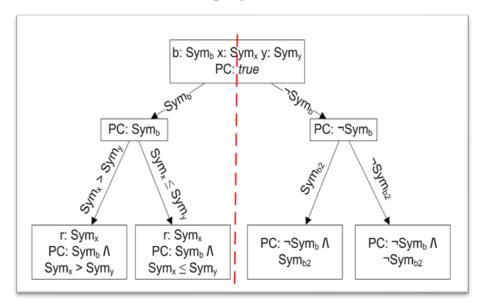
Solutions

- Dealing with loops and recursion
 - Put bound on search depth or on number of PCs
 - Stop search when desired coverage achieved
 - Loop abstraction [Saxena et al ISSTA'09] [Godefroid ISSTA'11]
- Solutions addressing path explosion
 - Parallel Symbolic Execution
 - Abstract State Matching
 - Compositional DART = SMART

Parallel Symbolic Execution

- Path explosion
 - Increases exponentially with the number of inputs specified as symbolic
 - Very expensive in terms of time (weeks, months)
- Solution
 - Speed-up symbolic execution using parallel or distributed techniques
- Symbolic execution is amenable to parallelization
 - No sharing between sub-trees

Balancing partitions



Nicely Balanced – linear speedup

Poorly Balanced – no speedup

- Solutions
 - Simple static partitioning [ISSTA'10]
 - Dynamic partitioning [Andrew King's Masters Thesis at KSU, Cloud9 at EPFL, Fujitsu]

Simple Static Partitioning

- Static partitioning of tree with light dynamic load balancing
 - Flexible, little communication overhead
- Constraint-based partitioning
 - Constraints used as initial pre-conditions
 - Constraints are disjoint and complete
- Approach
 - Shallow symbolic execution => produces large number of constraints
 - Constraints selection according to frequency of variables
 - Combinatorial partition creation
- Intuition
 - Commonly used variables likely to partition state space in useful ways
- Results
 - maximum analysis time speedup of 90x observed using 128 workers and a maximum test generation time speedup of 70x observed using 64 workers.

Abstract State Matching

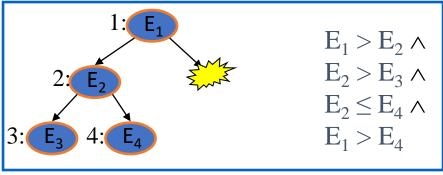
- State matching subsumption checking [SPIN'06, J. STTT 2008]
 - Obtained through DFS traversal of "rooted" heap configurations
 - Roots are program variables pointing to the heap
 - Unique labeling for "matched" nodes
 - Check logical implication between numeric constraints
 - Not enough to ensure termination
- Abstraction
 - Store abstract versions of explored symbolic states
 - Use subsumption checking to determine if an abstract state is re-visited
 - Decide if the search should continue or backtrack

Abstract State Matching

- Enables analysis of under-approximation of program behavior
- Preserves errors to safety properties -- useful for testing
- Automated support for two abstractions (inspired by shape analysis [TVLA]
 - Singly linked lists
 - Arrays

State Matching: Subsumption Checking

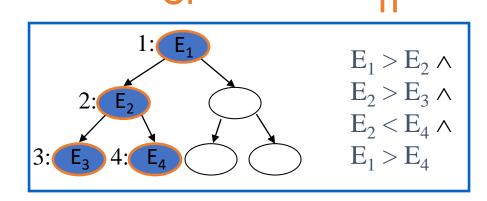




Set of concrete states represented by stored state

U

New state:



Set of concrete states represented by new state

Normalized using existential quantifier elimination

Abstractions for Lists and Arrays

- Shape abstraction for singly linked lists
 - Summarize contiguous list elements not pointed to by program variables into summary nodes
 - Valuation of a summary node
 - Union of valuations of summarized nodes
 - Subsumption checking between abstracted states
 - Same algorithm as subsumption checking for symbolic states
 - Treat summary node as an "ordinary" node
- Abstraction for arrays
 - Represent array as a singly linked list
 - Abstraction similar to shape abstraction for linked lists

Compositional DART [POPL'07]

- Idea: compositional dynamic test generation
 - use summaries of individual functions like in inter-procedural static analysis
 - if f calls g, analyze g separately, summarize the results, and use g's summary when analyzing f
 - A summary $\phi(g)$ is a disjunction of path constraints expressed in terms of input pre-conditions and output post-conditions:

```
\phi(g) = V\phi(w), with \phi(w) = pre(w) \land post(w)
```

- g's outputs are treated as symbolic inputs to calling function f
- SMART: Top-down strategy to compute summaries on a demand-driven basis from concrete calling contexts
- Same path coverage as DART but can be exponentially faster!
- Follow-up work
 - Anand et al. [TACAS'08], Godefroid et al. [POPL'10]

Example

```
int is positive(int x) {
 if (x>0) return 1;
 return 0:
#define N 100
void top (int s[N]) {// N inputs
int i, cnt=0;
for (i=0;i<N; i++)
 cnt=cnt+is positive(s[i]);
if (cnt == 3) error(); // (*)
return;
```

Program P = {top, is_positive} has 2^N feasible paths
DART will perform 2^N runs
SMART will perform only 4 runs

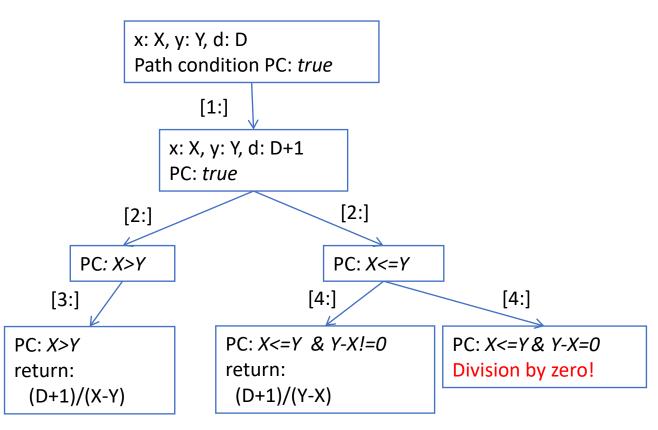
- 2 to compute summary
 φ(is_positive) = (x>0∧ ret=1) ∨ (x≤0∧ ret=0)
- 2 to execute both branches of (*) by solving:

Applications – An Example

Symbolic execution tree:

Method m:

1: d=d+1; 2: if (x > y) 3: return d / (x-y); else 4: return d / (y-x);



Solve path conditions \rightarrow test inputs

Auto-generated JUnit Tests

```
@Test public void t1() {
    m(1, 0, 1);
}
@Test public void t2() {
    m(0, 1, 1);
}
@Test public void t3() {
    m(1, 1, 1);
}
Fail  X PC: X<=Y & Y-X=0 ⇔ X=Y
```

Achieves full path coverage

Invariant Generation

- Pre-condition:
 - "x!=y"
- Post-condition:
 - "\result==((x>y)? (d+1)/(x-y): (d+1)/(y-x)"

Symbolic Execution and Software Testing

- Tools, many open-source
 - NASA's Symbolic (Java) Pathfinder
 http://babelfish.arc.nasa.gov/trac/jpf/wiki/projects/jpf-symbolic
 - UIUC's CUTE and jCUTE
 http://osl.cs.uiuc.edu/~ksen/cute
 - Stanford's KLEE
 http://klee.llvm.org/
 - UC Berkeley's CREST and BitBlaze http://code.google.com/p/crest
 - Microsoft's Pex, SAGE, YOGI, PREfix
 http://research.microsoft.com/en-us/projects/pex/
 http://research.microsoft.com/en-us/projects/yogi
 - IBM's Apollo, Parasoft's testing tools etc.