Testing (Part 3/3)

Martin Kellogg

Today's agenda:

- Finish up mutation testing from part 2
- Reading Quiz
- Test input generation (fuzzing)
- Test oracle generation
- Test prioritization & test suite minimization

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- Detecting these "equivalent mutants" is a big deal. How hard is it?
- It is undecidable! (= there is no algorithm for it that can always give the correct answer)
 - by direct reduction to the Halting Problem (or by Rice's theorem)

```
def foo():  # foo halts if and only if
if p1() == p2(): # p1 is equivalent to p2
   return 0
foo()
```

Takeaways: test & test suite quality

- Individual tests should be hermetic and focused
 - avoid flaky and brittle tests
- Three lenses for test suite quality: logic, statistics, and adversity
- Lens of Logic: "no visit X → no find bug in X"
 - leads to statement and branch coverage.
- Lens of Statistics: "sample the inputs the users will make"
 - leads to beta testing, A/B testing.
- Lens of Adversity: "poke realistic holes in the program and see if you find them"
 - leads to mutation testing.

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Q2: Which of the following testing strategies do the SQLite developers put the most effort into (in terms of CPU cycles):

- **A.** fuzz testing
- **B.** branch coverage
- C. regression testing
- **D.** differential testing

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Announcements:

- preliminary project plan due today
- IP2 due in one week (TODO: number of submissions as of tomorrow AM)

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 - But what else is "read in" by a program and may influence its behavior?

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What else besides "input" can influence program behavior?

- User Input (e.g., GUI)
- Environment Variables, Command-Line Args
- Scheduler Interleavings
- Data from the Filesystem
 - User configuration, data files
- Data from the Network
 - Server and service responses

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 - Lens of Adversity: choose inputs that kill mutants

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foo(a,b,c,d,e,f):
    if a < b: this
    else: that
    if c < d: foo
    else: bar
    if e < f: baz
    else: quoz</pre>
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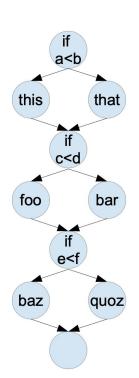
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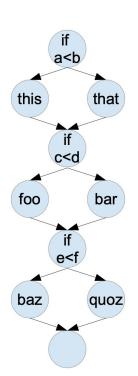
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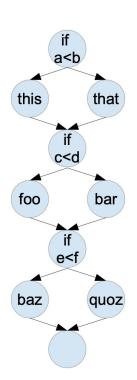
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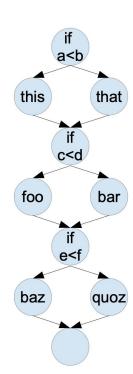


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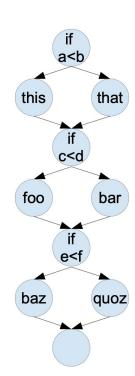
maximize:

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- branch coverage?
- path coverage?

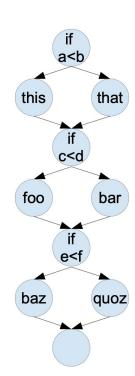
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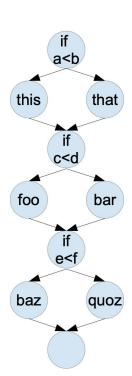
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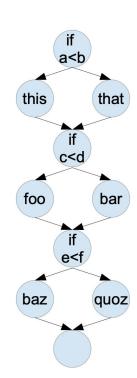
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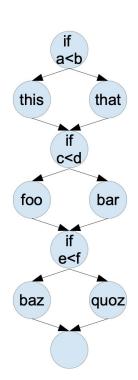
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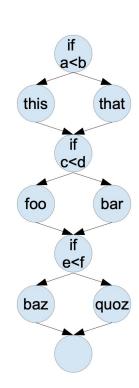
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- There are 2N branch edges
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- Path coverage subsumes branch coverage



Consider generating test inputs to cover a path

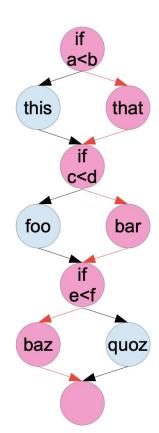
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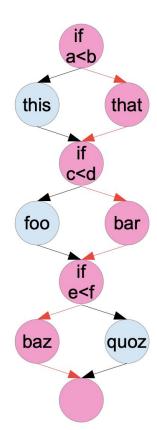
Definition: a *path predicate* (or *path condition*, or *path constraint*) is a boolean formula over program variables that is true when the program executes the given path

- Consider the highlighted (in pink) path
 - o i.e., "false, false, true"
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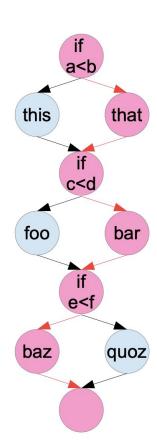
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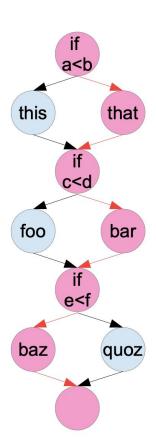
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- So, given a path predicate, how do we choose a test input that covers the path?



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- \blacksquare a=5, b=4, c=3, d=2, e=1, f=2
- \blacksquare a=0, b=0, c=0, d=0, e=0, f=1
- ... many more

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 - works surprisingly well (when answers are not sparse)
 - Option 3: use an automated theorem prover
 - cf. Wolfram Alpha, MatLab, Mathematica, Z3, etc.
 - works very well for a restricted class of equations (e.g., linear but not arbitrary polynomials, etc.)

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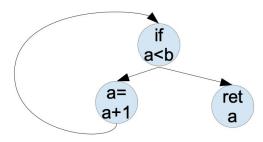
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Lens of Logic: enumerating paths

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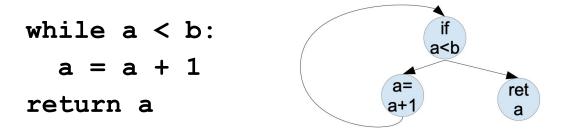
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 One path corresponds to executing the loop once, another to twice, another to three times, etc.

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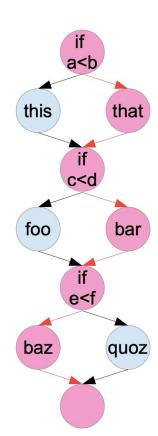
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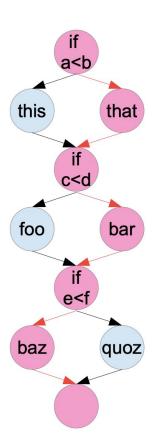
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- For more on this topic, take a graduate-level course on program analysis or compilers

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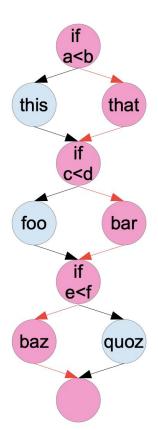


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Suppose we want to exercise the path that calls bar. One predicate is str1==str2. What do you assign to a and b?

baz

quoz

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 - If it can't (because the math is too hard, we don't control the input, etc.), we give up

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- Key question: if we generate an input for a given path, how do we tell if the program behaved correctly?

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 - "What should the program do?"
 - It is expensive both for humans and for machines.
 - and, for machines, sometimes impossible!

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Implicit oracles like these are used by most test generation tools in the real world.

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high-quality invariants can serve as test oracles

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 - For more information (e.g., how to build one) take a class on program analysis or read the Daikon paper (Feb 7 optional reading!)

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Definition: differential testing is a technique for testing two related programs by comparing their output on generated test inputs. Any difference indicates non-conformance in one of the two.

Advantages and disadvantages of differential testing:

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 - and sometimes neither is!
- but, differential testing provides a much stronger oracle than other automated techniques

Testing (part 3)

Today's agenda:

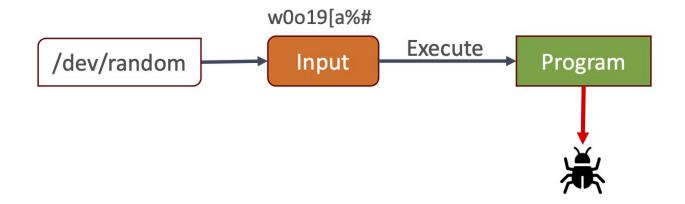
- Finish up mutation testing from part 2
- Reading Quiz
- Test input generation (fuzzing)
- Test oracle generation
- Test prioritization & test suite minimization

Test input generation

- As a human, often choosing good test inputs is the hardest part of writing a test
- For a computer, that's not true: computers can pick inputs very fast (given some policy)
- Key problem: which inputs should we pick?
 - Lens of Logic: choose inputs that will maximize coverage
 - Lens of Statistics: choose inputs "at random"
 - Lens of Adversity: choose inputs that kill mutants

Key idea: provide inputs "at random" to the program and use an implicit oracle

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Definition: fuzzing (or fuzz testing) is an automated testing technique that involves providing random or semi-random inputs to a program and monitoring for violations of an implicit oracle.

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- totally random input rarely works well
 - most programs have structured input
 - so modern fuzzers use some kind of semi-random, directed search

Modern fuzzers deal with structured input in a few ways:

mutating seed inputs:

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- combination with path predicates:
 - add inputs that are guaranteed to increase coverage to the seed pool

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- Fuzzing is machine-intensive
 - most inputs aren't useful
- Fuzzing finds real bugs
 - especially useful for finding security bugs

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Lens of Adversity: killing mutants

- Actually, not as useful as it seems for automatic test generation
 - still need to use either path predicates or fuzzing to choose inputs
- Can be a useful fitness function or guide for other automated test input generation approaches

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 - Which many produce many tests but lower-quality ones than humans would produce
 - A big cost problem!

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Which of these tests would you pick to minimize the number that need to be run?

Definition: given a budget of time, number of tests to run, or similar, the *test suite prioritization problem* is deciding which tests to run to maximize coverage while staying within the budget

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- question: how hard are these problems?
 - theory strikes again!
 - answer: it's "hard" (similar "traditional" problem that you might consider a reduction to: knapsack)

Takeaways

- two typical ways to generate test inputs:
 - solve path constraints
 - "at random" via fuzzing
- both common in practice
- both suffer from the oracle problem
 - implicit oracles are most common solution
 - invariants, differential testing, etc. also options
- in practice, you often have too many tests
 - deciding which to run is a hard problem, too