# CS1632, Lecture 14: Property-Based Testing

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## What is Testing?

- Checking expected behavior against observed behavior
- What we have been doing so far:
  - 1. Split the set of input values into equivalence classes
  - 2. Choose a few representative values from each equivalence class
  - 3. Write test case for those few values
  - ... And hope that those few values cover all behavior
- But do they? Are you really confident?

#### So let's take a sort function

```
public int[] sort(int[] arrToSort) {
    ...
}
```

#### Possible test cases

- null
- []
- [1]
- [-1]
- [1, 2, 3, 4, 5]
- [5, 4, 3, 2, 1]
- [-9, 7, 2, 0, -14]
- [1, 1, 1, 1, 1, 1]
- [1, 2, 3, 4 ... 99999, 100000, 100001]

#### At what point would you be satisfied?

- There is a near infinite number of sequences you could test
- Each sequence is a unique equivalence class! (almost)
  - In the sense that each sequence will take a different execution path
  - Each execution path represents a unique behavior
- Verdict: it's impossible to write enough tests to cover all behavior

Well.. It's impossible for a human, but can't we auto-generate them?

#### Stochastic Testing

- Stochastic testing: testing using randomly generated input values
  - Note: we are still not testing all input values
  - We are just testing a large number of random values hoping good coverage
- Popularly called "monkey testing" (monkey on the typewriter)
  - Not a good analogy: implies no thought is given to generation of input values
  - Testers should give \*a lot\* of thought to how input values are generated
  - Values are generated from a distribution, and distribution affects coverage
  - Testers should choose a distribution most likely to uncover defects

#### Good Distribution of Values is Important

• For example, suppose we have a defective method foo:

```
void foo (int x, int y) {
  if (x == y) {
    // Defect occurs
  }
}
```

- If we blindly test random combinations of x and y
  - $\rightarrow$  Low probability that combination where x == y is chosen
  - → Low probability that defect will be detected even with randomization
- A good distribution will contain cases where x == y

#### The Test Oracle Problem

- So now we have a set of randomly generated input values
- Now how do we auto-generate the expected output values?
  - Using the tested code? No, that is observed, not expected output value.
  - We need an "oracle" to tell us what the expected outcome should be.
  - Otherwise, output values must be calculated by humans one by one.
  - This is called the Test Oracle Problem.

What if we tested *properties* instead of output values?

## Property-Based Testing

- Property-Based Testing: testing using properties of output values
- Property: something that must invariably hold true in output values
  - Also called invariants
- Does not test output values directly
  - No need to generate the output values as part of the test case!
- Idea: Write the property once and test against all output values
  - If not all, at least a large set of output values

#### Examples of Invariants

• Invariants on statement a = b + c;

```
if (b > 0 && c > 0) { assert a > 0; } if (b < 0 && c < 0) { assert a < 0; }
```

• Invariants on statement a = b - c;

```
if ( b > c ) { assert a > 0; }
if ( b < c ) { assert a < 0; }</pre>
```

- As you can see, invariants check only a subset of behavior
  - Testing properties of an output is not equivalent to checking its value outright (e.g. if 1 + 1 returns 11, it passes assert 11 > 0 but is still incorrect)
  - But if you check enough properties, you often can get pretty close

## Going back to our sort() example

```
public int[] sort(int[] arrToSort) {
     ...
}
```

- What are the invariants?
- 1. Output array is the same size as input array
- 2. Every element in input array is in output array
- 3. No element not in input array is in output array
- 4. Values in output array are always increasing or staying the same
- 5. Idempotent: sort(arr) == sort(sort(arr))
- If all these invariants hold, by definition, the sort is provably correct!

#### Property-Based Testing Pros / Cons

- Disadvantages
  - Does not always guarantee that each output value will be correct
- Advantages
  - Can check behavior without being provided expected output values
    - → Enables stochastic testing
    - → Can even be embedded in source code to check values at runtime (java -ea Foo enables all asserts in Foo)
  - Leads coder to think about invariants and improve code. E.g.

```
int[] result = sort(arr);
if ( result.length != arr.length ) {
   // Handling of the case where lengths differ
}
```

→ Invariant is result.length == arr.length, so why the if?

## Stochastic Testing Pros / Cons

#### Advantages

- Can check numerous input values with the effort of writing just one test case
  - → The more times test case is run, the better test coverage you get!
- Input values not biased by preconceptions tester has about code. E.g.: Boundary values are tested on preconception that off-by-one errors are common
- Due to property testing, coders think about invariants and improve code

#### Disadvantages

- Due to property testing, does not always guarantee output value correctness
- Tests are not repeatable

## Stochastic Testing and Repeatability

- Reproducibility vs. Repeatability
  - Reproducibility: Ability to reproduce a defect found by a test
  - Repeatability: Ability of a test to repeat itself every time it is run
- Stochastic tests are reproducible
  - Tester only needs to record randomly generated input value causing defect
  - Tester can pass on input value to coder to reproduce the defect
- Stochastic tests are not repeatable
  - Every time the test runs, a new input value is tested by design
  - Even when defect is present, test may pass or fail depending on time of day
  - Very hard to figure out when the defect was introduced (Where should we revert to? Previous version, 10 versions ago, 100 versions ago?)

#### Stochastic Testing Best Practice

Stochastic testing should not be your main method of testing

- Use repeatable tests to the fullest --- tests that use concrete values
  - Test all the edge cases and corner cases that are frequent sources of defects
  - This should be your first line of defense
- Use stochastic testing as a second line of defense
  - To catch unexpected defects that testers didn't even think to test
  - Hard to deal with due to non-repeatability, but better than not catching defect

#### QuickCheck: Tool for Stochastic Testing

- Presented at ICFP '00 in the paper, "QuickCheck: A Lightweight Tool for Random Testing of Haskell Programs"
   https://www.researchgate.net/publication/2449938 QuickCheck A Lightweight Tool for Random Testing of Haskell Programs
- More popular in functional programming languages
  - Because all functions in functional programming are pure by definition
  - Pure functions are easier to use property-based testing on (Don't need to reason about side-effects when writing properties)
- But becoming more mainstream in other languages too

## Not just used in functional programming!

- Java: junit-quickcheck
- Ruby: rantly
- Scala: scalacheck
- Python: pytest-quickcheck
- Node.js: node-quickcheck
- PHP: php-quickcheck
- Clojure: simple-check
- C++: QuickCheck++
- .NET: FsCheck
- Erlang: Erlang/QuickCheck

## Using QuickCheck

- Two simple steps:
  - 1. Specify the properties of the allowed input
  - 2. Specify the properties of the output that must hold (invariants)

#### Example junit-quickcheck tests

- @Property: a property-based test; many random values are tested (default: 100)
- @InRange: constrains range of input values to those acceptable to sqrt()

## Write the junit-quickcheck test for sort()

```
@Property public void testSort(int[] arr) {
  int[] result = sort(arr);
  assertEquals(arr.length, result.length);
  ...
}
```

Then sit back with a beverage of your choice

QuickCheck then runs randomized test cases for us!

#### COMPUTER – DOING HARD WORK!

```
[17, 19, 1] \rightarrow [1, 17, 19]  OK
[-9, -100] \rightarrow [-100, -9] \text{ OK}
[8, 2, 987, 287, 201] \rightarrow [2, 8, 201, 287, 987] OK
[101, 20, 32, -4] \rightarrow [-4, 20, 32, 101] \text{ OK}
[115] -> [115] OK
[2, -9, -9, 1, 2] \rightarrow [-9, -9, 1, 2, 2]  OK
[8, 3, 0, 4] \rightarrow [0, 3, 4, 8]  OK
[17, 1009, -2, 413] \rightarrow [-2, 17, 413, 1009] OK
[12, 12, 1, 17, -100] \rightarrow [-100, 1, 12, 12, 17]  OK
[] -> [] OK
```

YOU —
lying on
beach
taking
foot selfies!



#### This is what it sounds like when Invariants fail

```
[17, 19, 1] \rightarrow [1, 17, 19] OK
[-9, -100] -> [-100, -9] OK
[8, 2, 987, 287, 201] -> [2, 8, 201, 287, 987] OK
[101, 20, 32, -4] \rightarrow [-4, 20, 32, 101]  OK
[115] -> [115] OK
[2, -9, -9, 1, 2] \rightarrow [-9, -9, 1, 2, 2]  OK
[8, 3, 0, 4] \rightarrow [0, 3, 4, 8]  OK
[17, 1009, -2, 413] \rightarrow [-2, 17, 413, 1009] OK
[12, 12, 1, 17, -100] \rightarrow [-100, 1, 12, 12, 17]  OK
[9, 0, -6, -5, 14] \rightarrow [0, -6, -5, 9, 14] FAIL
[] -> [] OK
```

## Shrinking

```
[9, 0, -6, -5, 14] -> [0, -6, -5, 9, 14] FAIL
[9, 0, -6] -> [0, -6, 9] FAIL
[-6, -5, 14] -> [-6, -5, 14] OK
[9, 0] -> [0, 9] OK
[0, -6] -> [0, -6] FAIL
[0] -> [0] OK
[-6] -> [-6] OK
```

Shrunk Failure: [0, -6] -> [0, -6]

## Shrinking

- Finds the smallest possible input that triggers a failure
- Helps track down actual issue
- A "toy" failure is a great thing to add to a defect report

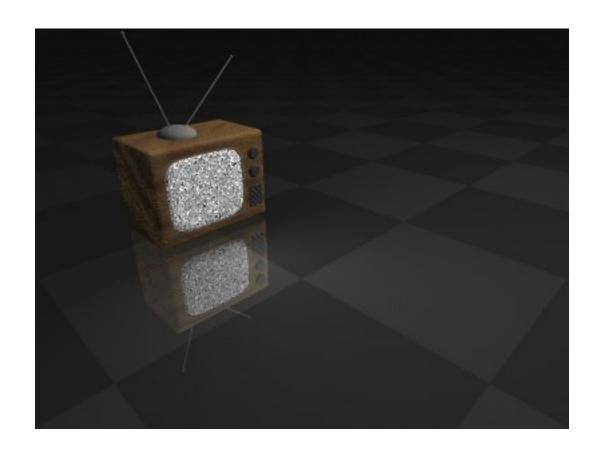
#### Adjusting Trials per Run

```
@Property(trials=1000) public void testSort(int[] arr) {
  int[] result = sort(arr);
  assertEquals(arr.length, result.length);
  ...
}
```

- By default, 100 randomly generated input values are passed to test method
- You can adjust the number of trials to 1000 using the trials attribute

## Fuzz Testing

Fuzz (noun): a blurred effect



#### Fuzz Testing

- A form of stochastic testing with a focus on "byte stream" inputs
- Idea: feed program with "fuzzy" or "noisy" byte streams and see if it
  - Exposes a defect (most commonly a system crash)
  - Exposes a security vulnerability
- A byte stream can be as varied as ...
  - An input file to an image viewer or video player
  - A network packet to a web server
  - JavaScript code interpreted by a web browser
  - A configuration file for a program

#### Fuzz Testing is not Dumb Testing

- Fuzz testing may sound like stupid testing
  - Generate some random static and force feed it to your program until it fails!
- But you will never be effective this way
  - In fact, you have to even more careful how you generate inputs
  - A byte stream is a complex form of input that leads to complex behavior
  - Unlike a simple integer input where all behavior can be covered relatively easily

## Why completely random input is ineffective

- Suppose we are testing a web browser
- Here is our test plan:
  - 1. Generate a set of randomized strings and store them into HTML files
  - 2. See if any of the HTML files crashes the web browser
- Is this the best way to test the robustness of our browser? Why (not)?
- I vote for NO
  - A browser starts by first checking the integrity of the HTML file (e.g. has necessary tags such as <a href="https://example.com/html">https://example.com/html</a>, is structured correctly, etc ...)
  - 99.99% of the randomized files will fail the initial check
  - 99.99% of the randomized files will achieve very poor code coverage
- We need a way to generate inputs with minimal integrity and structure

## A Smarter Algorithm for Fuzz Testing

- New test plan for web browser
  - 1. Collect a sample of existing HTML files (called the *corpus*)
  - 2. "Fuzz" HTML files in *corpus* to create new variants and add to *corpus*
  - 3. See if any file in the *corpus* crashes the web browser
- Steps to "fuzz" HTML files
  - 1. Parse an HTML file
  - 2. Mutate parts of parse tree with new values
    - Optionally from a dictionary; dictionary contains HTML tag names etc.
  - 3. Regenerate HTML file from parse tree and test on web browser
  - 4. Only add to *corpus* if the new HTML file increases code coverage
  - 5. Stop if sufficient code coverage is achieved, otherwise loop back to 1.
- Ensures HTML files have minimal integrity and structure

#### Now Please Read Textbook Chapter 18

- Other References:
- "Fuzzing with Code Fragments" (USENIX Security 2012)
  - <a href="https://www.usenix.org/system/files/conference/usenixsecurity12/sec12-final73.pdf">https://www.usenix.org/system/files/conference/usenixsecurity12/sec12-final73.pdf</a>
  - Fuzz testing found 105 security vulnerabilities on Firefox JavaScript engine
- libFuzzer: an LLVM compiler library for coverage-guided fuzz testing
  - https://llvm.org/docs/LibFuzzer.html
  - Still experimental but can be used with any language handled by clang (clang -fsanitize=fuzzer your\_app.c)