

Automated Testing

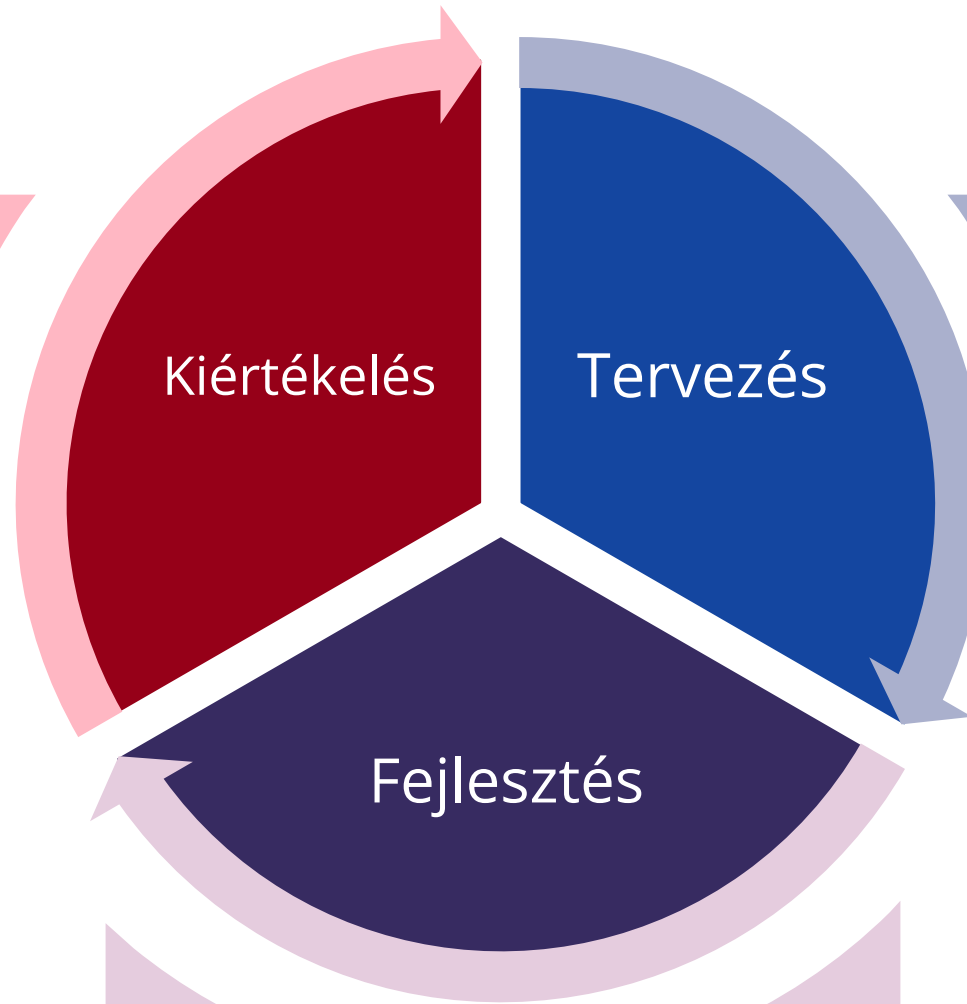


**Critical Systems
Research Group**

- Idő:
 - Most péntek: nincs tanítás.
 - Jövő Hétfő: hf.
 - Jövő Kedd: FONTOS!
 - Meghívott előadó
 - Vizsgakonzultáció, mintavizsga
 - 10 perc önlab tárgy adminisztratív eligazítás
- TDK5

Overview

- Performance evaluation
- Data Analysis
- Code Quality
- Static Analysis
- Testing & Coverage



- Graph-Based Modeling
- Textual Modeling
- Code Generation
- Model Intelligence
- Model Checking

- Build Automation
- Code Generation
- Code Intelligence

Basic concepts

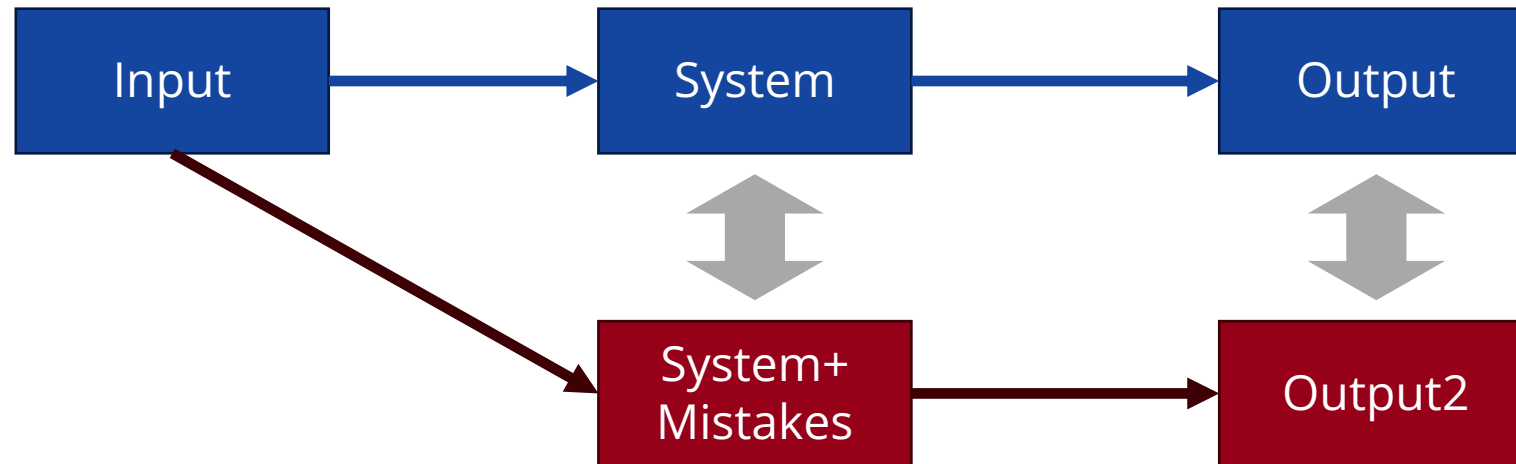
Testing setup

- Normal setup



Testing setup

- **Fault hypothesis:** instead of a correct system, we have an system with faults that can lead to failures



Mutation operators

- Potential faults are characterized by a set of mutations.

$$O = \{op_1, op_2, \dots, op_n\}$$

e.g.:

- Use + instead of –, Use = instead of ==

or more exotic:

- Accidentally delete an edge from system model
- Create biased training set
- Miss a requirement

- Mutation operations can create mutants from a system S .

$$M = \{op_1(S), op_2(S), \dots, op_n(S)\}$$

This M can be **huge**, so sometimes sampled, or used in statistics

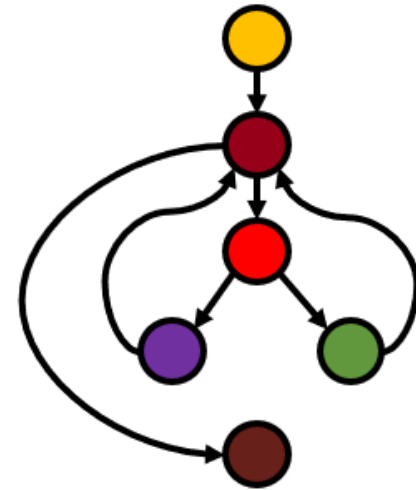
Usage of mutations

- identify weakly tested pieces of code
- identify weak tests
- compute the mutation score: $\frac{\text{mutants killed}}{\text{\#mutants}}$
- error propagation and state infection in the program

My takeaway: mutation is the basis of a testing activity.

- E.g., code coverage = proportion of mutants possible to reach.
- If new testing technique is required → mutation testing

Structure-based Testing



Structure-based Testing: Outline

- **Recap: basic concepts**
- Control-flow criteria
- Data-flow criteria (*optional*)
- Evaluation of structure-based testing

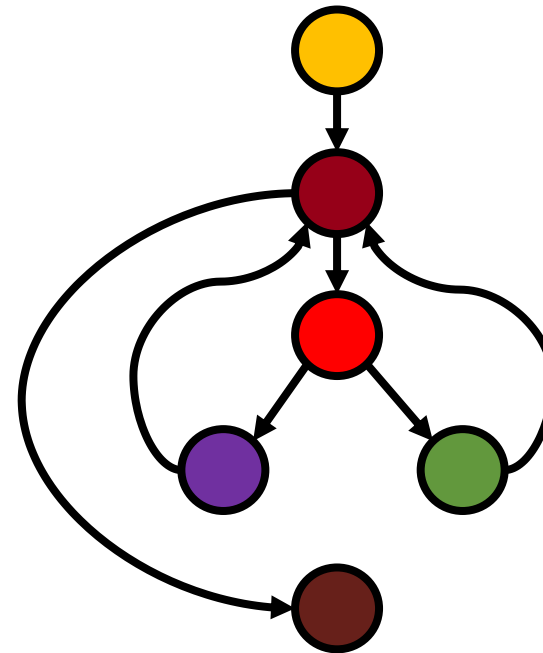
What is “internal structure”?

- In case of models: structure of the model
- In case of code: structure of the code (CFG)

Source code:

```
int a = 1;  
while(a < 16) {  
    if(a < 10) {  
        a += 2;  
    } else {  
        a++;  
    }  
}  
a = a * 2;
```

Control-flow graph:



Basic concepts

```
int t = 1;  
Speed s = SLOW;
```

```
if (! started){  
    start();  
}
```

```
if (t > 10 && s == FAST){  
    brake();  
}  
else {  
    accelerate();  
}
```

Statement

Block

Condition

Decision

Branch

Basic concepts

- Statement
- Block
 - A sequence of one or more consecutive executable statements containing no branches or function calls
- Condition
 - Logical expression without logical operators (and, or...)
- Decision
 - A logical expression consisting of one or more conditions combined by logical operators
- Branch
 - Possible outcome of a decision
- Path
 - A sequence of events, e.g., executable statements, of a component typically from an entry point to an exit point.

Example: decision and condition

- A decision with one condition:

```
if (temp > 20) {...}
```

- A decision with 3 conditions:

```
if (temp > 20 && (valveIsOpen || p == HIGH)) {...}
```

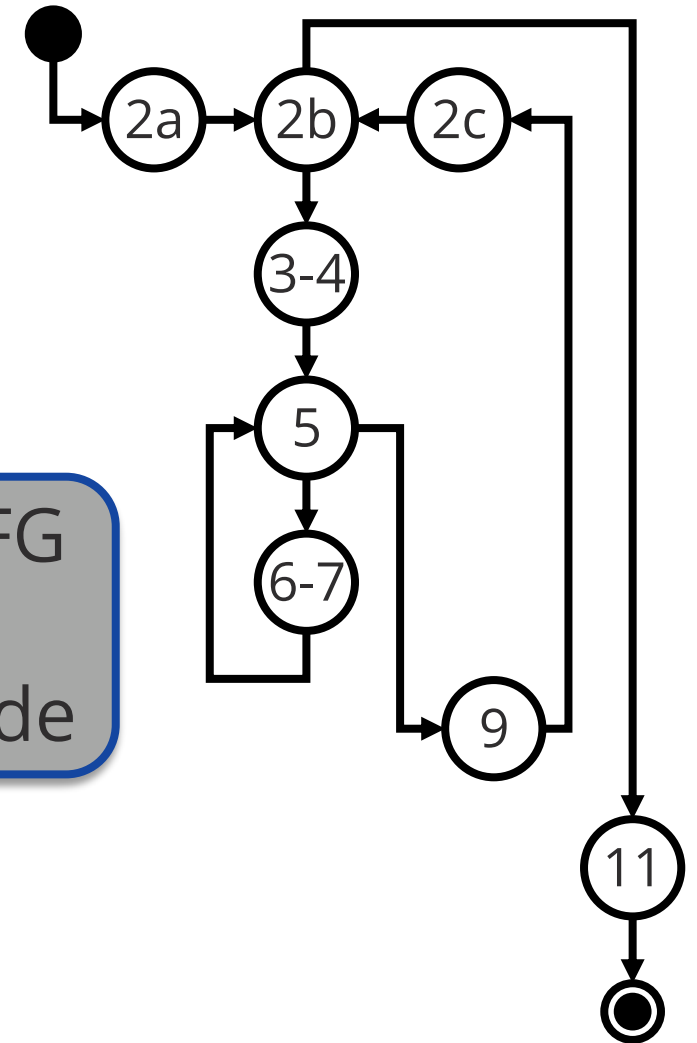

Control Flow Graph (CFG)

- A CFG represents the flow of control
- **$G = (N, E)$ directed graph**
 - Node $n \in N$ is a basic block
 - Basic block: Sequence of statements with exactly one entry and exit points.
 - Edge $e = (n_i, n_j) \in E$ is a possible flow of control from basic block n_i to basic block n_j

EXERCISE: Building a CFG

```
1: public void insertionSort(int[] a) {  
2:     for(int i = 0; i < a.size(); i++) {  
3:         int x = a[i];  
4:         int j = i - 1;  
5:         while(j >= 0 && a[j] > x) {  
6:             a[j+1] = a[j];  
7:             j = j - 1;  
8:         }  
9:         a[j+1] = x;  
10:    }  
11:    System.out.println("Finished.");  
12: }
```

Build the CFG
of this
program code



Structure-based Testing: Outline

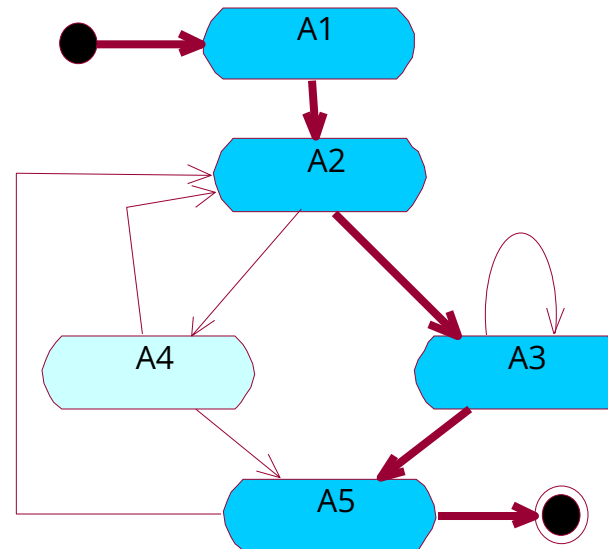
- Recap: basic concepts
- **Control-flow criteria**
- Data-flow criteria (*optional*)
- Evaluation of structure-based testing

Learning outcomes

- Explain the differences between different control-flow based coverage criteria (K2)
- Design tests using control-flow based coverage criteria for imperative programs (K3)

1. Statement coverage (recap)

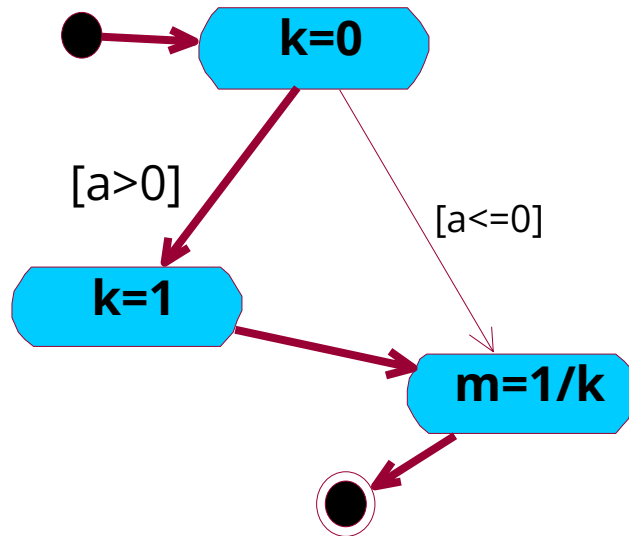
Number of statements executed during testing
Number of all statements



Statement coverage: $4/5 = 80\%$

Assessing statement coverage

All statement is executed at least once

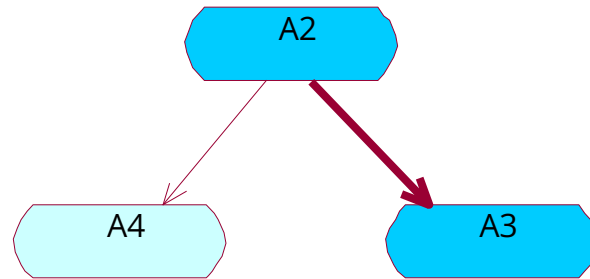


Statement coverage: 100%
BUT: **[a<=0]** branch missing!

Does not guarantee coverage of empty branches

2. Decision coverage (recap)

Outcomes of decisions taken during testing
Number of all possible outcomes



Decision coverage: $1/2 = 50\%$

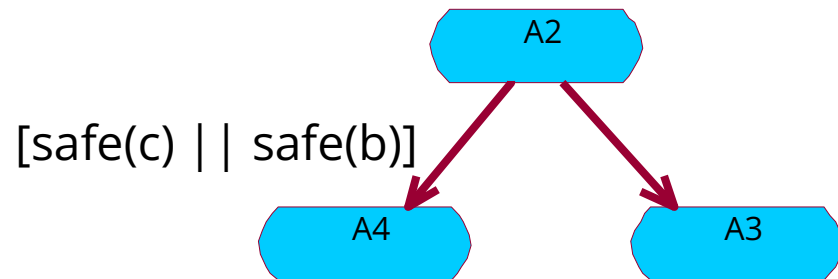
How many outcomes can a decision have?

Note: covering combinations of decisions is not required

Assessing decision coverage

All statement is executed at least once

All outcomes of decisions are covered



100% decision coverage:

#	safe(c)	safe(b)	decision
1	T	F	T
2	F	F	F

safe(b) == True missing!

Does not take into account all combinations of conditions!

3. Condition coverage

Generic coverage metric for conditions:

$$\frac{\text{Number of tested outcomes of conditions}}{\text{Number of aimed outcomes of conditions}}$$

Definition (what conditions are aimed):

- Every condition must be set to true and false during testing

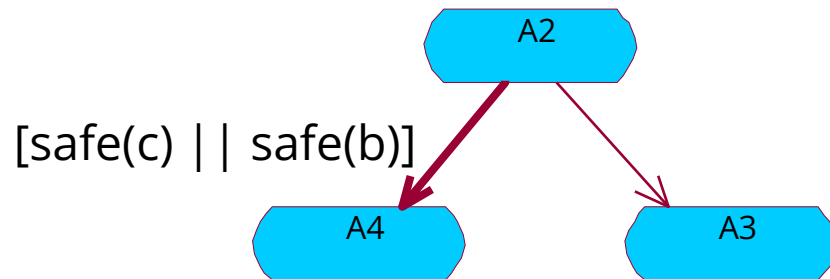
Other possible definition:

- Every condition is **evaluated** to both true and false
 - Not the same as above due to *lazy evaluation*

a	b	a b
F	F	F
F	T	T
T	F	T
T	T	T
T	*	T

Assessing condition coverage

Every condition has taken all possible outcomes at least once



100% condition coverage:

#	safe(c)	safe(b)	decision
1	T	F	T
2	F	T	T

False outcome of decision missing!

Does not yield 100% decision coverage!

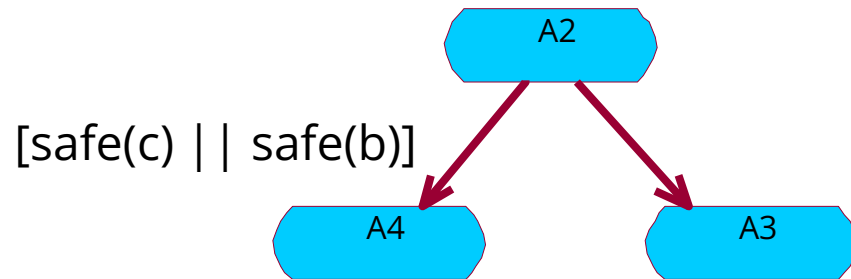
4. Condition/Decision Coverage (C/DC)

Combination of condition and decision coverage

Assessing C/DC Coverage

Every decision has taken all possible outcomes at least once.

Every condition has taken all possible outcomes at least once



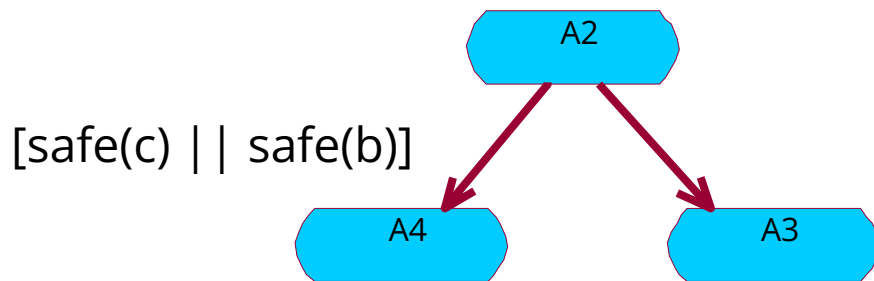
100% C/DC coverage:

#	safe(c)	safe(b)	decision
1	T	T	T
2	F	F	F

Does not take into account whether the condition has any effect!

5. Modified Condition/Decision Coverage (MC/DC)

- Each entry and exit point has been invoked at least once,
- every condition in a decision in the program has taken all possible outcomes at least once,
- every decision in the program has taken all possible outcomes at least once,
- each condition in a decision is shown to independently affect the outcome of the decision.



#	safe(c)	safe(b)	decision
1	T	F	T
2	F	T	T
3	F	F	F

100% MC/DC
coverage

Test generation techniques

Motivation

- **Given a software to test**
 - Availability: source code or binary
- **Extend existing testing**
 - Cover incomplete parts, find specific bugs etc.
- **Idea: generate tests somehow!**
 - Based on various criteria (e.g., coverage)

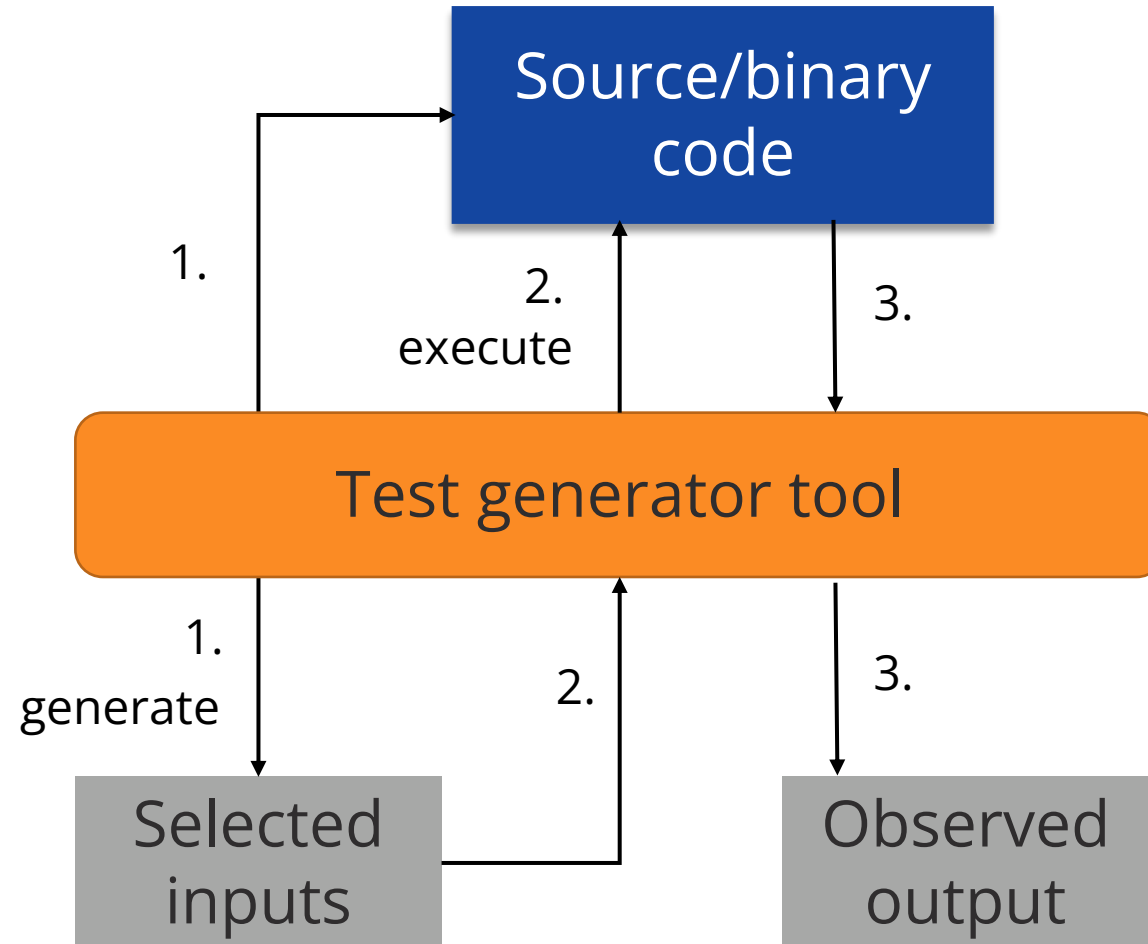
Test selection based on source code

```
int fun1(int a, int b){  
    if (a == 0){  
1      printf(ERROR_MSG);  
2      return -1;  
    }  
    if (b > a)  
3      return b*a + 5;  
    else  
4      return (a+b) / 2;  
}
```

a	b	statement
0	*	1, 2
a!=0	b > a	3
a!=0	b <= a	4

This can be (easily) automated!

Idea of white-box test generation



What is missing?

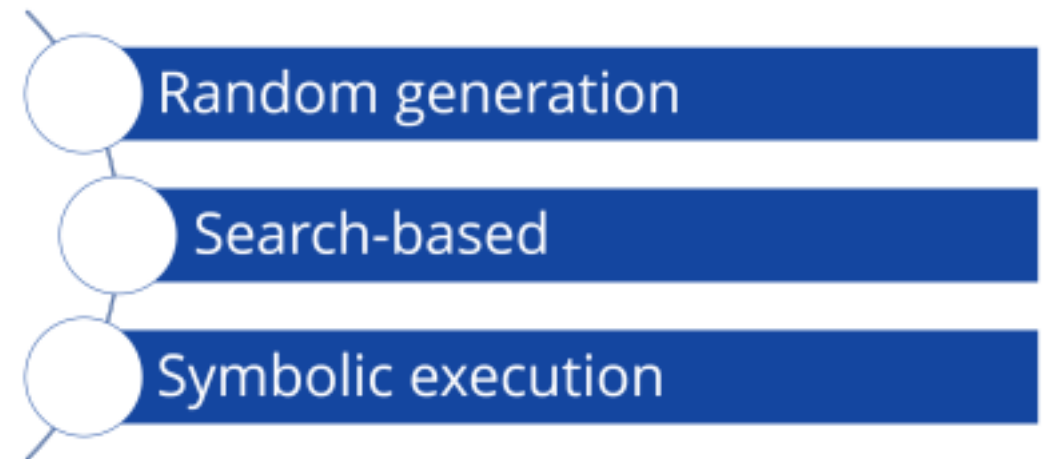
test case = input + *test oracle*

What can be checked without expectations?

- Generic, **implicit** errors (exception, segfault...)
- **Failing assert** statement for different inputs
- Manually extending **assertions** can improve this
- **Derive** from already existing outputs
 - Regression testing, different implementations

see Test oracles lecture

Techniques



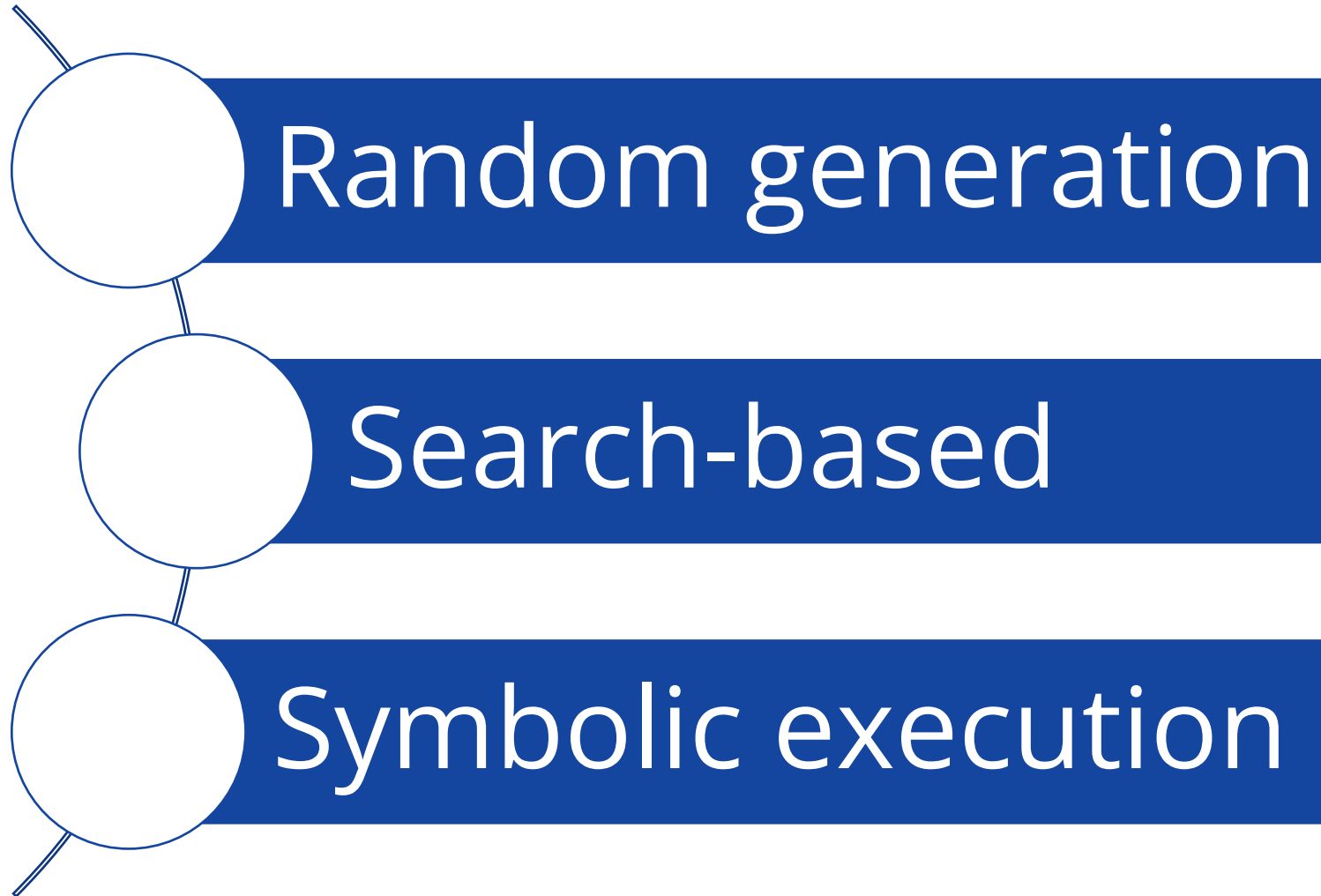
Overview of techniques and goals

Two main questions:

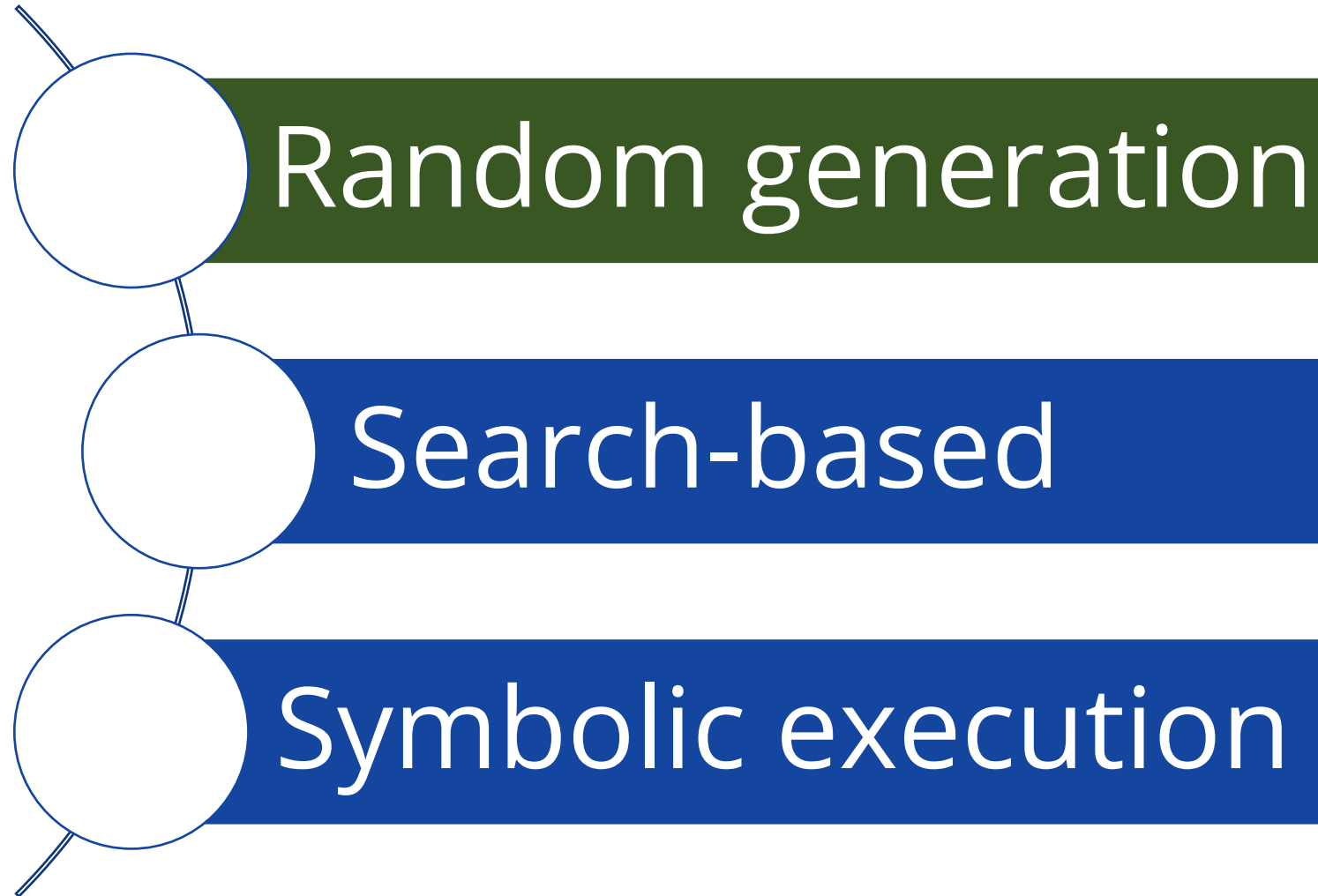
- How to select/search test inputs?
- What to cover with selected inputs? How to evaluate?

		What to cover with search? Evaluation?		
		Code coverage	Crash, exception...	Property, assertion
How to search?	Random			
	Search-based (Metaheuristics)			
	Symbolic execution			

Basic techniques



Basic techniques



Random test generation

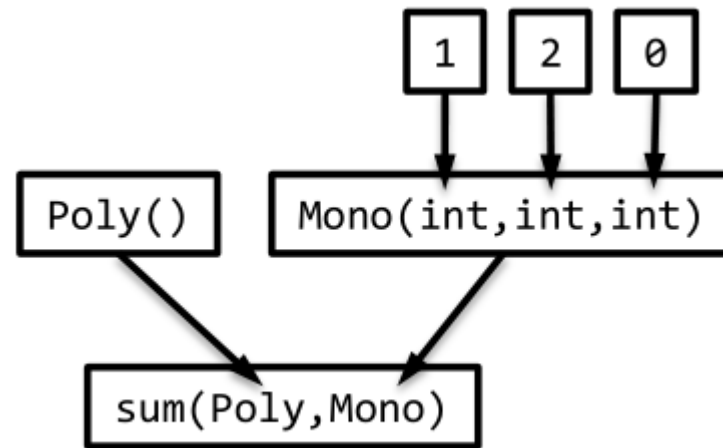
Random selection from input domain

- **Advantage:**
 - Very fast, very cheap
 - Surprisingly easy to find bugs
- **Ideas** (guiding random selection):
 - Try to define input format
 - If no error found: trying different parts of domain
 - Selection based on: "diff", "distance", etc.
- **Tool for Java:**



Randoop: feedback-driven generation

- Generation of method sequence calls
- Compound objects:



- Heuristics:
 - Execution of selected case
 - Throwing away invalid, redundant cases

Cases studies of random testing

- **Robustness testing**

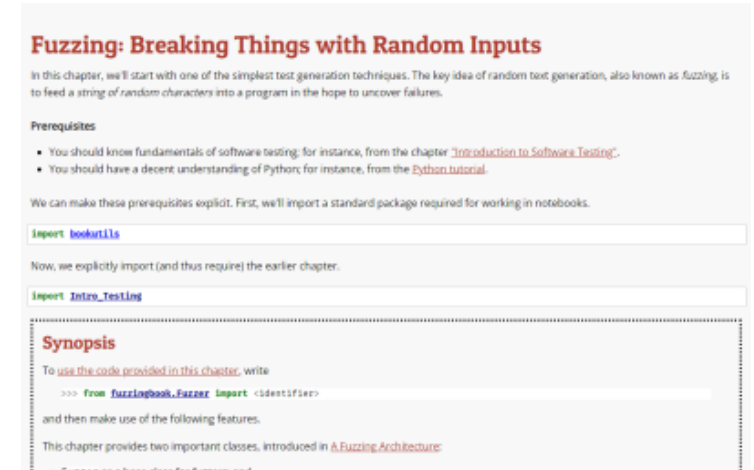
- Fuzz: random inputs for console programs (→ original [assignment](#))
 - Unix (1990), Unix (1995), MacOS (2007)
- NASA: flash file system
 - Simulating HW errors, comparison with references
 - (Model checking did not scale well)

- **Randoop**

- JDK, .NET libraries: checks for basic attributes (e.g.: `o.equals(o)` returns `true`)
- Comparison of JDK 1.5 and 1.6
- Was able to found bugs in well-tested components

Advanced fuzzing techniques

- Mutation-based fuzzing
 - Slightly change existing, valid inputs
 - Get past simple input checking to cover more
- Grammar-based fuzzing
 - Specify format of valid inputs
 - CLI parameters, API...
- ...



[The Fuzzing Book](#)
(Jupyter+Python)

Killer use case: Automated vulnerability detection
(see [security courses](#))

Using annotations in random generation

- If the code contains:
 - pre- and post-conditions (e.g., design by contract)
 - properties or other annotations
- These are able to guide test generation

```
/*@ requires amt > 0 && amt <= acc.bal;  
    @ assignable bal, acc.bal;  
    @ ensures bal == \old(bal) + amt  
    @    && acc.bal == \old(acc.bal - amt); */  
public void transfer(int amt, Account acc) {  
    acc.withdraw(amt);  
    deposit(amt);  
}
```

AutoTest: Bertrand Meyer et al., "Programs that Test Themselves", [IEEE Computer](#) 42:9, 2009.

Tools for property-based test generation

• QuickCheck

- Goal: replace manual input values with generated ones
- Tries to cover laws of input domains
- Minimizes failing tests (e.g., shortest call sequences)

Methods to tests:

byte[] encrypt(byte[] plaintext, Key key)

byte[] decrypt(byte[] ciphertext, Key key)

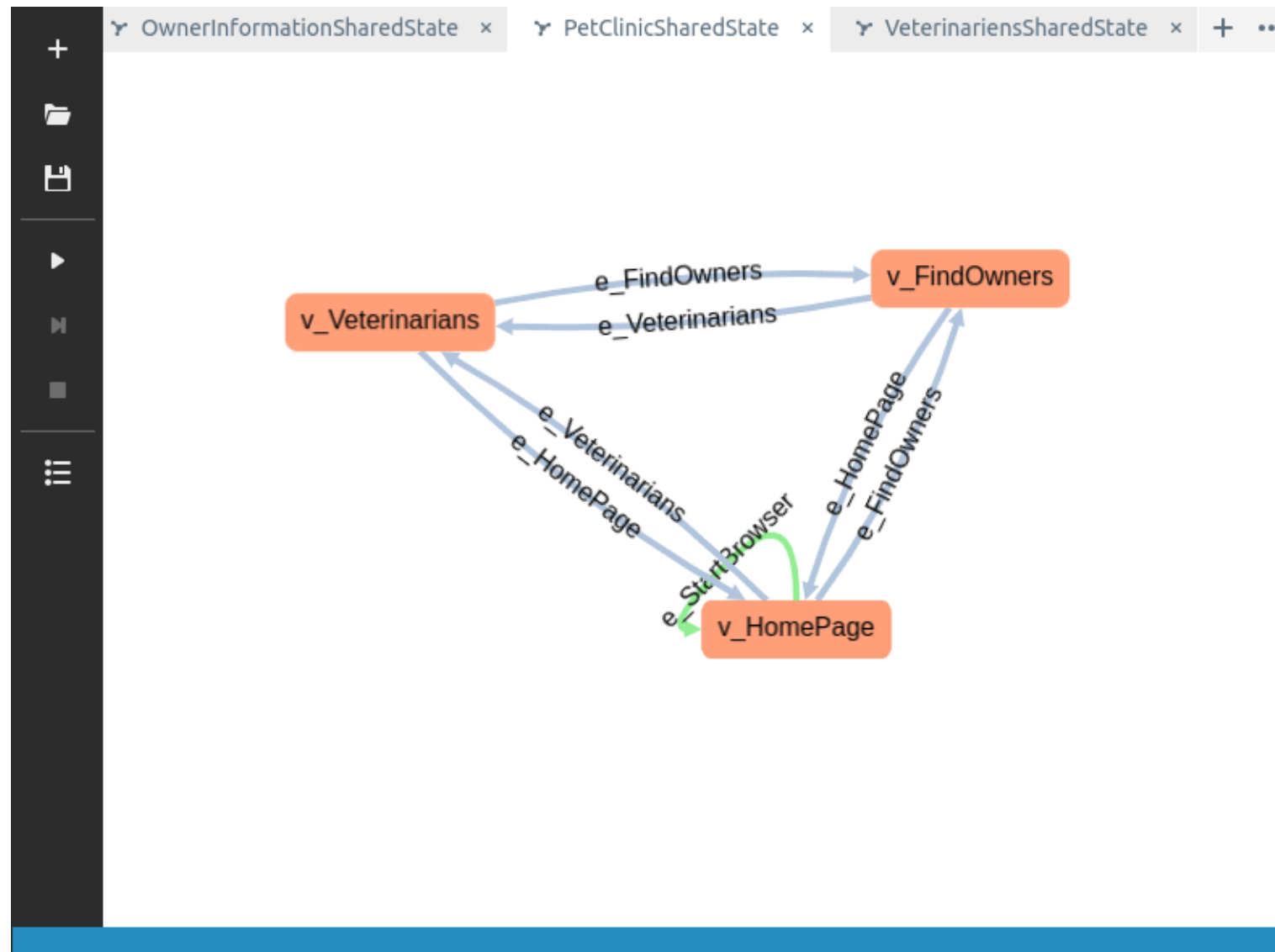
Property:

```
@Property public decryptReversesEncrypt(String plaintext, Key key){  
    Crypto.encrypt(plaintext.getBytes("US-ASCII"), key);  
    assertEquals(plaintext, new String(Crypto.decrypt(ciphertext, key)));  
}
```

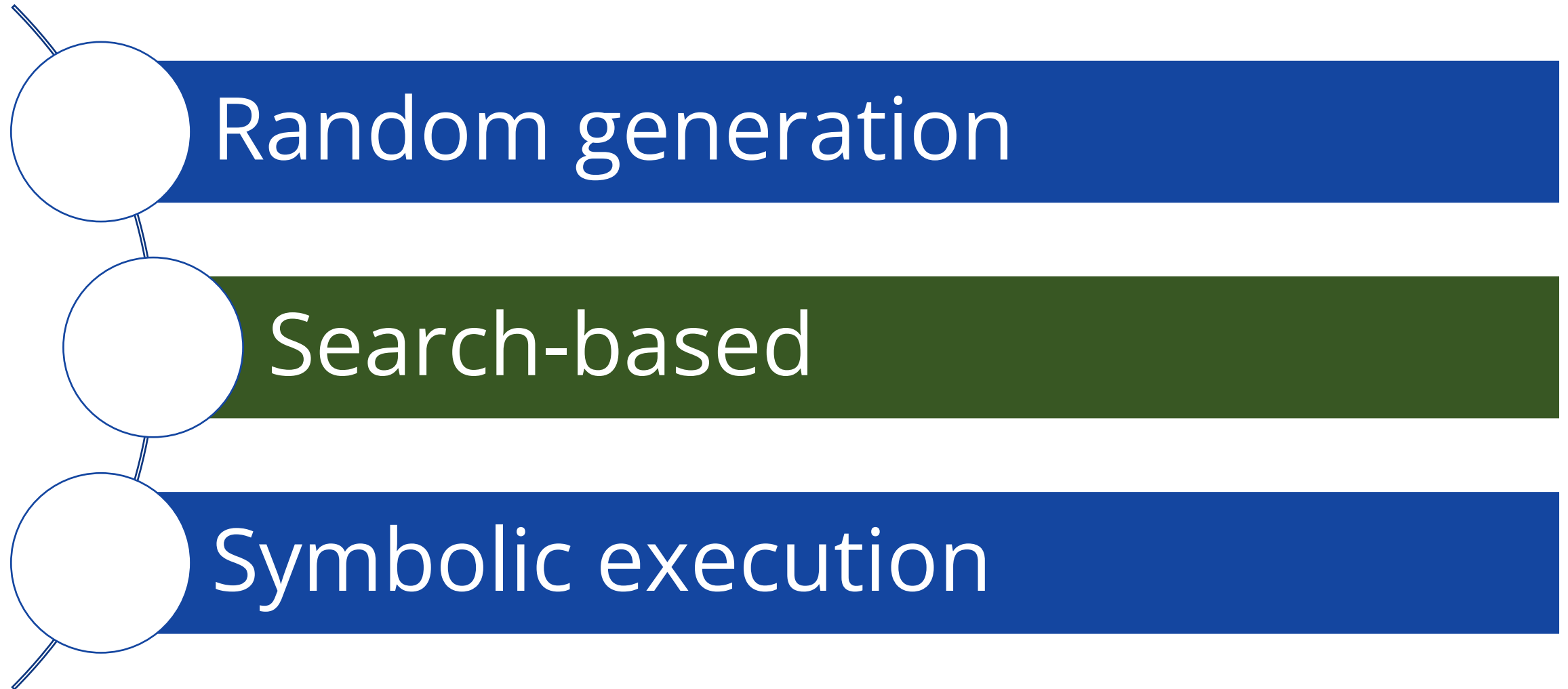
More information: [video](#)

Claessen et al. "QuickCheck: a lightweight tool for random testing of Haskell programs" ACM Sigplan Notices 46.4 (2011): 53-64

GraphWalker



Techniques



Search-based techniques

Search-based Software Engineering (SBSE)

- Metaheuristic algorithms
 - genetic alg., simulated annealing, hill climbing...
- Representing a problem as a search:
 - Search space: program structure + possible inputs
 - Objective function: reaching a test goal (e.g., covering all branches of decisions)

A tool for search-based test generation



- „Whole test suite generation”
 - All test goals are taken into account
 - Searches based on multiple metrics
 - E.g., high coverage with minimal test suite
- Specialties:
 - Minimizes test code, maintains readability
 - Uses sandbox for environment interaction

Basic techniques



Random generation

Search-based

Symbolic execution

Symbolic execution: the idea

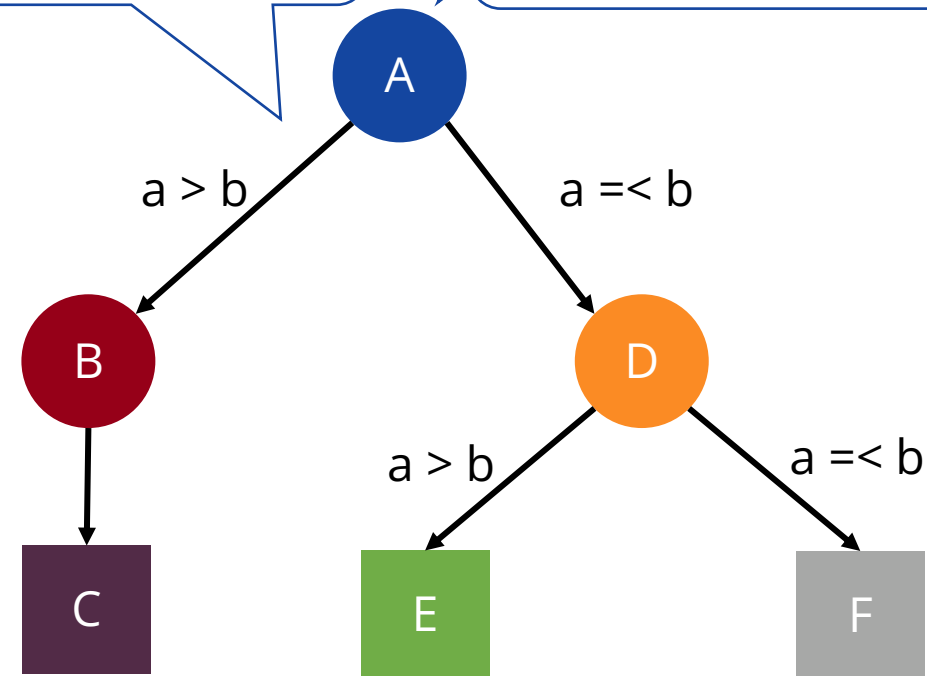
- **Static program analysis technique from the '70s**
 - Symbolic variables instead of concrete ones
 - Form constraints for each path with symbolic variables
 - Constraint solving (path feasibility)
 - Explore program states and paths (e.g. DFS)
- **Application for test generation**
 - A solution yields an input to execute a given path
- **Note:** SE tries to explore all paths (e.g. unrolls loops)
 - SE tree is not a CFG!
 - SE and abstract interpretation are different techniques!

Example: Static symbolic execution

```
int fun1(int a, int b){  
    if (a > b){  
        printf(ERROR_MSG);  
        return -1;  
    }  
    if (a > b)  
        return b*a + 5;  
    else  
        return (a+b) / 2;  
}
```

- Execute statement
- if: fork execution
- Select a branch

- Symbolic var.: a, b
- Statement: if (a > b)
- Path constraint (PC): true



After reaching a leaf, select a non-explored branch

- PC: $a \leq b$ AND $a > b$
- FALSE: not feasible

- PC: $a \leq b$ AND $a \leq b$
- Solve PC to get test inputs, e.g.
 - $a = 0, b = 0$

Static symbolic execution: detailed example

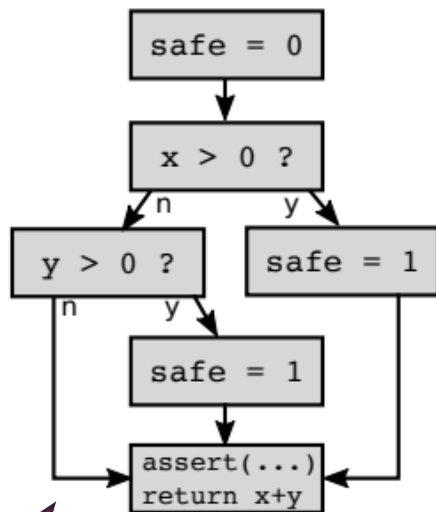
```
uint8_t nzsum(uint8_t x, uint8_t y) {
1.   int safe = 0;
2.   if (x > 0)
3.       safe = 1;
4.   else if (y > 0)
5.       safe = 1;
6.   assert(safe, "Error");
7.   return x + y;
}
```

SE tree

Execution state:

- Symbolic values of variables
- PC: path constraint

Executing a statement updates variables symbolically



CFG

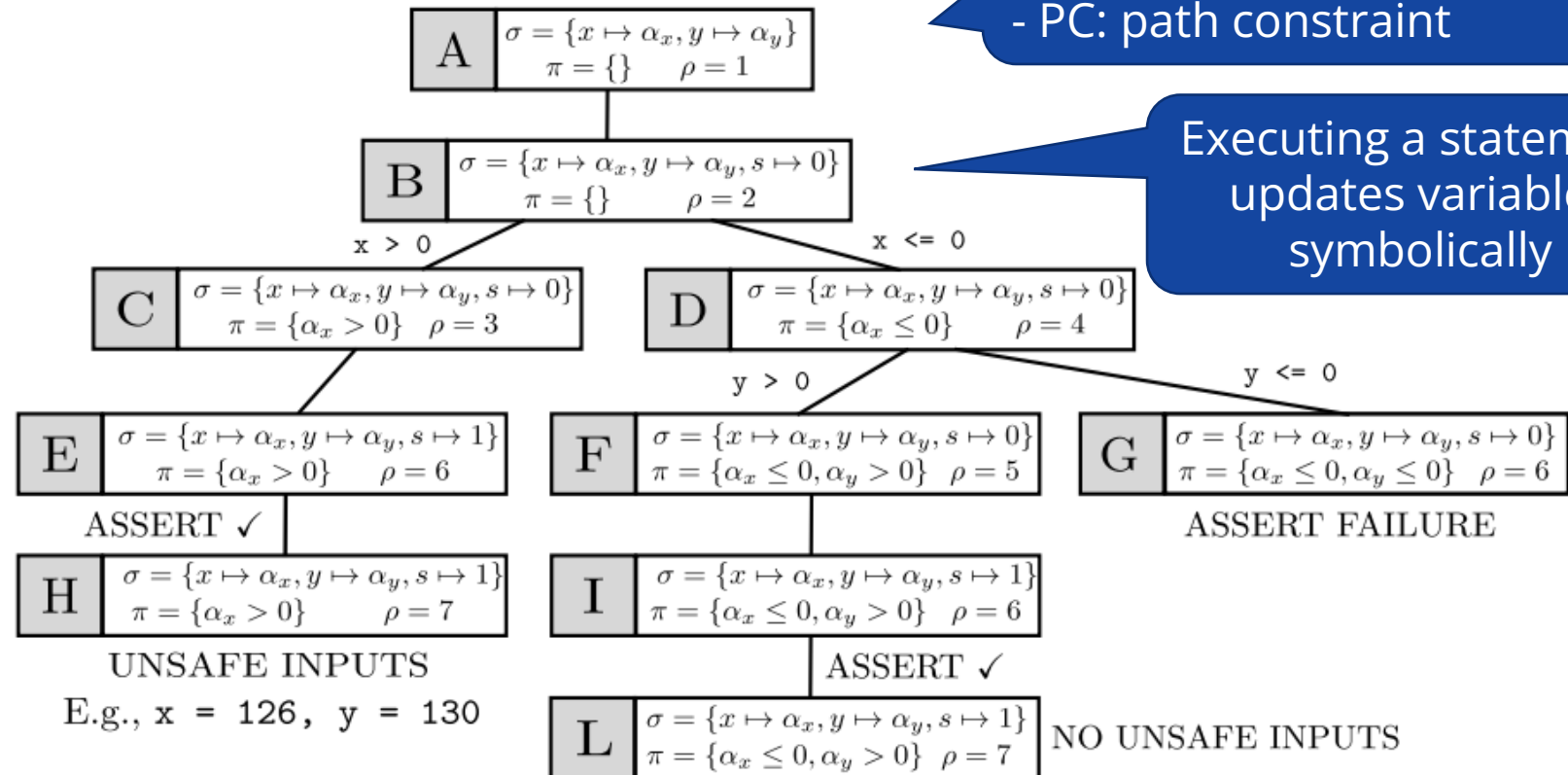


Figure 1: Introductory example: code and control flow graph (left) and symbolic execution tree (right).

EXERCISE: Building a SE tree

```

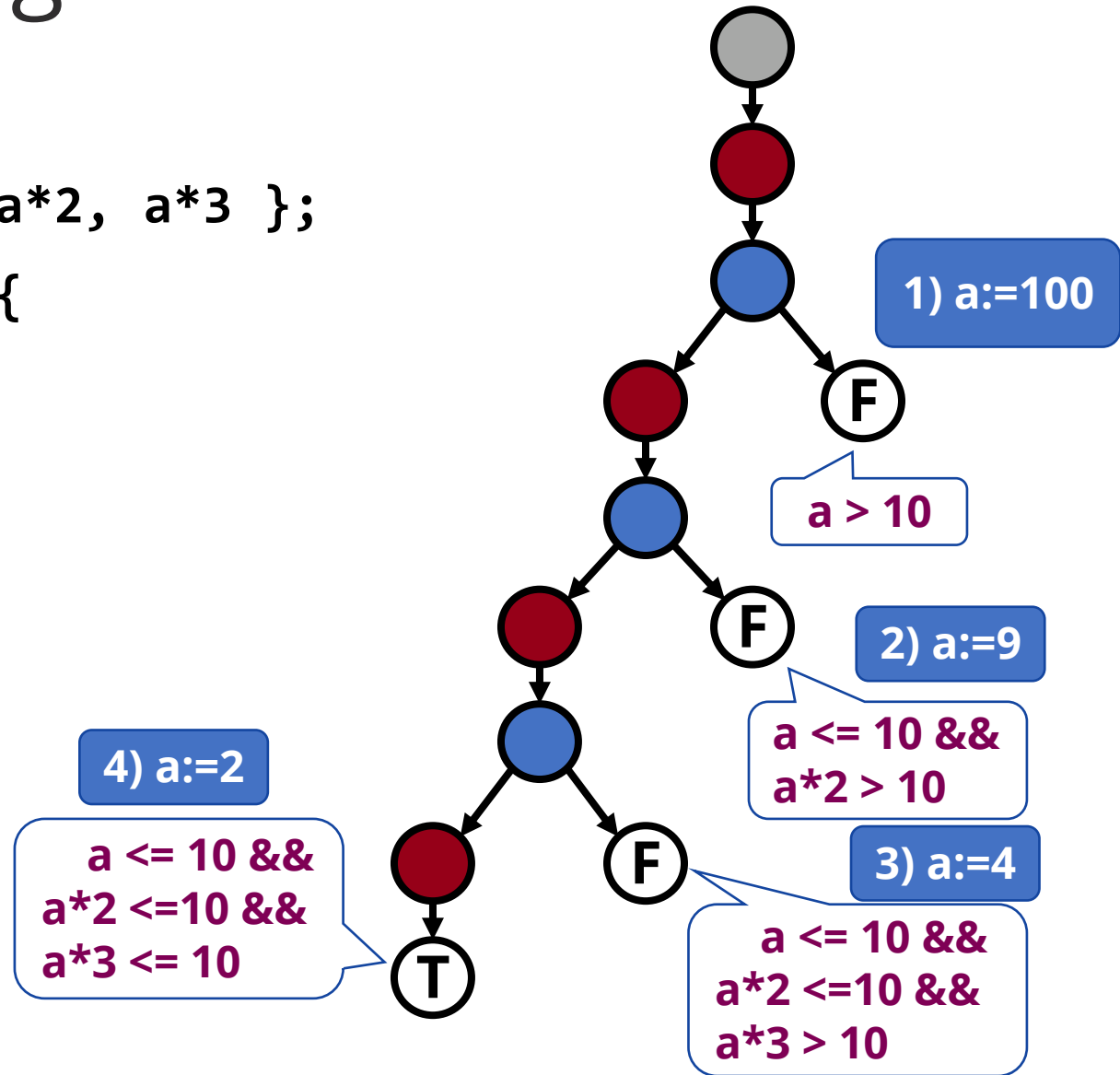
public bool fun2(int a) {
    int[] arr = new int[] { a, a*2, a*3 };
    for(int i = 0; i < 3; i++) {
        if(arr[i] > 10) {
            return false;
        }
    }
    return true;
}

```

4) a:=2

$a \leq 10 \wedge 8$

Build the SE tree of this method



Extending static symbolic execution

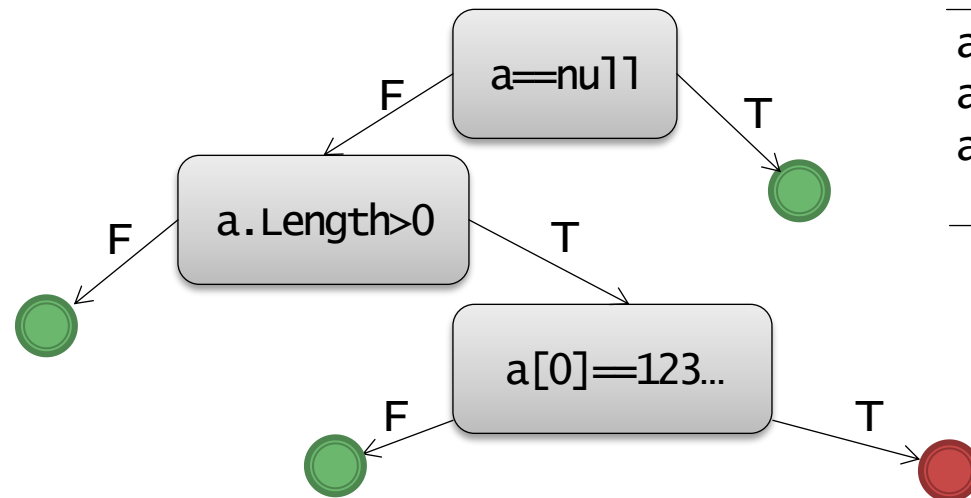
- **Static SE fails** in several cases, e.g.
 - Too long paths → too many constraints
 - Cannot decide if a path is really feasible or not
- **New century, new progress:**
 - Enough computing power (e.g., for SMT solvers)
 - New ideas, extensions, algorithms and *tools*
- **Idea:** mix symbolic with concrete executions
 - Dynamic Symbolic Execution (DSE) or
 - Concolic Testing (concrete + symbolic)

Dynamic symbolic execution

Code to generate inputs for:

```
void CoverMe(int[] a)
{
    if (a == null) return;
    if (a.Length > 0)
        if (a[0] == 1234567890)
            throw new Exception("bug");
}
```

Negated
condition



Choose next path		
Solve		Execute&Monitor
Constraints to solve	int[] a	Observed constraints
	null	a==null
a!=null	{}	a!=null && !(a.Length>0)
a!=null && a.Length>0	{0}	a!=null && a.Length>0 && a[0]!=1234567890
a!=null && a.Length>0 && a[0]==1234567890	{123..}	a!=null && a.Length>0 && a[0]==1234567890

Done: There is no path left.

Source: T. Xie, N. Tillmann, P. Lakshman:
Advances in Unit Testing: Theory and Practice

Tools available

Name	Platform	Language	Notes
KLEE	Linux	C (LLVM bitcode)	
Pex	Windows	.NET assembly	Included in Visual Studio (IntelliTest)
SAGE	Windows	x86 binary	Security testing, SaaS model
Jalangi2	*	JavaScript	
Symbolic PathFinder	*	Java	
Sonar Java SE	*	Java	https://github.com/SonarSource/sonar-java/tree/master/java-symbolic-execution

More tools: <http://mit.bme.hu/~micskeiz/pages/cbtg.html>

Parameterized Unit Testing

- **Idea: Using tests as specifications**

- Easy to understand, easy to check, etc.
- *But:* too specific (represents a single execution), oracle needed, etc.

- **Parameterized Unit Test (PUT)**

- Wrapper method for method/unit under test
 - Partitions the space of concrete test cases
- Main elements
 - Inputs of the unit
 - Assumptions for input space restriction (partitioning)
 - Call to the unit
 - General assertions for expected results (oracle)
- Serves as a **specification** → Test generators can use it

Example: Parameterized Unit Testing

```
/// The method reduces the quantity of the specified
/// product. The product is known to be NOT null, also
/// the sold amount is always more than zero. The method
/// has effects on the database, and returns the new
/// quantity of the product. If the quantity would be
/// negative, the method reduces the quantity to zero.
int ReduceQuantity(Product prod, int soldCount) { ... }
```

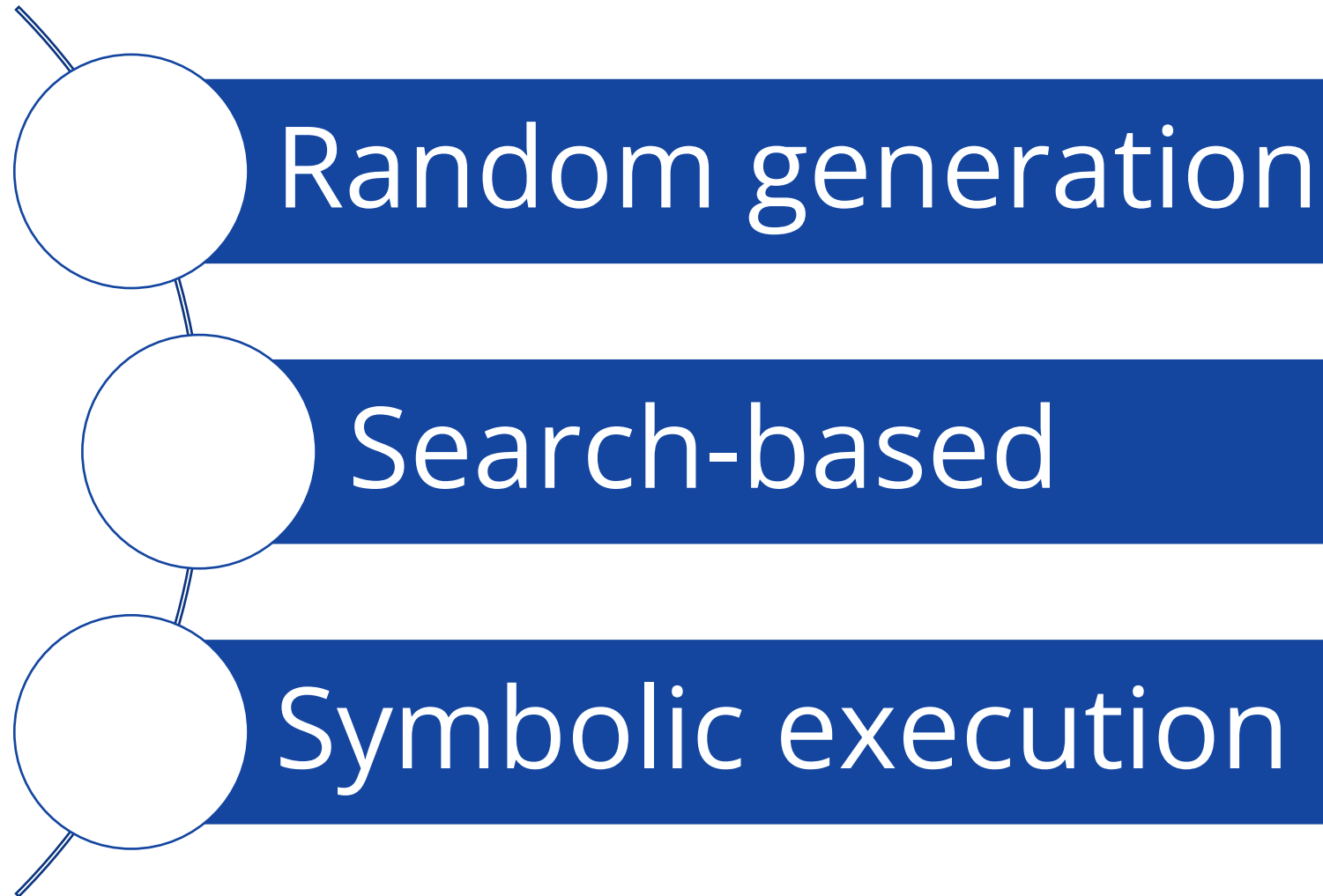
```
void ReduceQuantityPUT(Product prod, int soldCount) {
    // Assumptions
    Assume.IsTrue(prod != null);
    Assume.IsTrue(soldCount > 0);
    // Calling the UUT
    int newQuantity = StorageManager.ReduceQuantity(prod, soldCount);
    // Assertions
    Assert.IsTrue(newQuantity >= 0);
    int oldQuantity = StorageManager.GetQuantityFor(prod);
    Assert.IsTrue(newQuantity < oldQuantity);
}
```


Example: Parameterized Unit Testing

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/// The method reduces the quantity of the specified
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}
```

Basic techniques



Summary of techniques and goals

Two main questions:

- How to search/select test inputs?
- What to cover with selected inputs? How to evaluate?

Examples		What to cover with search? Evaluation?		
How to search?		<i>Code coverage</i>	<i>Crash, exception...</i>	<i>Property, assertion</i>
	<i>Random</i>	Coverage-based fuzzers	Robustness testing	Property-based testing
	<i>Search-based (Metaheuristics)</i>	SBSE tools (EvoSuite)		
	<i>Symbolic execution</i>	Static SE	SAGE (security)	Parametric Unit Test (Pex)

Evaluations

Applying these techniques on real code?

- SF100 benchmark (Java)
 - 100 projects selected from SourceForge
 - EvoSuite reaches branch coverage of 48%
 - Large deviations among projects

G. Fraser and A. Arcuri, "Sound Empirical Evidence in Software Testing," [ICSE 2013](#)

- A large-scale embedded system (C)
 - Execution of CREST and KLEE on a project of ABB
 - ~60% branch coverage reached
 - Fails and issues in several cases

X. Qu, B. Robinson: A Case Study of Concolic Testing Tools and Their Limitations, [ESEM 2011](#)

Are these techniques really that good?

- Does it help software developers?
 - 49 participants wrote and generated tests
 - Generated tests with high code coverage did not discover more injected failures

G. Fraser et al., "Does Automated White-Box Test Generation Really Help Software Testers?," [ISSTA 2013](#)

- Finding real faults
 - Defects4J: database of 357 issues from 5 projects
 - Tools evaluated: EvoSuite, Randoop, Agitar
 - Only found 55% of faults

S. Shamshiri et al., „Do automatically generated unit tests find real faults? An empirical study of effectiveness and challenges." [ASE 2015](#)

Comparison of test generator tools

- Various source code snippets to execute
 - Covering most important features of languages
- 363 Java/.NET snippets
 - Executed on 6 different tools
- Experience:
 - Huge difference in tools
 - Some snippets challenging for all tools

L. Csepentő, Z. Micskei: „Evaluating code-based test input generator tools,” [STVR 2017](#)

Comparison of test generator tools

