# Software Testing Basics

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#### Testing: basic concepts

- Test case (or, simply test): an execution of the software with a given test input, including:
  - Input values
  - Sometimes include execution steps
  - Expected outputs (test oracle)
- Test suite: a finite set of tests
  - Usually can be run together in sequence
- Test adequacy: a measurement to evaluate the test quality
  - Such as code coverage



### Testing: levels

- Unit Testing
  - Test each single module in isolation
- Integration Testing
  - Test the interaction between modules
- System Testing
  - Test the system as a whole, by developers
- Acceptance Testing
  - Validate the system against user requirements, by customers with no formal test cases

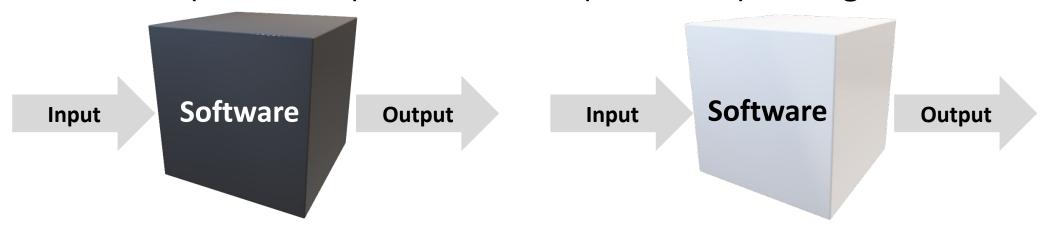
System /
Acceptance
Testing

**Integration Testing** 

**Unit Testing** 

#### Types of test generation

Black-box (functional) vs. white-box (structural) testing



- Black-box test generation: generates tests based on the functionality of the program
- White-box test generation: generates tests based on the source-code structure of the program

## Today's focus

- Unit testing: involves testing individual units (e.g., methods or classes) of a software to ensure that each part is correct, typically
  - Unit level
  - White-box
  - Deterministic
  - . . .

- Fuzz testing (fuzzing): involves providing invalid, unexpected, or random data as inputs to a software, typically
  - System level
  - Black-box
  - Non-deterministic
  - . . .

#### This class

- Unit Testing
  - Feedback-directed Random Test Generation (ICSE'07)
- Fuzz Testing
  - Finding and Understanding Bugs in C Compilers (PLDI'11)
  - Fuzzing with Code Fragments (SEC'12)
  - Compiler Validation via Equivalence Modulo Inputs (PLDI'14)
  - AFL: American Fuzzy Lop (<a href="https://github.com/google/AFL">https://github.com/google/AFL</a>)
- LLM-based Fuzz Testing
  - Large Language Models are Zero-Shot Fuzzers: Fuzzing Deep-Learning Libraries via Large Language Models (ISSTA'23)

## Problem: unit test generation

```
Program under test:
public class Math{
  public static int sum(int a, int b) {
    return a+b;
  }
  ...
}
```

```
Example JUnit test:
public class MathTest{
    @Test
    public void testSum () {
        int a=1;
        int b=1;
        int c=Math.sum(a, b);
        assertEquals(2,c);
        Test oracle
    }
    ...
}
```

#### Is this an important problem?



# How to perform random white-box test generation?

```
public class HashSet extends Set{
  public boolean add(Object o) {...}
  public boolean remove(Object o) {...}
  public boolean isEmpty() {...}
  public boolean equals(Object o) {...}
  ...
}
```

Generation

#### **Program under test**

- Need to generate a random sequence of invocations, where each requires
  - A random method
  - Some random arguments
  - A random receiver object
    - Not required for static methods

```
Set s = new HashSet();
s.add("hi");
```

#### Generated test t1

```
Set s = new HashSet();
s.add("hi");
s.remove(null);
```

#### Generated test t2

```
Set s = new HashSet();
s.isEmpty();
s.remove("no");
s.isEmpty();
s.add("no");
s.isEmpty();
s.isEmpty();
...
```

#### Generated test t3

... 8

# Random method-sequence generation: limitations

- Does not have test oracles
  - E.g., an ideal test oracle for the test below: assertEquals(1, s.size())
- Cannot generate complex tests
  - E.g., the arguments of some method invocations can be generated by other method invocations
- Can have many redundant&illegal tests

```
Set s = new HashSet();
s.isEmpty();
s.remove("no");
s.isEmpty();
s.add("no");
s.isEmpty();
s.isEmpty();
```

A random test

# Random method-sequence generation: redundant&illegal tests

```
1. Useful test:
Set s = new HashSet();
s.add("hi");
```

```
2. Redundant test:
Set s = new HashSet();
s.add("hi");
s.isEmpty();
```

Should not output

```
3. Useful test:
Date d = new Date(2006, 2, 14);
```

```
4. Illegal test:
Date d = new Date(2006, 2, 14);
d.setMonth(-1); // pre argument >= 0
```

#### Should not output

```
5. Illegal test:
Date d = new Date(2006, 2, 14);
d.setMonth(-1); // pre: argument >= 0
d.setDay(5);
```

# Randoop: feedback-directed (adaptive) random test generation

- Use code contracts as test oracles
- Build test inputs incrementally
  - New test inputs extend previous ones
  - In this context, a test input is a method sequence
- As soon as a test is created, use its execution results to guide generation
  - away from redundant or illegal method sequences
  - towards sequences that create new object states

## Randoop input/output

#### • Input:

- Classes under test
- Time limit
- Set of contracts
  - Method contracts (e.g. "o.hashCode() throws no exception")
  - Object invariants (e.g. "o.equals(o) == true")
- Output: contract-violating test cases

```
HashMap h = new HashMap();
Collection c = h.values();
Object[] a = c.toArray();
LinkedList l = new LinkedList();
l.addFirst(a);
TreeSet t = new TreeSet(l);
Set u = Collections.unmodifiableSet(t);
assertTrue(u.equals(u));
```



fails on Sun's JDK 1.5/1.6 when executed

#### Randoop: algorithm

- Seed value pool for primitive types
  - pool = { **0, 1, true, false, "hi", null** ... }
- Do until time limit expires:
  - Create a new sequence
    - Randomly pick a method call  $m(T_1...T_k)/T_{ret}$
    - For each input parameter of type  $T_i$ , randomly pick a sequence  $S_i$  from the value pool that constructs an object  $v_i$  of type  $T_i$
    - Create new sequence  $S_{new} = S_1; ...; S_k; T_{ret} v_{new} = m(v_1...v_k);$
    - if **S**<sub>new</sub> was previously created (lexically), go to first step
  - Classify the new sequence **S**<sub>new</sub>
    - May discard, output as test case, or add to pool



```
Program under test:
                                                    Test1:
                                                    B b1=new B(0);
public class A{
 public A() {...}
 public B m1(A a1) {...}
public class B{
 public B(int i) -{-...}
 public void m2(B b, A a) {...}
 Value pool:
 S1: B b1=new B(0);
         0,-1, null, "hi", ...}
                                                                                      14
```

```
→ Method
 Parameter
 ➤ Receiver object
```

```
Program under test:
                                                     Test1:
                                                     B b1=new B(0);
public class A{
 public A() - (-.-) - -.
 public B m1 (A a1) {...}
                                                     Test2:
                                                     A a1=new A();
public class B{
 public B(int i) {...}
 public void m2(B b, A a) {...}
 Value pool:
 S2: A a1=new A();
 S1: B b1 = new B(0);
          {0, 1, null, "hi", ...}
                                                                                       15
```

```
--> Method
--> Parameter
--> Receiver object
```

```
Program under test:
public class A{
 public A() {...}
 public B m1 (A a1) -{-...}
public class B{
 public B(int i) {...}
 public void m2(B b, A a) {...}
 Value pool:
 S3: A a1=new A();
     B b2=a1.m1(a1);
 S2: A a1 --- A();
 S1: B b1 = new B(0);
```

{0, 1, null, "hi", ...}

```
Test1:
B b1=new B(0);
```

```
Test2:
A a1=new A();
```

```
Test3:
A a1=new A(); //reused from s2
B b2=a1.m1(a1);
```

```
- - - > Method- - - > Parameter- - - > Receiver object
```

```
Program under test:
public class A{
  public A() {...}
  public B m1(A a1) {...}
}
public class B{
  public B(int i) {...}
  public void m2(B b, A a)-{...}
}
```

```
Value pool:
S3: A al=new A();
B b2=al.ml(al);
S2: A al=new A();
S1: B b1=new B(0);

{0, 1, null, "hi",...}
```

```
<u>Test1:</u>
B b1=new B(0);
```

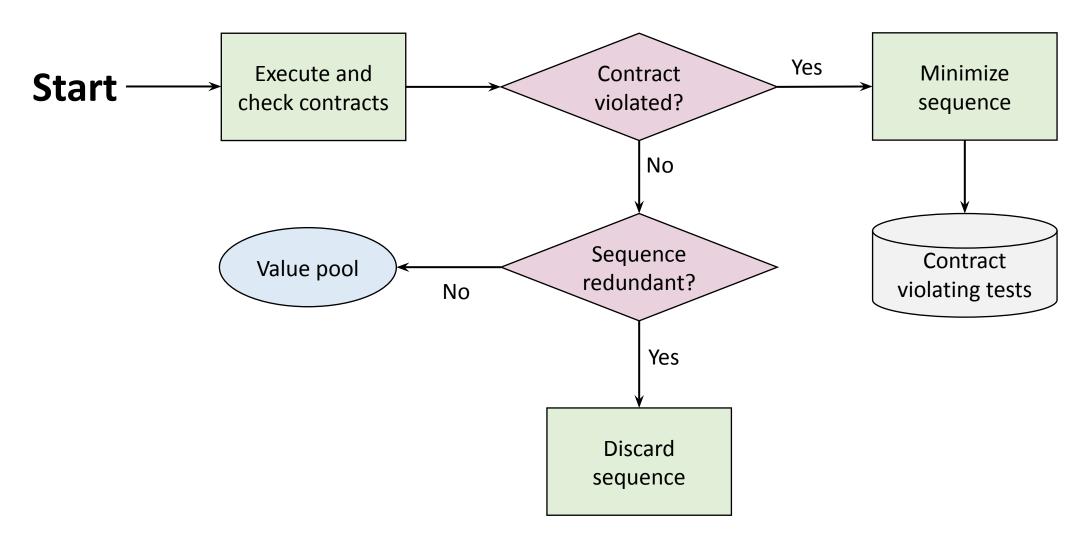
```
Test2:
A a1=new A();
```

```
Test3:
A a1=new A();
B b2=a1.m1(a1);
```

```
Test4:
B b1=new B(0); //reused from s1
A a1=new A();
B b2=a1.m1(a1); //reused from s3
b1.m2(b2, a1);
```

```
•••
```

## Classifying a sequence



#### Redundant sequences

- During generation, maintain a set of all objects created
- A sequence is redundant if all the objects created during its execution are members of the above set (using *equals* to compare)
- Could also use more sophisticated state equivalence methods
  - E.g. heap canonicalization used in model checkers

#### Tool support

#### • Input:

- An assembly (for .NET) or a list of classes (for Java)
- Generation time limit
- Optional: a set of contracts to augment default contracts
- Output: a test suite (JUnit or Nunit) containing
  - Contract-violating test cases
  - Normal-behavior test cases



#### Randoop outputs oracles

Oracle for contract-violating tests:

```
Object o = new Object();
LinkedList l = new LinkedList();
l.addFirst(o);
TreeSet t = new TreeSet(l);
Set u = Collections.unmodifiableSet(t);
assertTrue(u.equals(u));//expected to fail
```

Oracle for normal-behavior tests (regression tests):

```
Object o = new Object();
LinkedList l = new LinkedList();
l.addFirst(o);

1.add(o);
assertEquals(2, l.size());//expected to pass
assertEquals(false,l.isEmpty());//expected to pass
```

#### Some Randoop options

Avoid use of null

```
Statically:
Object o = new Object();
LinkedList l = new
LinkedList();
1.add(null);
```

```
Dynamically:
Object o = returnNull();
LinkedList l = new
LinkedList();
1.add(o);
```

- Bias random selection
  - Favor shorter sequences
  - Favor methods that have been less covered
  - Use constants mined from source code
- Source code available:
  - https://randoop.github.io/randoop/

# Code coverage by Randoop

Data structure programs	Time (s)	Branch cov.
Bounded stack (30 LOC)	1	100%
Unbounded stack (59 LOC)	1	100%
BS Tree (91 LOC)	1	96%
Binomial heap (309 LOC)	1	84%
Linked list (253 LOC)	1	100%
Tree map (370 LOC)	1	81%
Heap array (71 LOC)	1	100%

## Bug detection by Randoop: subjects

Subjects	LOC	Classes
JDK (2 libraries) (java.util, javax.xml)	53K	272
Apache commons (6 libraries) (logging, primitives, chain, jelly, math, collections)	114K	974
.Net libraries (6 libraries)	615K	3455

## Bug detection by Randoop: methodology

- Ran Randoop on each library
  - Used default time limit (2 minutes)
- Contracts:
  - o.equals(o)==true
  - o.equals(o) throws no exception
  - o.hashCode() throws no exception
  - o.toString() throw no exception
  - No null inputs and:
    - Java: No NPEs
    - .NET: No NPEs, out-of-bounds, of illegal state exceptions

# Bug detection by Randoop: subjects

Subjects	Failed tests		Error-revealing tests	Distinct errors
JDK	613	32	29	8
Apache commons	3,044	187	29	6
.Net framework	543	205	196	196
Total	4,200	424	254	210

#### Errors found: examples

- JDK Collections classes have 4 methods that create objects violating
   o.equals(o) contract
- Javax.xml creates objects that cause hashCode and toString to crash, even though objects are well-formed XML constructs
- Apache libraries have constructors that leave fields unset, leading to NPE on calls of equals, hashCode and toString (this only counts as one bug)
- Net framework has at least 175 methods that throw an exception forbidden by the library specification (NPE, out-of-bounds, of illegal state exception)
- .Net framework has 8 methods that violate o.equals(o)
- Net framework loops forever on a legal but unexpected input

#### Regression testing scenario

- Randoop can create regression oracles
- Generated test cases using JDK 1.5
  - Randoop generated 41K regression test cases
- Ran resulting test cases on
  - JDK 1.6 Beta
    - 25 test cases failed
  - Sun's implementation of the JDK
    - 73 test cases failed
  - Failing test cases pointed to 12 distinct errors
  - These errors were not found by the extensive compliance test suite that Sun provides to JDK developers

```
Object o = new Object();
LinkedList l = new LinkedList();
l.addFirst(o);
l.add(o);
assertEquals(2, l.size());//expected to pass
assertEquals(false,l.isEmpty());//expected to pass
```

## Randoop: applications







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#### Fuzz testing

#### ./Program < /dev/random



- Fuzzing strategies
  - Mutation-based
  - Generation-based
  - Learning-based

- Feedback guide
  - New coverage?
  - Shorter execution?
  - Valid input?

- Targeted programs
  - Binaries
  - Compilers
  - Browsers
  - DB systems
  - ML systems
  - ..

#### Generation-based fuzzing

Create test inputs based on predefined structure/grammar

```
<start> ::= <expr>
<expr> ::= <term> + <expr> | <term> - <expr> | <term>
<term> ::= <term> * <factor> | <term> / <factor> | <factor>
<factor> ::= +<factor> | -<factor> | (<expr>) | <int> | <int>.<
<int> ::= <digit><int> | <digit>
<digit> ::= 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9
```

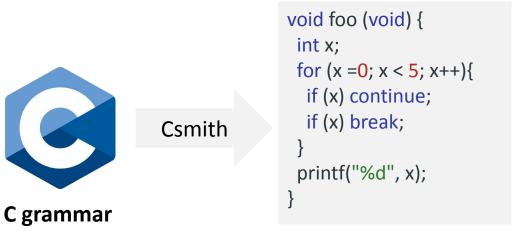
Generate

**Javascript** 

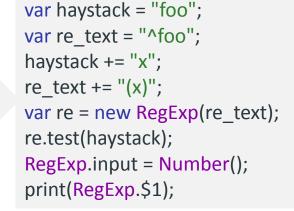
grammar

```
(-+++((1+(+7-(-1*(++-+7.7--+-4.0)))))*+--4--(6)+64)8.2-27-
-9 / +((+9 * --2 + --+-+-((-1 * +(8 - 5 - 6)) * (-((-+(((+(4))))) - ++4) /
+(-+--((5.6 - --(3 * -1.8 * +(6 * +-(((-(-6) * ---+6)) / +--(+-+-7 * (-0 *
(+(((((2)) + 8 - 3 - ++9.0 + ---(--+7 / (1 / +++6.37) + (1) / 482) /
+++-+0)))) * -+5 + 7.513)))) - ...
```

#### **Arithmetic expression grammar**



**Arithmetic expression** 



C program triggering an LLVM bug



JS program crashing Mozilla

### Generation-based fuzzing: examples

- Finding and Understanding Bugs in C Compilers (PLDI'11)
  - Targeting C compilers
  - Cited for 1,000+ times
  - 400+ GCC/LLVM bugs found





- Fuzzing with Code Fragments (SEC'12)
  - Targeting JS browsers/engines
  - Cited for 400+ times
  - USD 50,000+ bug bounties in the first month
  - 2,000+ bugs found for Mozilla Firefox, Google Chrome, and Microsoft Edge to date



#### Mutation-based fuzzing

 Apply small mutations on high-quality seed inputs to generate more test inputs

```
int a, b, c, d, e;
int main() {
                                                           int a, b, c, d, e;
 for (b = 4; b > -30; b--)
                                                           int main() {
  for (; c;)
                                                            for (b = 4; b > -30; b--)
   for (;;) {
                                                              for (; c;)
                                         FMI
     b++;
                                                               for (;;) {
     e = a > 2147483647 - b:
                                                                <del>b++;</del>
     if (d) break;
                                                                e = a > 2147483647 - b:
                                                                if (d) break;
 return 0;
                                                             return 0;
```

: not executed

- Structured mutation not generalizable?
- Mutation at the binary level!

**AFL** 

### Mutation-based fuzzing: examples

- Compiler Validation via Equivalence Modulo Inputs (PLDI'14)
  - Insight: EMI takes existing input programs and generates equivalent variants on a particular set of inputs (by removing unexecuted statements) for compiler fuzzing
  - 147 confirmed bugs found in the paper
  - Found 1,000+ LLVM/GCC bugs together with follow-up work
- AFL: American Fuzzy Lop (<a href="https://github.com/google/AFL">https://github.com/google/AFL</a>)
  - The pioneer binary fuzzing tool leveraging coverage feedback
  - Highly scalable and generalizable due to the practical design
  - Found numerous bugs in real-world software systems











#### This class

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### How can we fuzz test Deep Learning libraries?

Deep Learning (DL) libraries 🎓 🖒 serve as the fundamental building block for all DL pipelines

```
d = 3
fold = nn.Fold(output_size=(4,
                                                               all spatial dimensions
                                                                                           torch.nn.Fold
5), kernel_size=(2, 2))
                                                                                          (output_size, kernel_size,
                                                                                          dilation, padding, stride)
input = torch.randn(1, 3 * 2 *
                                                                Input Shape
                                                                                                         SATISFY
                                                                                                                        Output Shape
                                                                                          MUST.
                                                          (N,C*∏(kernel_size), L)
                                                                                                                     (N,C,output size[0],..)
2, 12)
                                                                        \color{red} \color{red} \textbf{output\_size}[d] + 2 \times \textbf{padding}[d] - dilation[d] \times (\textbf{kernel\_size}[d] - 1
output = fold(input)
```

#### **Challenges:**

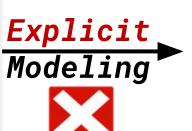
- Python syntax/semantics
- Complex tensor computation constraints
- Other implicit/explicit API constraints

### Large Language Models (LLMs) for fuzzing!

They are trained on trillions of open-source code tokens and can autoregressively generate human-like code!



Traditional fuzzing techniques (generation- & mutation-based)





- Language constraints
- Tensor shape constraints
- Other API constraints



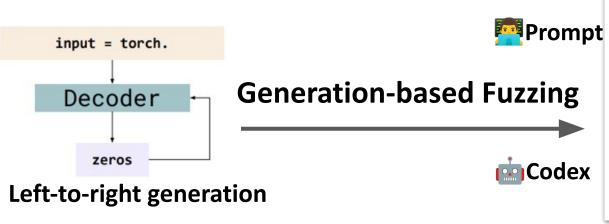




Trained on >400K
Tensorflow/PyTorch projects

Directly use LLMs to generate programs satisfying *heterogeneous* constraints for fuzzing DL libraries and beyond!

### LLMs for fuzzing: design



```
Task 1: Import TensorFlow 2.10.0 target library

Task 2: Generate input data

Task 3: Call the API tf.nn.conv2d(input,filters,strides,

padding,data_format='NHWC',dilations=None,name=None)

import tensorflow as tf

target API signature

tf.__version__
input = tf.Variable(tf.random.normal([1, 10, 10, 1]))

filter = tf.Variable(tf.random.normal([3, 3, 1, 1]))

op = tf.nn.conv2d(input, filter, strides=[1, 1, 1, 1], padding='VALID')
```

```
Encoder

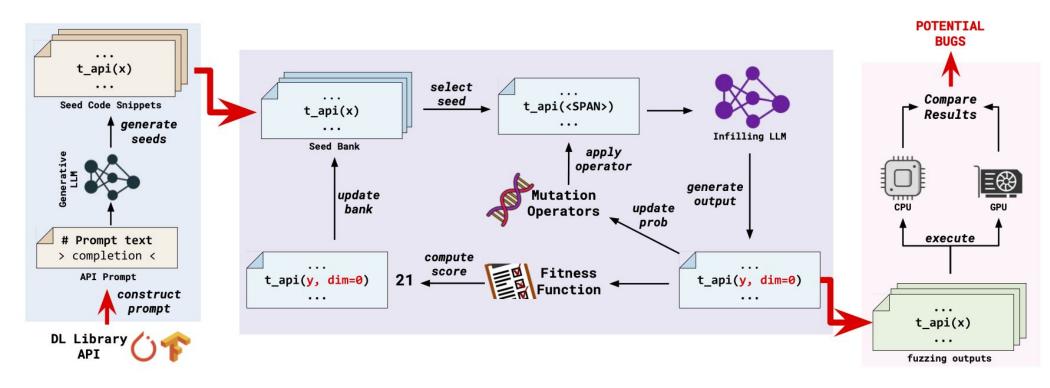
Decoder

SPAN>:zeros(3,

Infilling
```

```
A = torch.rand(50, 50)
                      Seed
                                 B = torch.clone(A)
                     Input
                                   = torch.mm(A, B)
                                                          -target API
                                                        prefix-only
           argument-replacement
                                                         <SPAN>
           A = torch.rand(50, 50)
                                                        B = torch.clone(A)
           B = torch.clone(A)
                                                        C = torch.mm(A, B)
           C = torch.mm(<SPAN>)
argument<
                                               prefix<sup>4</sup>
            keyword-insertion
                                                         prefix-argument
           A = torch.rand(50, 50)
                                                         <SPAN>
           B = torch.clone(A)
                                                        B = torch.clone(A)
           C = torch.mm(A, B, <SPAN>=<SPAN>)
                                                        C = torch.mm(<SPAN>)
           -suffix-only
           A = torch.rand(50, 50)
           B = torch.clone(A)
           C = torch.mm(A, B)
                                                         method
           <SPAN>
                                                         A = torch.rand(50, 50)
           suffix-argument
                                                        B = torch.clone(A)
           A = torch.rand(50, 50)
                                                        C = torch. \langle SPAN \rangle (A, B)
           B = torch.clone(A)
           C = torch.mm(<SPAN>)
           <SPAN>
```

### **TitanFuzz**



- Seed program generation using LLM-based generation
- 2. Evolutionary program mutation via LLM-based infilling
- 3. Differential testing oracle



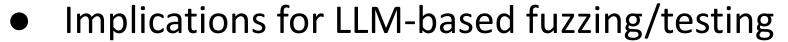






### TitanFuzz: Summary

- The first LLM-based approach for fuzzing (DL libraries and beyond)
  - Seed generation/mutation using generative/infilling LLMs
  - Up to 50.84% higher coverage than traditional fuzzers
  - Detect 65 bugs, with 44 confirmed



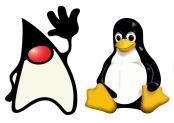
- LLMs can directly perform generation- and mutation-based fuzzing (with minimal engineering efforts)!
- Applicable to challenging domains with heterogeneous constraints
- Easily generalizable to other system domains!













Our recent studies:

400+ bugs found

300+ confirmed



### Recent trends for LLM-based fuzzing

#### More application domains

- LaST [ASE'23]
- BusyBoxFuzzer [Security'24]
- Fuzz4All [ICSE'24]



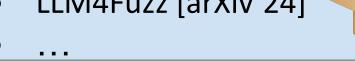
#### Edge-case test generation

- FuzzGPT [ICSE'24]
- InputBlaster [ICSE'24]
- Yanhui [ISSTA'24]



#### Opening the blackbox

- ChatAFL [NDSS'24]
- WhiteFox [OOPSLA'24]
- CovRL [ISSTA'24]
- LLM4Fuzz [arXiv'24]



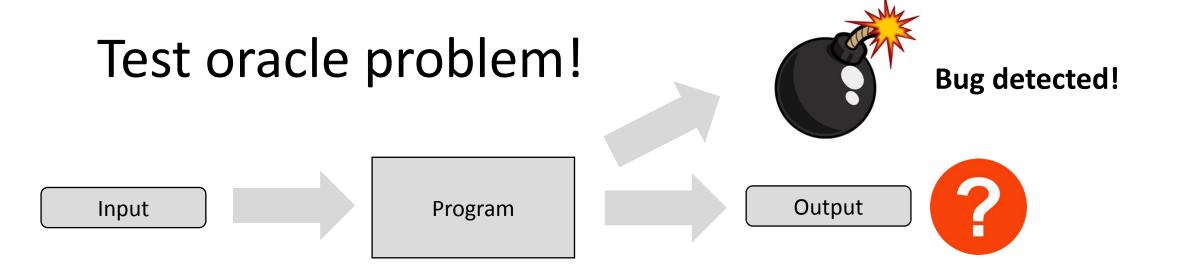
#### Fuzzer (not input) generation

- MetaMut [ASPLOS'24]
- PromptFuzz [CCS'24]
- KernelGPT [ASPLOS'25]



# Thanks!

# Backup slides

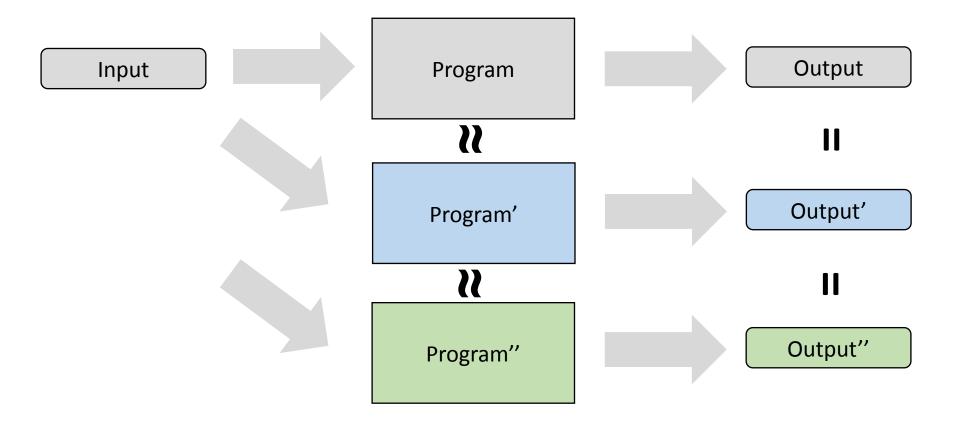


**Test oracle:** a mechanism for determining whether software executed correctly for a test<sup>1</sup>.

One of the hardest problem in Software Engineering!

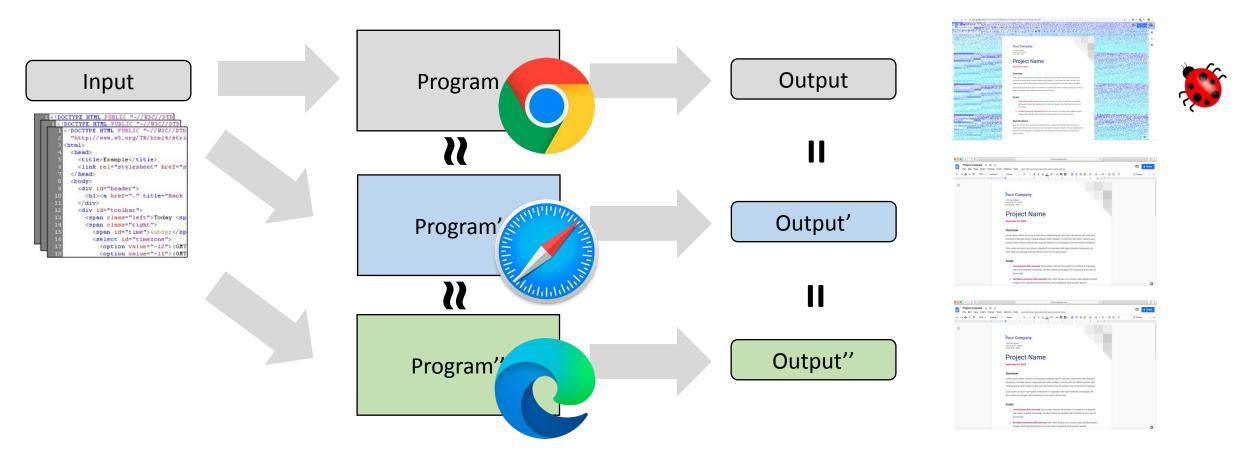
How to mitigate it?

### Differential testing



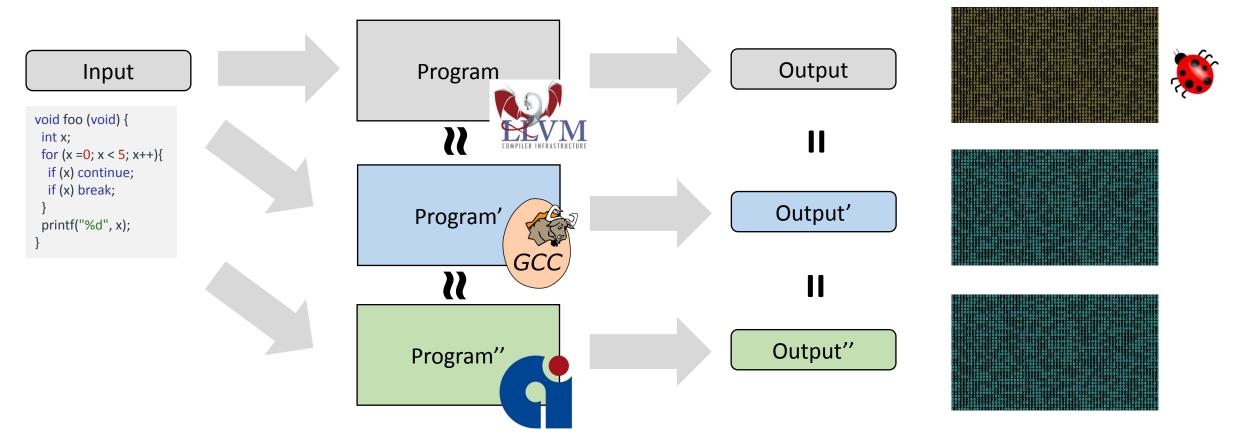
Provide the same input to similar applications, and observe output differences

### Differential testing: browsers



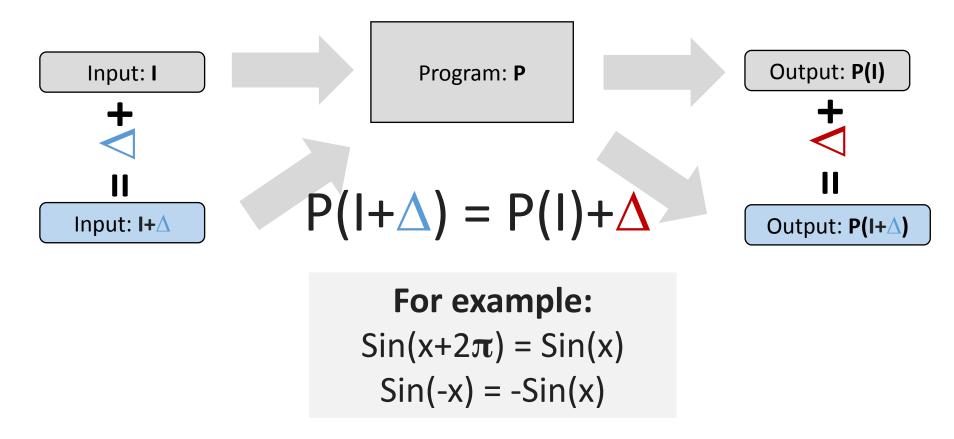
Provide the same input to similar applications, and observe output differences

### Differential testing: compilers (Csmith)



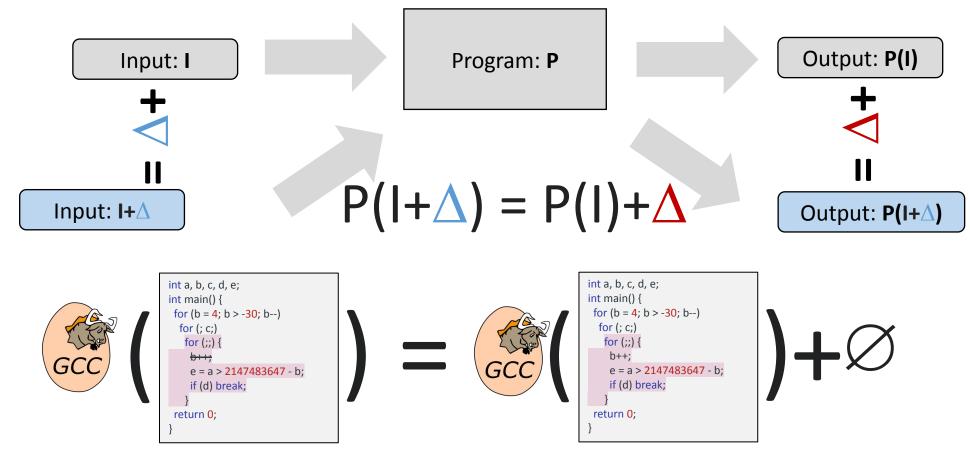
Provide the same input to similar applications, and observe output differences

### Metamorphic testing



Provide the manipulated inputs to same application, and observe if output differences are as expected

## Metamorphic testing: compilers (EMI)



Provide the manipulated inputs to same application, and observe if output differences are as expected