High Volume Automated Testing with Yeager

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Overview

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Relevant URLs

- `github.com/elementc/yeager`
  - The library presented in part 2.
- `github.com/elementc/monica-tests-yeagerized`
  - Some example code, and a rudimentary test of a CRM webapp.
- `github.com/elementc/thesis`
  - The thesis and all related materials, including these slides (`conference-presentation.{tex, pdf}`)
  - See (`presentation.{tex, pdf}`) and (`main.{tex, pdf}`) for more detail, including a primer on modern web testing.

This Work Emerges From Academia

I have never worked in "the real world" on advanced testing. This work presents a tool I think is useful in some testing contexts, and a framing for already-existing techniques in the academic body of knowledge. A key theme of my thesis is that many technqiues in the family described are lost to history because nobody in industry wrote it down. Omissions are omissions of documentation, not value judgement– I welcome relevant interruptions and anecdotes, and hope the conversation continues outside of these walls.

Bugs That Traditional Testing Finds

- Known bugs, whether previously fixed or bugs that are defended against
- Unfinished features
  - As in Test Driven Development
- Clear and obvious program faults
  - Obvious to the computer
  - Crashes, for instance
  - Nonzero return codes

What Traditional Testing Does Not Find

- Faults the tester did not think to test for
- Faults that are not obvious
- Faults the tester deems improbable
- Faults that only emerge when the system is under some duress

How To Find What Traditional Testing Does Not Find

- All the bugs missed are failures of imagination.
  - If a scenario can be imagined, a test can be written for it.
- Computers are really bad at imagining, too, but are passable at rolling dice.

Examples of The Bugs We Want to Find

- Digital phone system that crashes when the 22nd line is put on hold
- Flakey text editor that has been running for months on a grad student's laptop
- System that buckles when 200k users log on at the start of a workday
- OS password prompt that sets your password to the empty string when you try using the empty string to log in
- Launching from Vostochny instead of Baikonur crashes the Fregat
- Other "hard to reproduce" failures

What Is High Volume Test Automation (HiVAT)?

Tests that algorithmically generate, execute, and evaluate the results of arbitrarily many test actions on a system, in such volume as to:[Kaner, 2013]

1. Exceed the volume a reasonable testing staff could do manually.
2. Expose behaviors of the system not normally exposed during traditional testing techniques.
3. Simulate use and abuse of the system more realistically and dynamically than would be attainable through traditional techniques.
4. Generate test scenarios that are not outside the realm of possibility or even probability due to the high-availability nature of modern software systems.

Generators

- How test cases are generated
- How the system is driven
- An engineering consideration

Interface

- Black box or white box
- Shades of grey, maybe hitting a private REST service instead of the UI directly
- A consideration of engineering and testing goals

[Hoffman, 2013]

Oracle

- How to programmatically determine correctness of generated tests
- Comparison of some sort
  - To assertions in previously written code
  - To expectations from a formal Finite State Machine
  - To a previous version of the system
  - To a competitor's system
  - To systemic expectations, like not crashing
  - Room for research here
- A consideration of engineering and testing goals

Loggers and Diagnostics

- Keeping track of test trace
- Keeping track of system health during test
- Possibly characterizing system degradation
- A consideration of testing goals

Context

- Testing objectives regardless of engineering
  - Surveying the system for new bugs
  - Determining system resillience through abuse
  - Cornering hard-to-replicate bugs in suspect modules
  - Characterizing system resource consuption over time

Scalability

- How volume in these tests is generated
  - A single, long-running thread
  - A cluster of many threads
  - A swarm of many cheap cloud servers [Parveen and Tilley, 2010]
  - A virtualization service testing a breadth of configurations
- A consideration of the testing context and engineering constraints

Purported Inventors

- HP's "evil"
  - Oldest in my literature review from 1966
- TI
- Bell
- AT&T
- Microsoft
- Telenova
- Rohm
- FAA contractors
- Automotive industry
- Miller et al. [1989] with the Fuzz Tester
  - First from academia, 1989 technical report and 1990 article.

Industrial Inventors Are Reticent To Publish

- HiVAT is perceived as a competititve advantage
- Disclosing these practices would expose testers to risk of termination or legal retaliation
- Swept away as part of efforts to minimize maintenance-related tasks
- Perceived to not be transferrable from project to project
  - "This won't help anybody else."

LSRT: Long Sequence Regression Testing

- Accomplished by modifying exisiting test suites
- Set tests to run continuously
- Remove cleanup between test runs

State Model Testing

- Build a detailed Finite State machine
- Algorithmically exercise the machine to generate testable theorems about the system

[Lee and Yannakakis, 1996]

Software Is A Finite State Machine

- Software representable as a machine with states, state transitions, inputs, outputs, and other tuples
- FSMs exactly describes the software's behavior
- Technique is popular in Electrical Engineering and for testing protocols

Exhaustive Testing

- Lower level
- Test every single possible parameter value to a function
- Needs another implementation for an oracle
- Gets prohibitively slow for multiple parameters
- Analysis, using slicing for instance [Gallagher and Lyle, 1991], can prove parameter independence and eliminate the need to test combinations of parameters

A Tale Of Two Exhaustive Tests

Hoffman [2003]

- Suspected a trig function of bugs
- Used another implementation
- Fed both functions every number in the range of a 32 bit float
- Found two errors in a few minutes

Dawson [2014]

- Suspected a trig function of bugs
- Used another implementation
- Fed both functions every number in the range of a 32 bit float
- Found one error 826k times in about 90 seconds

(see thesis page 36, "3.2 On The Recorded History of High Volume Test Automation" for more)

Fuzz Testing

- Miller's tool generates streams of random bytes and feeds them as input to UNIX command line utilities. [Miller et al., 1990]
- A test fails if the program crashes.
- Fuzz testing has grown into a diverse family of subtechniques, popular among security researchers.

Load/Performance Testing

- API tests put into a massive thread pool
- The "accepted" way to verify many users won't crash a system
- Popular tool in this family: Locust [Heyman et al., 2011]

Testing In Production

- A practice at Microsoft [Andrews, 2012]
- Candidate builds of Bing fed actual user input
- Output compared to current build
- Enables automated, staged deployments

A/B Testing

- A practice from the field of marketing
- Release candidate revisions to a subset of users and monitor for desireable behavior
- Promote the most effective revision to general availability
- Email marketing, site homepages, search engine ads, (fictitious) news stories

[Kohavi and Thomke, 2017]

Testers Write Based On The System's States

- Test scripts often are developed emulating the system's underlying state model.
- Implied state model is significantly simplified compared to a formal FSM specification.

Context: What Simplified State Models Don't Capture

If we could capture just the states and transitions implied by the test scripts being written, we'd still be missing:

- Input typed into the program
- Data the program read from some external source
- Overheating CPUs, full disks, cosmic rays, etc.

These "Simplified State Models" might still be useful.

Simplified State Models Can Be Represented As Directed Multigraphs

- System states are vertexes, or nodes.
- Test functions are edges, connecting an in-node to an out-node.
- Each edge connects one in-node to one out-node, however
  - a given function might work as a transition to an out-node from multiple compatible in-nodes.
  - This behavior is a byproduct of convenicence features in the software under test, like having a logout button on every page.
  - For brevity's sake, treat a list of in-nodes on an edge's definition as a separate edge definition for each listed in-node.

Random Walks: Generating New Test Plans Automatically

Given one of these simplified state models represented as a graph, and a source of random numbers, automatically generating test plans is straightforward.

- For a given node, the current state, from the set of nodes
- Gather all of the edges, the transition functions, which have that state as their from-node
- Select and execute one of the gathered functions
- The selected function's to-node becomes the new current state
- Repeat until some planned condition is met or execution of a selected function is not possible

What Bugs Look Like From A Modeling Perspective

- Bugs manifest as nodes which the model says should be reachable, but execution cannot successfully reach.
- Such occurrences might be bugs in the software.
- Such occurrences might be bugs in the tester's model.

Prior Art: Model Based Testing

- Jonathan Jacky, in Radiation Oncology, of the University of Washington, made an excellent Python model-based tester called PyModel.
- PyModel consumes a handcrafted model.
- It can emit a test plan that covers the whole model.
- It can emit a test plan that takes a random, should-be valid walk of the software under test.

Weaknesses in PyModel

- It requires a handcrafted model in a finicky domain-specific language.
  - Not Plain Old Python.
- It is difficult to connect to test execution.
- It requires a lot of time to get running.

What Is Yeager?

- Python version 3 module
- Annotate funtions indicating that they cause a state transition
- Infers a state model
- Can take a random walk on that model
  - Can terminate random walks under selectable conditions
- Has debug tools to understand the inferred model

Yeager's API Fits On A Notecard

- `import yeager`
- `@yeager.state_transition(from, to)`
- `yeager.walk()`
- Tweak: `yeager.add_state_to_blacklist()`, `yeager.add_transition_to_blacklist()`, `yeager.remove_state_from_blacklist()`, `yeager.remove_transition_from_blacklist()`, and `yeager.set_edge_weight()`
- Debug: `yeager.enumerate_transitions()`, `yeager.reachable_states()`, `yeager.orphaned_states()`

Write a Function

```
def login(driver):
    from pages.login import LoginPage
    lp = LoginPage(driver)
    lp.log_in_correctly(USERNAME, PASSWORD)
```

Annotate the State Transition

```
@yeager.state_transition("login", "dashboard")
def login(driver):
    from pages.login import LoginPage
    lp = LoginPage(driver)
    lp.log_in(USERNAME, PASSWORD)
```

Debug Yeager Models

- Using `enumerate_transitions` function as show in `enumerate_transitions.py`
- Using `orphaned_states` & `reachable_states` functions as shown in `orphaned_states.py` & `reachable_states.py`

Plan And Execute A Test Run

- `yeager.walk()`
- `yeager.walk(50)`
- `yeager.walk(exit_state="state-to-exit-on")`
- In development: after some visitation goal

Intuitive States in a CRM

- login page
- dashboard
- contacts list
- looking at a contact
- editing a contact
- logging a phone call or meeting with a contact
- writing in the journal
- etc.

States Necessitate Transitions

- Filling in the login form transitions from the login page to the dashboard
- Clicking a contact in the contacts list transitions to the viewing-a-contact state
- etc.

Write Some Glue and Go

Break up existing test code into logical state-transitioning units. Then for each unit:

- extract a relatively stateless function from it.
- annotate whatever transition that function triggers.

A Note on "Relative Statelessness"

- This will vary from tester to tester according to their gumption (and requirements).
- It is reasonable for a test function to require a shared webdriver so in a web test.
- It might be reasonable for a test function to require a list of all the contact names put into the system so far.
- It is unreasonable for a test function to require a memoizing key-value store with hundreds or thousands of entries.
- All extra arguments passed to `walk` are forwarded to test functions.
- Mutable arguments can be modified and these modifications persist across execution.

Visualizing the Simplified State Model

It is straightforward to use the Yeager graph inference with graph visualization software. A routine is provided to allow users to visualize with the `graph_tool` module, which can further export to the more-standardized graphviz package natively.

```
python3 visualize_graph.py
```

Take a Walk

- Execution begins with a call to `yeager.walk()`
- A demo: `python3 yeager_test.py`

What It Looks Like The Test Is Going Well

- No crash
- No assertions being tripped
- Software appears to be being executed

What It Looks Like When The Model Is Wrong

- Crash on an illogical sequence
- Example:
  - Click "Create Contact"
  - Click "Add this Contact"
  - Expected: On Contact pages
  - Actual: On Add Contact Page with an error message about needing to input a name
- A suite can generate this fault: `yeager_bad_model_test.py`

What It Looks Like When The Software Is Wrong

- Crash on a perfectly logical sequence
- Example:
  - Open a contact
  - Click "Add Reminder"
  - Fill in a date
  - Fill in a title
  - Check the "Remind me about this just once" box
  - Click the save button
  - Expected: On the contact's page, with a new reminder
  - Actual: On a 500 internal server error page
- `https://github.com/monicahq/monica/issues/326`

SSM-Based LSRT

- Benefits of LSRT by building on existing test automation investment, and exposing behavior under arbitrarily long test sequences
- Benefits of FSM modeling by thoroughly exploring the system, as well as providing valuable insight into the construction of the system

Quick To Implement

- Tests can be built as quickly as the tester can write Python.
- Tests benefit from good engineering practices elsewhere in the testing effort.
- Tests can focus on areas of the system under inspection, an incomplete model is still valuable unlike in FSMs.

Selective Detail

- Testers can hammer small details like keystrokes into a textbox or focus only on big-picture program flow.
- Testers make as many or few assertions as they wish.
- Testers can control the flow of their walks depending on the testing context.

Another Tool To Fight Failures of Imagination

- In some contexts, against some classes of bug, with sufficient resources
- Good use of idle systems
- Fill in a gap on the HiVAT family tree

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