# 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)

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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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  - 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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#### What is fuzzing?

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

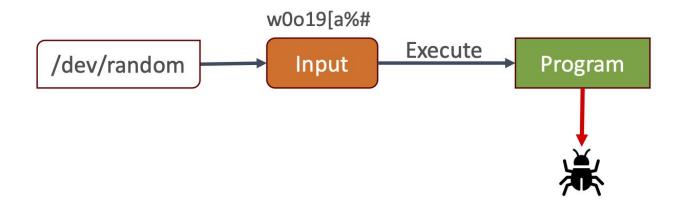
## What is fuzzing?

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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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- the simplest fuzzers use truly-random input
  - but that rarely works well in practice except to test the code Ο that reads input (why?)

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  - 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
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As an example, consider a prog choosing random characters?

scheme://netloc/path?query#fragment

What do you think are the odds

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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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- 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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- 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

- consider a new generation of test inputs containing:
  - 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

# Mutational fuzzing: putting it all together

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#### Population of inputs:

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https://web.njit.edu/~mjk76/
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      - suppose that it doesn't

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    - repeat the process...

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### Fuzzing: agenda

- story time
- mutational fuzzing
- grammar-based fuzzing
- fuzzing in the real world
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scheme://netloc/ Key idea: provide that structure to the fuzzer, and only select inputs that are valid!

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### Grammar-based fuzzing: review of grammars

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

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

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- It is straightforward to augment a fuzzer to detect buffer overflows in addition to crashes
  - ~doubles running time for most C programs, but fuzzing is already resource-intensive
  - fuzzers have detected many important security issues
    - e.g., Heartbleed in OpenSSL

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  - bad news: you can't start this homework the day before it's due
- 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)