COMP5111 – Fundamentals of Software Testing and Analysis Random Testing (Feedback Directed)



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Some slides are adapted from https://randoop.github.io/randoop/files/thesis_talk_post.pdf

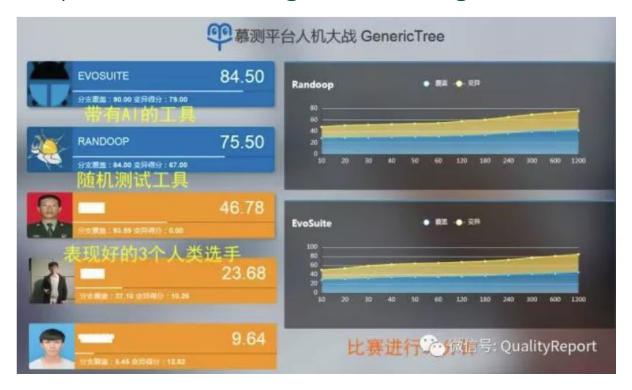


Dream of a developer...





Three selected top testers among a few thousand contest participants vs two test generation algorithms in 2017



After 30 mins

After 60 mins



After 90 mins



After 120 mins



Automatic Software Testing

- Random testing
- Symbolic analysis
- Concolic testing
- Search-based testing

Can unit tests be automatically generated without program spec?



Can generated tests detect real faults?

Random Testing

```
foo (int &x, int &y) {
 if (x>y) {
  X = X + Y;
  y = x - y;
  X = X - Y;
  if (x - y > 0) {
      assert (false); // bug
```

Options:

- Random input data generation
- Random user interaction sequence
- Random data selection from database
- Combinations of all above

```
1st trial: x = 1321, y = 456;

2nd trial: x = -2908, y = 89;
...

n^{th} trial: ...
```

Random Testing

- Mentioned first time by Glenford J. Myers in 1979.
- Popularly used by industry as fuzzing tests.

Slashdof News FOR NERDS. STUFF THAT MATTERS.

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Stories

Recent Popular Search



Posted by timothy of from the running-th

Microsoft uncovered more than 1,800 bugs in Office 2010 by running millions of 'fuzzing' tests using idling PCs.

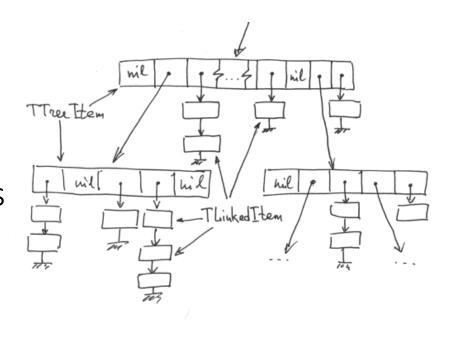
CWmike writes

"Microsoft uncovered more than 1,800 bugs in Office 2010 by tapping into the unused computing horsepower of idling PCs, a company security engineer said on Wednesday. Office developers found the bugs by running millions of 'fuzzing' tests, a practice employed by both software developers and security researchers, that searches for flaws by inserting data into file format parsers to see where programs fail by crashing. 'We found and fixed about 1,800 bugs in Office 2010's code,' said Tom Gallagher, senior security test lead with Microsoft's Trustworthy Computing group, who last week co-hosted a presentation on Microsoft's fuzzing efforts at the CanSecWest security conference. 'While a large number, it's important to note that that doesn't mean we found 1,800 security issues. We also want to fix things that are not



Random Unit Test Generators

- Challenge 1: Generate complex data structures
- Challenge 2: Avoid generating redundant tests
- Challenge 3: Generate test oracles (i.e., the assert statements)



A Polynomial Library

Dynamic data structure

```
class Poly {
 List <Mono> elements;
 Poly() { ... }
 Poly plus(Mono m) { ... }
 Poly mult(Poly p) { ... }
 Poly sum(Poly p) { ... }
 Poly deriv() { ... }
```

```
class Mono {
 int num, den, exp;
 Mono(int num, int den, int exp) {
```

How can we generate tests with complex data structures?

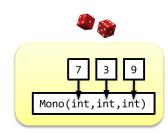
Insight: Complex data structures can be built incrementally from primitive values



Common Generation Strategy

operation	input	output
Mono(int,int,int)	3 ints	a new Mono
Poly()	none	a new Poly
Poly plus(Mono)	a Poly, a Mono	a new Poly

random terms



```
public void test1() {
   p = new Poly()
       .mult(new Poly());
   checkInvariant(p);
}
```

```
public void test2() {

  p = new Poly()
     .plus(new Mono(1,2,0));

  checkInvariant(p);
}
```

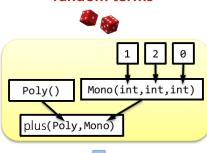
```
public void test3() {
    m = new Mono(7,3,9);
    checkInvariant(m);
}
```

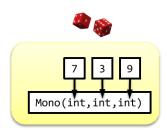
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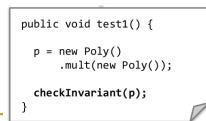
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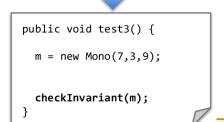
random terms







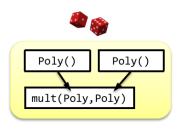
```
public void test2() {
   p = new Poly()
        .plus(new Mono(1,2,0));
   checkInvariant(p);
}
```

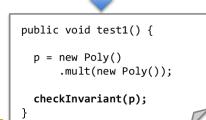


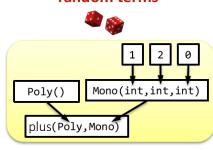
Common Generation Strategy

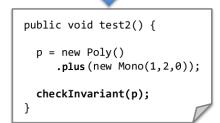
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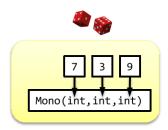
random terms

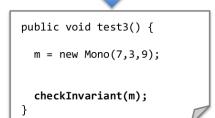












How can we build complex data structures mechanically?

Insight: Feed generated data structures

back to a repository



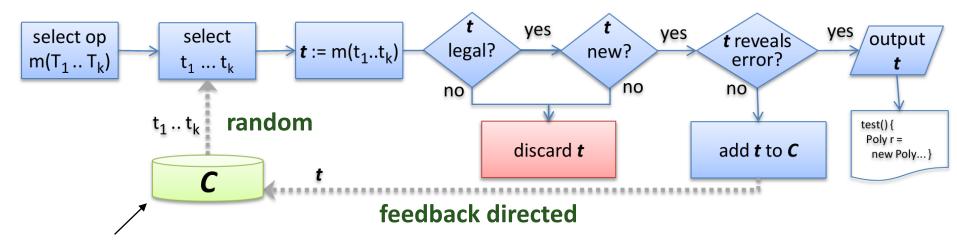
- Carlos Pacheco [ICSE 2007, ISSTA 2008]
- Randoop (https://code.google.com/p/randoop/)
- Able to reveal unknown faults in widely used libraries

distinct errors revealed (Java)

code base	Randoop	JPF (model checker)	JCrasher (random tester)
Sun JDK (272 classes,43KLOC)	8	0	1
Apache libraries (974 classes, 114KLOC)	6	1	0

distinct errors revealed (.NET)

code base	Randoop	symbolic execution unit test generator
.NET library (1439 classes, 185KLOC)	30	0



 C is a repository containing possible terms used by the Test Generator.

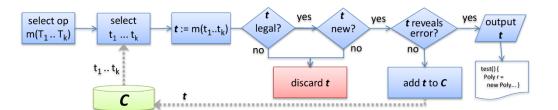
 Generate tests with simple inputs randomly from repository C.



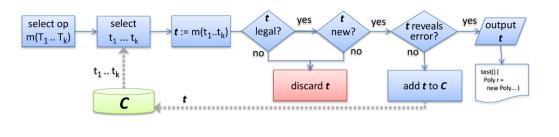


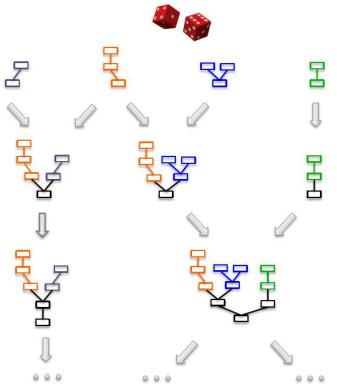






- Generate tests with simple inputs randomly from repository C.
- Build new test inputs incrementally from previous ones.





illegal input to Mono generates useless inputs illegal, repetitive num xexp den Mono(int,int,int) throws IllegalArgException

Mono(int num, int den, int exp) { ... } Expects den ≠ 0 and exp ≥ 0



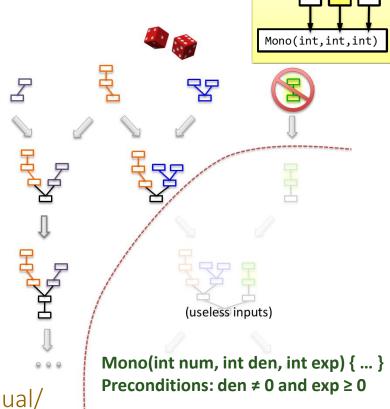
How can we avoid redundant tests?

Insight: Prune the input space with preconditions and equivalence



Pruning Input Space

- Executes inputs
- Discards the ones useless for extension
 - □ illegal, redundant
- Prune input space
 - Specify pre-conditions on method parameters
 - See method pre-conditions at https://randoop.github.io/randoop/manual/



Example: Mono(int, int, int)

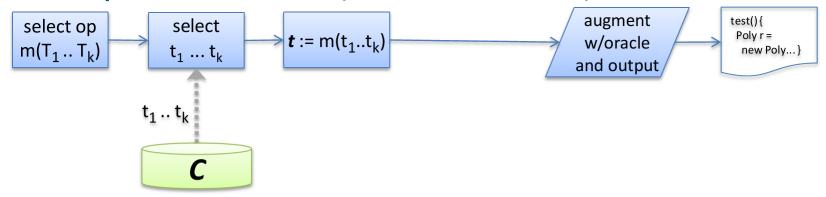
select op $m(T_1 ... T_k)$



component set of terms

C = {0, 1, 2, null, false, etc. }

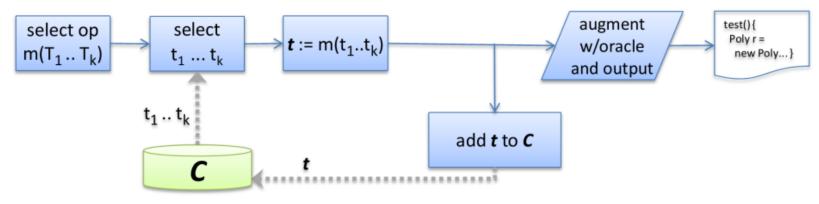
Example: Mono(int, int, int)



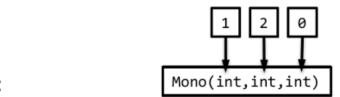
component set of terms



Example: Mono(int, int, int)

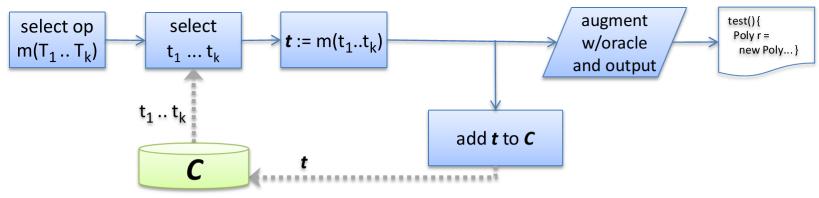


component set of terms



Example:

Example: Poly()



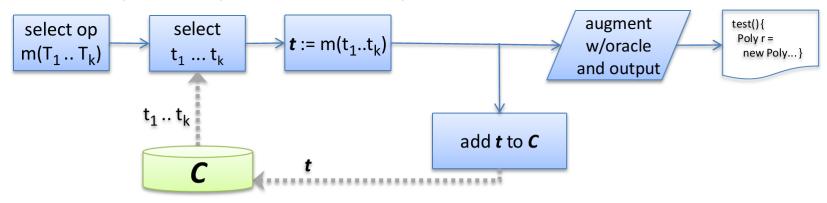
component set of terms

Example:

Poly()

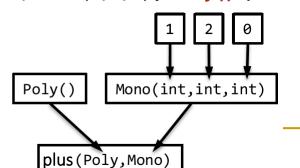
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Example: plus(Poly, Mono)

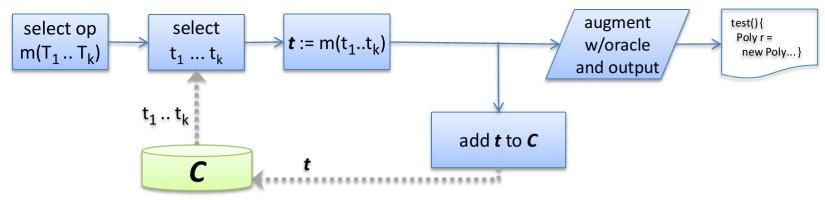


component set of terms

Example:

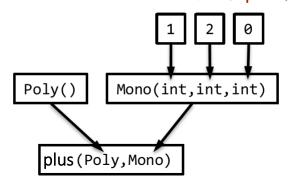


Example: plus(Poly, Mono)



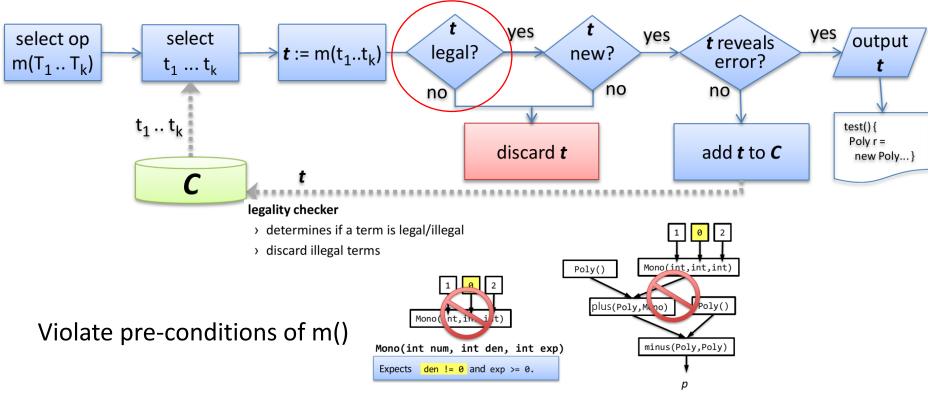
component set of terms

$$C = \{0, 1, 2, null, false, Mono(1,2,0), Poly(), plus(Poly(), Mono(1,2,0)) \}$$



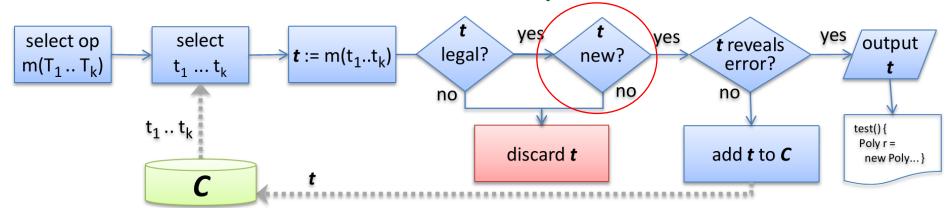
Example:

Guided Generator - Legality



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Guided Generator - Equivalence



- Determines if two terms are equivalent
 - e.g., Mono(1,2,1) and Mono(2,4,1)
- Discard a term if equivalent to one in C
- Implement equivalence as t.equals(t')

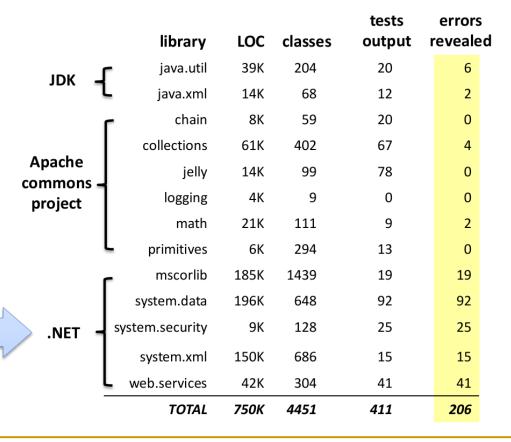
Applied Randoop to 13 libraries

- > built-in oracles
- > heuristic guidance
- default time limit(2 minutes/library)

Outputs one test per violating method

.NET libraries specification:

"no method should throw NPEs, assertion violations, or IllegalMemAccess exception"



How can we define test oracles?

Insight: Define generic oracles and regression oracles



Five Built-in Oracles

- Contracts over Object.equals()
 - Reflexivity: o.equals(o) == true
 - Symmetry: o1.equals(o2) == o2.equals(o1)
 - Transitivity: o1.equals(o2) && o2.equals(o3) implies o1.equals(o3)
 - Equals to null: o.equals(null) == false
 - It does not throw an exception

Five Built-in Oracles

- Contracts over Object.hashCode()
 - o1.equals(o2) == true implies o1.hashCode() == o2.hashCode()
 - It does not throw an exception
- Contracts over Object.toString()
 - It does not return null
 - It does not throw an exception

Five Built-in Oracles

- Contracts over Object.clone()
 - It does not throw an exception, including CloneNotSupportedException
 - It does not throw an exception

Five Built-in Oracles

- Contracts over Comparable.compareTo() and Comparator.compare()
 - Reflexivity: o.compareTo(o) == 0
 - \square Anti-symmetry: sgn(o1.compareTo(o2)) == -sgn(o2.compareTo(o1))
 - Transitivity: o1.compareTo(o2)>0 && o2.compareTo(o3)>0 implies o1.compareTo(o3)>0
 - Substitutability of equals: x.compareTo(y) == 0 implies sgn(x.compareTo(z)) == sgn(y.compareTo(z))
 - Consistency with equals(): x.compareTo(y)==0 implies x.equals(y)
 - It does not throw exception

Regression Test Oracle

Generate assertion using the current test output

```
public class ClassExampleWithFailure {
  public static int twice(int x) { return x+x;}

public static int foo(int x, int y) {
  int z = twice(x);
  if (z == 144 && y > 20) {
    assert(false); // assert failure
  }
  return y*z;
}

**Test

public void test022() throws Throwable {
    ...
  int i2 = ClassExampleWithFailure.foo(24832, 388);
    org.junit.Assert.assertTrue(i2 == 19269632);
  }
}
```

Evaluation and Industry Adoption

Comparison

JDK

- > 6 methods that create objects violating reflexivity of equality
- > 2 well-formed XML objects cause hashCode/toString NPEs

Apache

> 6 constructors leave fields unset, leading to NPEs

.NET

- > 175 methods throw forbidden exceptions
- > 7 methods that violate reflexivity of equals

.NET

> library hangs given legal sequence of calls

without guidance

none revealed

66% fewer revealed

70% fewer revealed

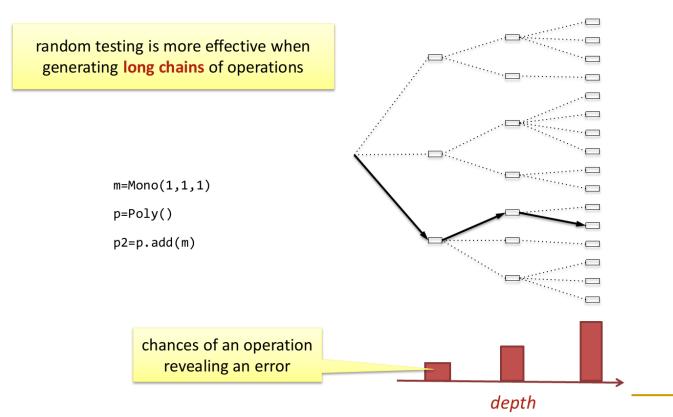
not revealed

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Comparison

- Why is Randoop more effective?
 - Prune useless inputs
 - Generates longer tests
 - Regression oracles

Test Length vs. Test Effectiveness



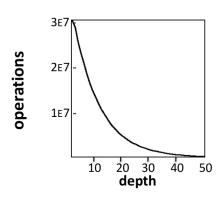
An Experiment

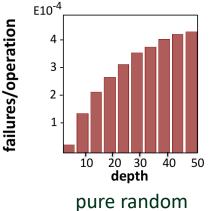
- Pure random testing
 - Start from empty sequence
 - Take random steps
 - Restart if error or exception
- Exercise 10M operations per library
 - Take several days

library	classes	LOC
java.util	204	39K
collections	402	61K
primitives	294	6K
trove	336	87K
jace	164	51K

Results

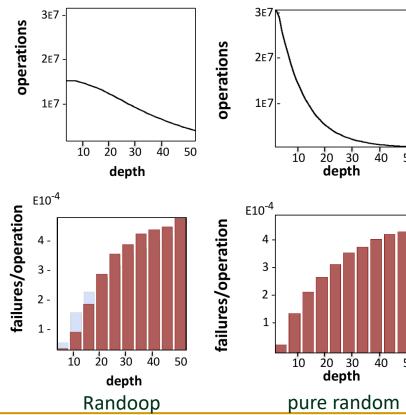
- Fail to create long chains (due to exceptions)
- Failure rate is higher at greater depths
- Ineffective → performs most operations where failure rate is the lowest





Using Randoop (i.e., Feedback Directed Random)

- More tests with longer chains
- Able to reveal more failures (under the same budget)



Comparison

library	random walk	Randoop only leg.	Randoop
java.util	20	21	27
collections	28	37	48
primitives	16	13	19
trove	20	27	27
jace	15	15	26
TOTAL	99	113	147

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Include also

equivalence checking

Failures

detected

Other findings ...



Case study [ISSTA 2008]

- Microsoft test team
- Randoop (.NET version)
- Applied to highly-tested library
 - □ Tested over 5 years by 40 engineers

Findings: revealed more errors in 15 hours than the team typically discovers in 1 person-year of effort

- Can generated tests detect real bugs?
- Is automated test generation cost effective?

Case Study Statistics

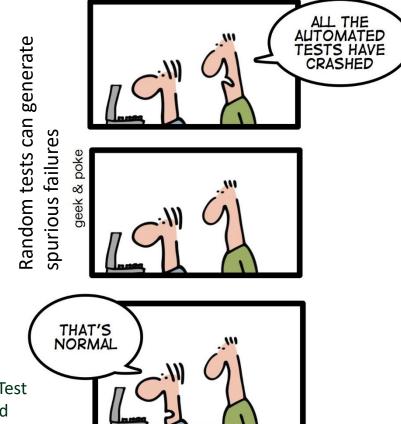
- Facts
 - Human time interacting with Randoop: 15 hours
 - CPU time running Randoop: 150 hours
 - Total distinct method sequences: 4 million
 - New errors revealed: 30
- Randoop
 - 30 new errors in 15 hours of human effort
 - 1 new error for ½ hour effort
- Existing team methods
 - 20 new errors per year
 - 1 new error for 100 hours human effort

Interacting with Randoop
Inspecting the resulting tests
Discarding redundant failures

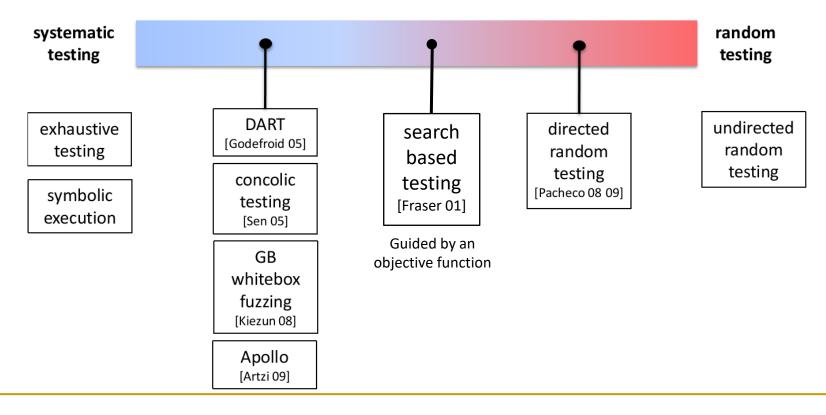
Limitations

- Can generate spurious failures
 - Unaware of implicit pre-conditions*
- Weak assert statements
- Low coverage (< 50%) for reactive programs
 - Android apps, GUI applications
 - Not driven by coverage

*Mijung Kim, Shing-Chi Cheung, Sunghun Kim. Which Generated Test Failures Are Fault Revealing? Prioritizing Failures Based on Inferred Precondition Violations using PAF. In ESEC/FSE 2018.



Spectrum of Testing Techniques



Using Randoop

https://randoop.github.io/randoop/manual/index.html

Video: https://www.youtube.com/watch?v=nPdb-72-EJY

Further Readings

- Other popular random testing tools
 - Jubula for GUI testing (http://www.eclipse.org/jubula/)
 - Monkey & Stoat for Android software testing (http://developer.android.com/tools/help/monkey.html) (https://github.com/tingsu/Stoat)
 - Sapienz for Android random testing (https://github.com/Rhapsod/sapiens)
 - Facebook prototype to be replaced by an official release in 2019
- API invariance inference
 - Robillard, M.P.; Bodden, E.; Kawrykow, D.; Mezini, M.; Ratchford, T.,
 "Automated API Property Inference Techniques," *IEEE Transactions on Software Engineering*, vol.39, no.5, pp.613-637, May 2013.