High Volume Automated Testing with Yeager

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November 29, 2017

Acknowedgements

This work would not be possible without the support of:

- ► Cem Kaner, CSTER, and WTST participants
- ► Curtis Chambers, Jeff Farr, Mike DeCabia at Dycom Industries
- ▶ the Ruckus, the Harbor City Hooligans, the Samuels family
- ▶ Rob Atilho and Ryan Bomalaski, and many more on campus
- kbg, Richard Ford, actual and adopted family

Overview

Automated Testing

Technologies
System Under Test: Monica CRM

Patterns and Practices

Long Sequence Testing in Yeager

Software as a State Machine Usage Yeager In Action

High Volume Automated Testing

Anatomy History Family Tree

Relevant URLs

- ▶ github.com/elementc/yeager
- ▶ github.com/elementc/monica-tests-traditional
- ▶ github.com/elementc/monica-tests-yeagerized
- ▶ github.com/elementc/thesis
- ▶ github.com/monicahq/monica
- ▶ monica-doran.herokuapp.com

Why Automate Testing?

- Save time
- ► Save money
- ► Test thoroughness
 - ► Humans miss details
 - ► Humans get bored or tired

Languages

- ▶ Test frameworks exist for many languages
- ► Testers prefer "easier" scripting languages like Perl, Ruby, Python
- ► This discussion will center around Python
 - ▶ Much can be implemented in Ruby

How is Automation Achieved?

- ▶ Write functions that exercise the system under test
- ▶ Put these functions in a format that can be consumed by a test runner
- ► Call test runner
- ▶ Interpret test runner's output

Frameworks

- ▶ Has a suite of assertion convenience methods
- ► Has logging/reporting facilities
- ► Has a runner
- ▶ Python: unittest, nose, pytest
- unittest is in the Python Standard Library

Glass Box Testing

source

- ► Test code interacts directly with the System Under Test's
- ► Can probe very deeply into execution
- ▶ Use mock interfaces & shims to isolate tests

Selenium

- ► Programmatic control of web browsers for testing and other automation[Holmes and Kellogg, 2006]
- ▶ Driver class allows navigation and document queries
- ► Node class allows interaction, data retrieval, and limited Driver-like queries for children

Black Box Testing

- ► Test code interacts with the user or service interface of the running program
- ▶ Use external toolkits like Selenium to drive user interfaces
- ► Often in a special test environment but otherwise the unmodified software

HTML (summary)

- ► XML- based documents for the web
- ► Tree-structured
- ▶ Nodes have properties, including text, in addition to children

CSS (summary)

- ► Language for styling HTML documents
- ► Format- selector: rule;
- ► Selectors: strings that identify one, many, or none of the nodes in an HTML document
- ► Rules: specific styling attributes to apply to each node matched by attached rule

Page Object Modeling

- ▶ Each page on a site corresponds to a Python class.
- ▶ Fields or important strings on pages get getters and setters.
- ► Clickable buttons or links get click() functions.
 - ► If the click should transition to a new page, construct and return that new page's class.
- ▶ In class constructors, assert invariants about that page.

[Kung et al., 2000]

Monica: A Personal CRM

- ► Open-Source
- ► Life-tracker
- ► Friend-keeper
- Journal
- ▶ In the cloud
- ► Inspired by sales' Client Relationship Management (CRM) suites

How Web Test Suites Come Together

- ▶ Build all the page objects and put them in /pages/.
- Write step-by-step test plan as comments in the body of a function in the runner's format.
- ► Translate English steps into Python code.

[Nguyen, 2001]

Running Tests

- ► Same as running any other Python script
- python3 test_contacts.py
- ► Some frameworks have a multi-script runner
- ▶ python3 -m unittest

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

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

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
- ▶ Other "hard to reproduce" failures

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

Testers Write Based On The System's States

- ► Page Object Model testing pattern emulates the system's underlying state model, and includes state transitions.
- ► Implied state model is significantly simplified compared to a formal FSM specification.
- ▶ POM provides a detailed look at how the system is built.

State Models Can Help Us Plan New Tests

- ► Given a printout of a state diagram, one can trace a pen along the model and plan a new test sequence.
- ▶ What parts of the System Under Test are tested and what parts are not yet tested becomes obvious.

Context: What Simplified State Models Don't Capture

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

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

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.

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.

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

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.

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

Annotate the State Transition

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

```
@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

Test Monica With Yeager

- ► Have a robust suite of Page Object Models
- ▶ Intuitive and meaningful system
- ▶ Public service

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.

Intuitive States of Monica

- ► 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.

Use Existing Page Object Models As A Guide

- ► Emulates the Page Object Models' structure
- ► States are pages
- ► Methods are state transitions
 - ► Some transitions can be loopbacks

Write Some Glue and Go

For each method in the page object models:

- create a relatively stateless function that calls it.
- ▶ annotate any state transition that function triggers.

Example Suite's 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 graphviz natively.

python3 visualize_graph.py

A Note on "Relative Statelessness"

- ► This will vary from tester to tester according to their gumption.
- ▶ It is reasonable for a test function to require a shared webdriver so page objects can be used.
- ▶ 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.

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 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

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 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.
- 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 Interface

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

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

- ▶ 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]

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

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.

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

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

LSRT: Long Sequence Regression Testing

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

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 slices for instance [Gallagher and Lyle, 1991], can prove parameter independence and eliminate the need to test combinations of parameters

State Model Testing

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

[Lee and Yannakakis, 1996]

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

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.

Testing In Production

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

Load 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]

A/B Testing

- ► Marketing practice
- 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, news stories

[Kohavi and Thomke, 2017]

Model-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 thoroghly exploring the system, as well as providing valuable insight into the construction of the system

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.

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.

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