

# Fuzzing

Martin Kellogg

# Reading Quiz: fuzzing

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HW2 thoughts

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“I realized that it would be very time consuming and also difficult for me to manually collect a high coverage test suite...I **wrote a script** that would select an image **if it increases the coverage value**”

- this is an excellent approach to a problem like this!
  - always consider automation if a task is repetitive and manual
  - this student treated coverage as a **fitness function**, much like a mutational fuzzer (more details later)

# Fuzzing: agenda

- **story time**
- mutational fuzzing
- grammar-based fuzzing
- fuzzing in the real world
- start symbolic execution (if there is enough time left)

# Story Time



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- the fuzz caused many of the Unix utilities that the professor was using **to crash**
  - **insight**: just a few bits of random inputs are enough!

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Modern fuzzers combine these two ideas.

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- Many programs (especially student programs) read from a file

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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  - Ultimately programs are tested through **system calls** (e.g.,
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1. Fully **hermetic** tests should include all these inputs
  2. We want fully hermetic tests
  3. 1 & 2 imply test input generation must also **control the environment**



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# What is fuzzing?

**Key idea:** provide inputs “at random” to the program and use an *implicit oracle*

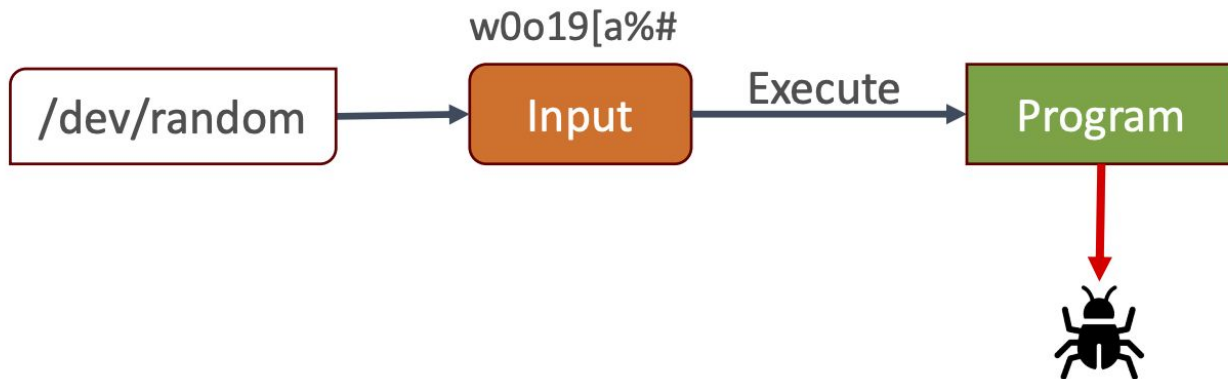
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An **implicit oracle** is an oracle that doesn't require an explicit spec from the programmer, such as “programs should not crash”.

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Truly-random input example:

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  - but that rarely works well in practice except to test the code that **reads input** (why?)

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  - because most programs accept input in a **defined format**
  - **implication:** fuzzing with random input produces tests that have **low coverage**

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- `path` is the path on that host
- `query` is a list of key/value pairs, such as `q=fuzzing`
- `fragment` is a marker for a location in the retrieved document

# Achieving high coverage

- As an example, consider a program that generates random URLs

What do you think are the odds of generating a valid URL by choosing random characters?

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  - by taking advantage of a **known grammar** for the inputs
  - by using program analysis to find **constraints** on the input that will allow it to pass various checks

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  - repeat until some stopping condition



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    - coverage!

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  - for example, in our URL parsing example, these would be URLs that we know are valid
- the **choice of seed inputs** is one of the most important inputs to the fuzzer
  - “garbage in, garbage out” is very true for this kind of fuzzer
  - can also significantly impact performance
  - HW3 hint: choose seed images carefully

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- using low-level mutations means we can inject mutants into **higher-order mutants** or seed inputs
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A **neutral** mutation is one that does not impact fitness. E.g., `google.com` is also a valid URL.

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- however, statement coverage is actually a bit **too coarse-grained** in practice
- practical fuzzers like AFL (used in HW3) use branch or **path** coverage
  - AFL's fitness function rewards an input for **any new path**, even if that path has the same branch coverage
    - this means e.g., that an input that causes a loop to go around **twice instead of once** is rewarded

# Mutational fuzzing: power schedules

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  - one input that covered a new branch or path that was created in the last round of mutation
  - $n-1$  inputs that have been in the population for at least a few generations
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  - we implement this intuition via **power schedules**

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    - when a seed is mutated to produce an input that increases fitness, its energy increases
    - when a seed is mutated, but doesn't produce an input that increases fitness, its energy decreases

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    - this technique can dramatically improve the fuzzer’s performance
  - change the power schedule to assign energy based on **distance to some objective**
    - called **directed fuzzing**

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## Population of inputs:

`https://www.google.com/`

`https://web.njit.edu/~mjk76/`

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# Fuzzing: agenda

- story time
- mutational fuzzing
- **grammar-based fuzzing**
- fuzzing in the real world
- start symbolic execution (if there is enough time left)

# Grammar-based fuzzing

- Mutating seed inputs is effective in practice to find inputs that are “near” the seeds
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`scheme://netloc/`

**Key idea:** provide that structure to the fuzzer, and only select inputs that are valid!

- In our previous example, the `scheme://netloc/` is just a seed, and we mutate each character in the URL
- But we know a lot more about how URLs are **structured**!

# Grammar-based fuzzing: review of grammars

- A *formal grammar* describes which strings from an alphabet of a formal language are valid according to the language's syntax.

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- For example, here is a grammar for URLs:

URL = S :// N / P?                      scheme://netloc/path?query#fragment

S = http | https | ftp | ...

N = any string

P = any string / P | P ? Q |  $\epsilon$

Q = any string | Q # F

F = any string

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- i.e., the seed inputs are **replaced** with the grammar, and the population is created by sampling from the grammar.
- mutation changes from “change a random character” or similar to “change a part of the **derivation tree** for a term”

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  - e.g., compilers, APIs, GUI applications
- for such programs, providing a grammar can dramatically improve fuzzing efficiency
  - downside: someone usually has to write the grammar
  - but this is an area of active research!

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- Fuzzing is **common in industry**
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- Fuzzing is **machine-intensive**
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- Fuzzing **finds real bugs**
  - especially useful for finding security bugs

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  - fuzzers have detected many important security issues
    - e.g., Heartbleed in OpenSSL

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  - but what if we just try to figure out **which inputs would improve coverage** directly?
- this is the key idea behind using **symbolic execution** to generate test inputs that improve coverage

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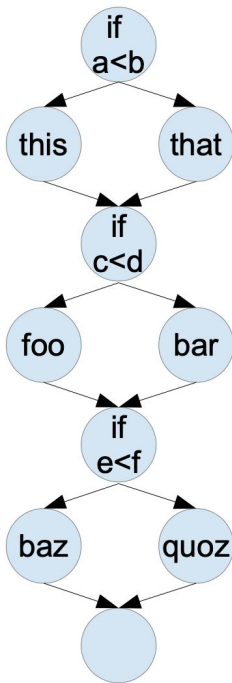
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- our plan: choose an uncovered bit of code, and then symbolically execute **backwards** from there to figure out what values the input variables would need to take on in order to cover the code
  - this is the **Lens of Logic** again, but applied in a different way

# Lens of Logic: maximize coverage

```
foo(a,b,c,d,e,f) :  
    if a < b: this  
    else: that  
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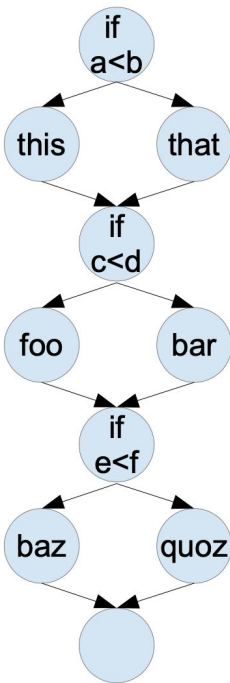
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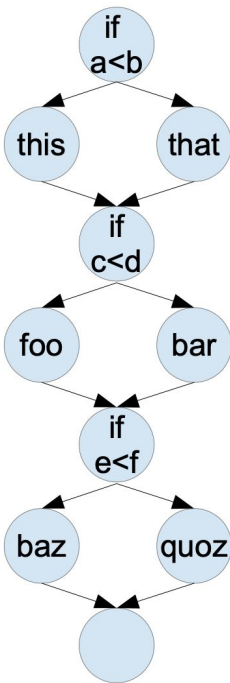
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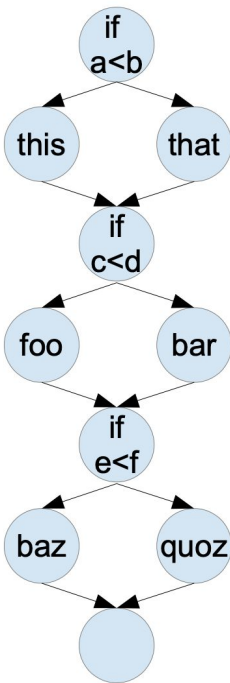
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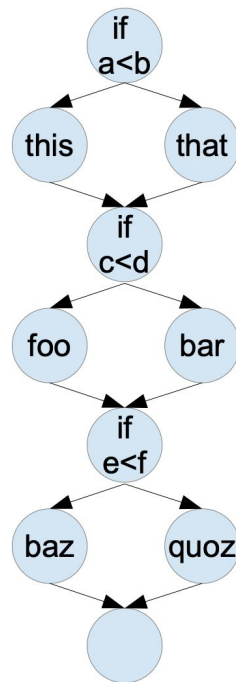


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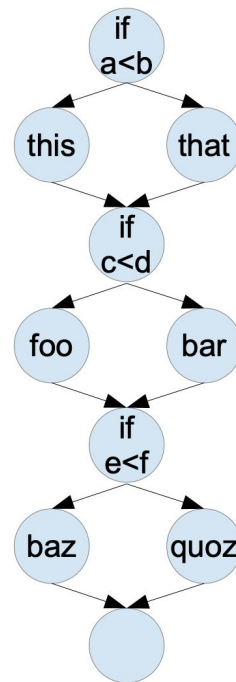
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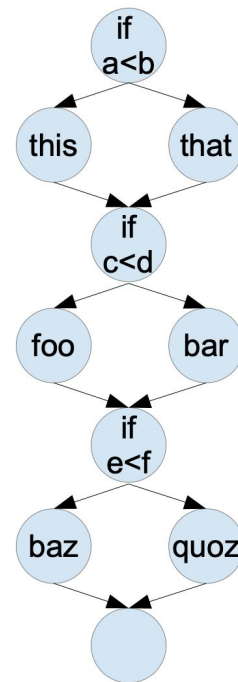
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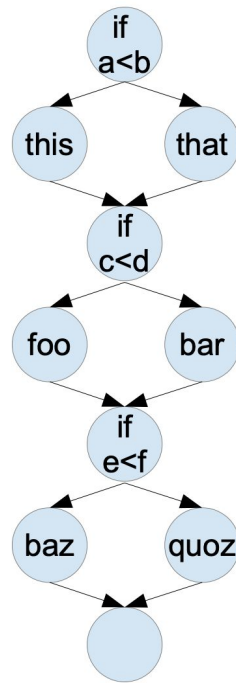
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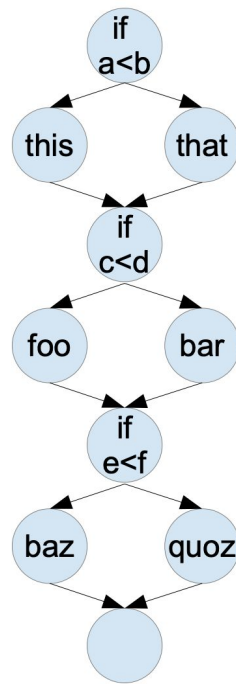
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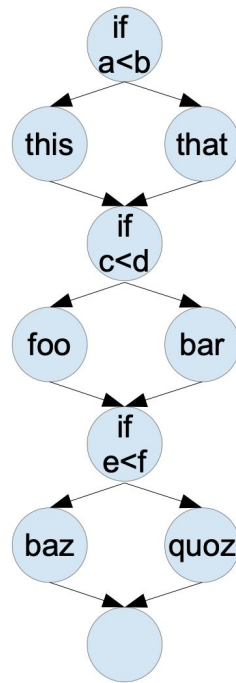
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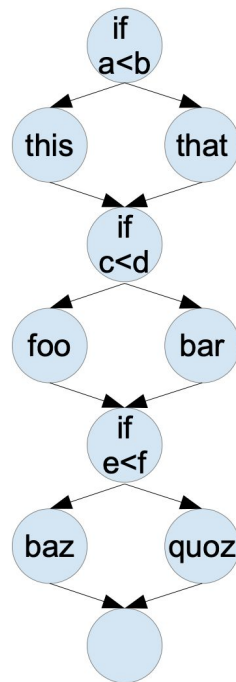
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- Path coverage **subsumes** branch coverage



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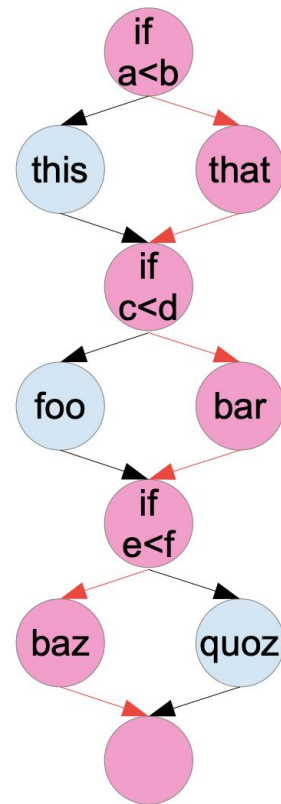
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**Definition:** a **path predicate** (or **path condition**, or **path constraint**) is a boolean formula over program variables that is true when the program executes the given path

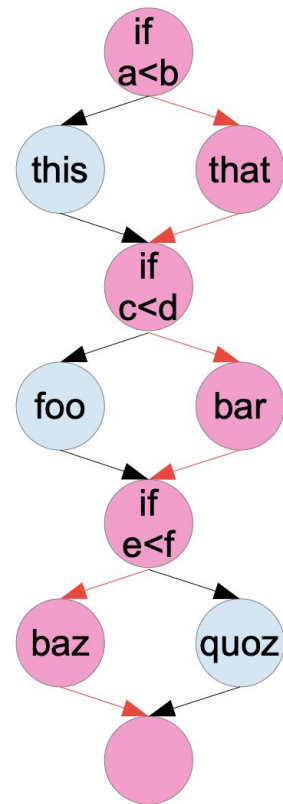
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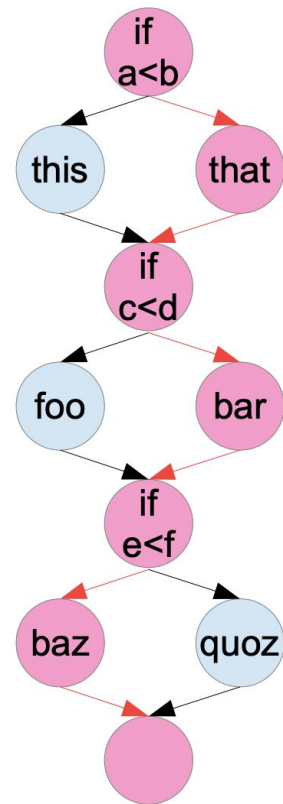
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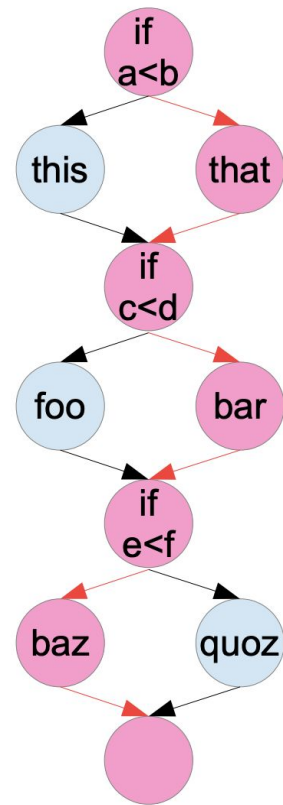
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- When the path predicate is true, control flow will follow the given path
- So, given a path predicate, how do we choose a test input that covers the path?



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  - `a >= b && c >= d && e < f ?`
    - `a=5, b=4, c=3, d=2, e=1, f=2`
    - `a=0, b=0, c=0, d=0, e=0, f=1`
    - ... many more

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  - Option 2: repeatedly **guess randomly**
    - 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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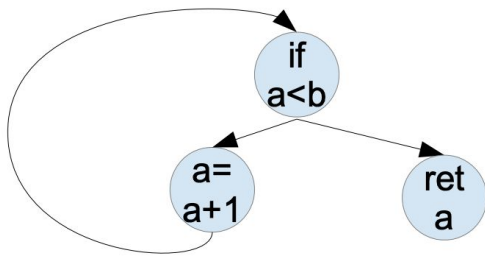
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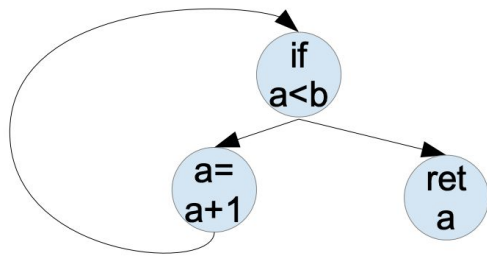
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- One path corresponds to executing the loop once, another to twice, another to three times, etc.

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

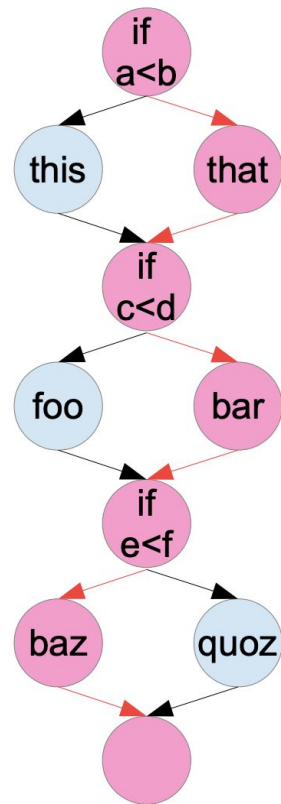


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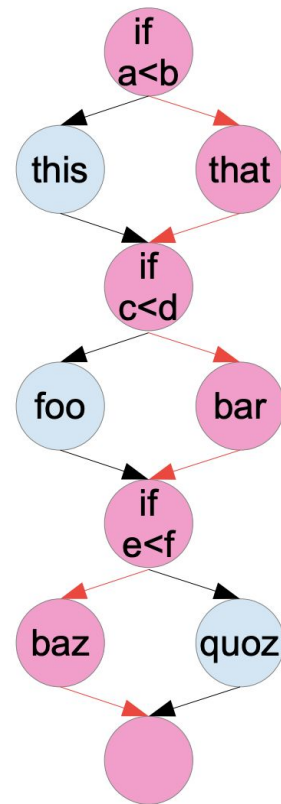
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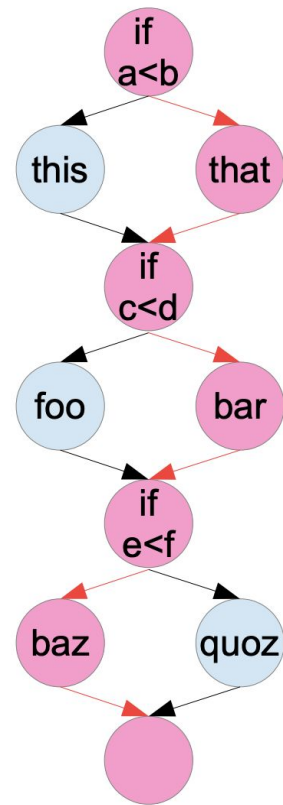
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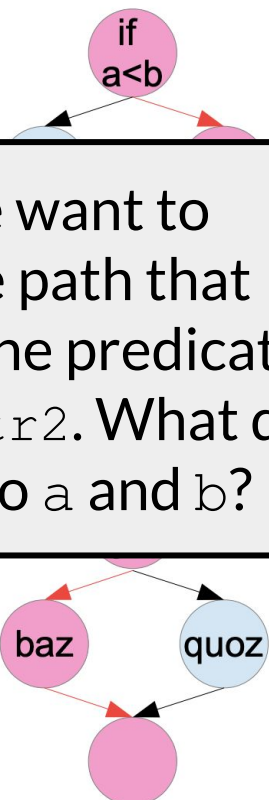
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Suppose we want to exercise the path that calls `bar`. One predicate is `str1==str2`. What do you assign to `a` and `b`?

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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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  - **Extract** their path constraints
  - **Solve** those path constraints

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- note: there is no autograder for this assignment. You only need to turn in a written report (but to write the report, you'll need data from AFL that you can only get by running it on libpng)