Exercise #1

Clone repo: <u>tiny.cc/zhd-workshop</u>

github.com/Zac-HD/escape-from-automanual-testing

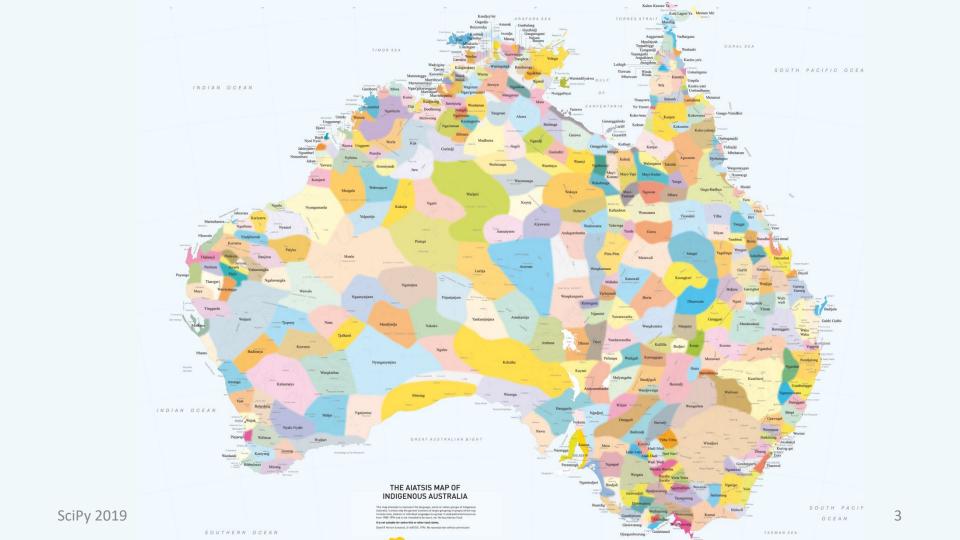
pip install pytest hypothesis

- pytest *.py
 - to check that it's all installed ☺



Escape from auto-manual testing with Hypothesis!

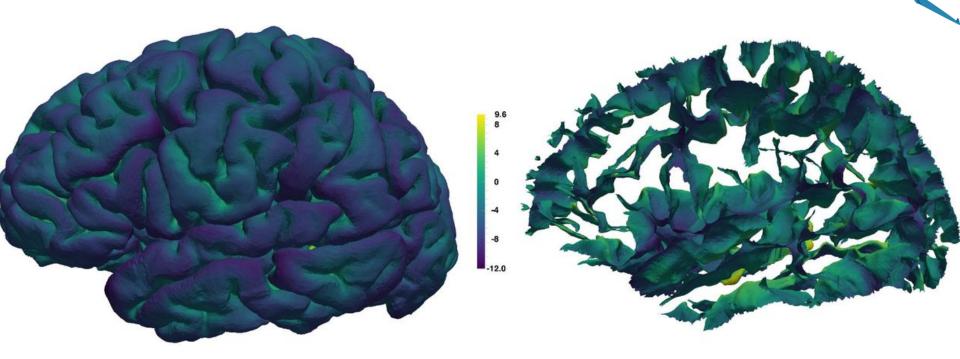
Zac Hatfield-Dodds



Outline

- 1. Motivation
- 2. Writing testable code
- 3. Property-based testing 101 Ex 1
- 4. Describing data with Hypothesis Ex 2
- 5. Testing tactics Ex 3

"This is your brain on software"



... any reported scientific result could be wrong if data have passed through a computer, and these errors may remain largely undetected ...

-- Soergal (2015) https://dx.doi.org/10.12688%2Ff1000research.5930.2

Why are we here?

- Code has become critical to our research
 - Open source libraries
 - One-off analyses

- We need faster, better ways to test it
 - Still with me? Great!



Secretly the most important part:

WRITING TESTABLE CODE

Design for testability

- Deterministic behaviour
- Immutable data
- Canonical data
- Keep I/O out of logic
- Lots of assertions

What's a assertion?



"an expression in a program which is **always true** unless there is a bug."

http://wiki.c2.com/?WhatAreAssertions

What should a test do?

- Execute the "system under test"
 - "arrange, act, assert"
 - "given, when, then"

Fail iff a bug is introduced

"Auto-manual tests"



- Humans
 - Decide the input
 - Write the test function
 - Determine and check expected results

Doing it in code is repeatable, not automated

Other kinds of tests

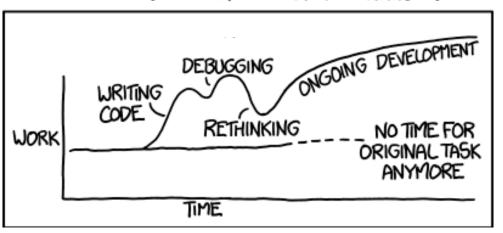
- Diff tests
 - Does new version reproduce known output?
- Doctests
 - Check that examples in docs still work
- Coverage tests
 - Identify unexecuted parts of your code
 - Please never use percent coverage!

Property-based testing



- User:
 - Describes valid inputs
 - Writes a test that passes for any valid input
- Engine:
 - Generates many test cases
 - Runs your test for each input
 - Reports minimal failing inputs (usually)

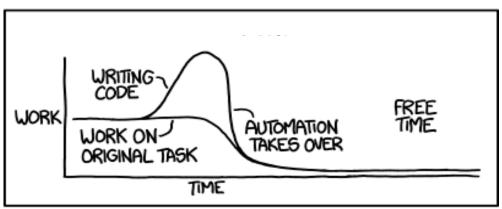
"I SPEND A LOT OF TIME Writing tests I SHOULD WRITE A PROGRAM AUTOMATING IT!"

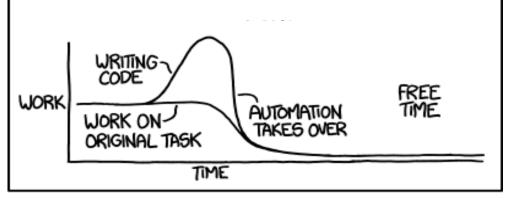




PROPERTY-BASED TESTING 101

"I SPEND A LOT OF TIME writing tests I SHOULD WRITE A PROGRAM AUTOMATING IT!"





PROPERTY-BASED TESTING 101

from hypothesis import given, strategies as st @given(st.lists(st.integers(), min_size=1) def test a sort function(ls): # we can compare to a trusted implementation, assert dubious sort(ls) == sorted(ls) # or check the properties we need directly. assert Counter(out) == Counter(ls) assert all(a<=b for a, b in zip(out, out[1:]))</pre>

Exercise #1

Clone repo: <u>tiny.cc/zhd-workshop</u>
 github.com/Zac-HD/escape-from-automanual-testing

- pip install pytest hypothesis
 - In your preferred environment, py2 or py3
- pytest pbt-101.py
- Open file, edit per comments, re-run tests



DESCRIBING DATA

hypothesis.strategies



Describes inputs for @given to generate

- Only construct strategies via the public API
 - SearchStrategy type is only public for type hints
 - Composing factories is nicer anyway!

Values

- Simplest strategies are for values
 - None, bools, numbers, Unicode or binary strings...

- Finer-grained than types
 - Optional bounds for value or length
 - Arguments like allow_nan or timezones

Collections



- Lists, sets, dicts, iterables, etc.
 - Take a strategy for elements (or keys/values)
 - Optional min_size and max_size

Map and Filter methods

s.map(f)

- applies function f to example
- shrinks before mapping

s.filter(f)

- retry unless f(ex)
- mostly for edge cases

```
s = integers()
s.map(str) # strings of digits
# even integers
s.map(lambda x: x * 2)
# odd ints, slowly
s.filter(lambda x: x % 2)
# Lists with some unique numbers
lists(s, 2).filter(
  lambda x: len(set(x)) >= 2
```

Complicated data

- Got a list of values?
 - sampled_from or permutations can help
- Recursive strategies just work
 - At least three ways to define them
- Combine strategies:
 - integers() | text()
 - Can't take intersection though.
- Call anything with builds()



Inferring strategies

A schema is a machine-readable description:

- Used for validating input
- Can generate input instead!

This tests both validation and logic.

regex, array dtype, django model, attrs classes, type hints, database...

```
\rightarrow \rightarrow from regex(r'^[A-Z]\w+$')
'Fgjdfas'
'D榙譞ć奇\n'
>>> from dtype('f4,f4,f4')
(-9.00713e+15, 1.19209e-07, nan)
(0.5, 0.0, -1.9)
>>> def f(a: int): return str(a)
>>> builds(f)
'20091'
'-507'
```

Beyond the standard library

- hypothesis.extra
 - Django, Numpy, Pandas, Lark, pytz, dateutil...
- Also many third-party extensions, e.g.
 - Geojson, SQLAlchemy, networkx, jsonschema,
 Lollipop, Mongoengine, protobuf...

Data dependencies

Custom strategies

Similar to interactive data in tests

Interactive data

- Run part of a test, then get more input
- Useful with complex dataflow

```
@composite
def str and index(draw, min size):
    s = draw(text(min size=min size))
    i = draw(integers(0, len(s) - 1))
    return (s, i)
str and index().example()
@given(data())
def test_something(data):
    i = data.draw(integers(...))
```

Minimal examples

- Strategies shrink
 - From the inside out
 - i.e. before map or filter are applied
 - Towards the smallest and shortest example
 - Based on the strategy definition
- Multiple errors possible per test!

Inline st.data()

- Draw more data within the test function
 - Great for complex or stateful systems
 - Use @composite instead if you can

```
@given(st.data())
def a_test(data):
    x = data.draw(integers(0, 100), label="First number")
    y = data.draw(integers(x, 100), label="Second number")
    # Do something with `x` and `y`
```



TACTICS FOR TESTS

Tactics: what do we test?

- "Auto-manual" testing
 - output == expected
- Oracle tests (full specification)
 - Does a magic "oracle" function say output is OK?
- Partial specification
 - Can identify some but not all failures
- Metamorphic testing
- Hyper-properties

Oracles

- Fantastic for refactoring or testing performance optimisations
- "reverse oracles"
 - Generate an answer, ask the oracle for a matching question, test that code gets the answer
- You may need to test the Oracle too

Partial specification

- We don't need an exact answer for tests!
 - $-\min(xs) \le \max(xs)$

- Lots of serialisation specs are like this
 - In fact almost all specs are partial

Special-case oracles

- If your oracle only works for some valid inputs, that's still useful to test those inputs
- Or a more precise test for a subset of inputs
 - Monotonic functions, positive numbers, etc.
 - Varying just one parameter to simplify results

Common properties



- Shared by lots of code
 - Often good API design generally
 - Or worth it just for testability

"Does not crash"

Just call your function with valid input:

```
@given(lists(integers()))
def test_fuzz_max(xs):
    max(xs) # no assertions in the test!
```

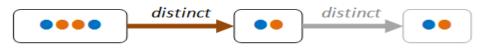
• This is embarrassingly effective.

Invariants



```
transform
```

Counter(ls) == Counter(sorted(ls))



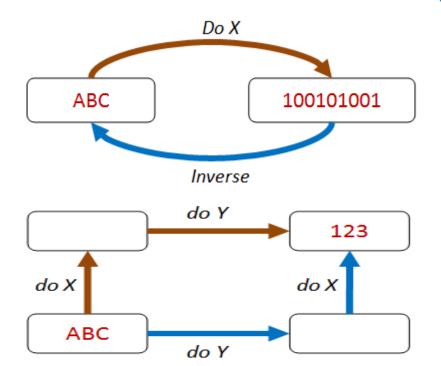
Round-trips

"inverse functions"

- add / subtract
- json.dumps / json.loads

or just related:

- factorize / multiply
- set_x / get_x
- list.append / list.index



Exercise #2



- Same plan as last time:
 - pytest strategies-and-tactics.py
 - Open, edit, re-test



Metamorphic Relations

TESTING BLACK BOXES

Metamorphic relations?

- We don't know how input relates to output
- BUT
 - Given an input and corresponding output
 - Make a known change to the input
 - We might know how the output should change (or not change)

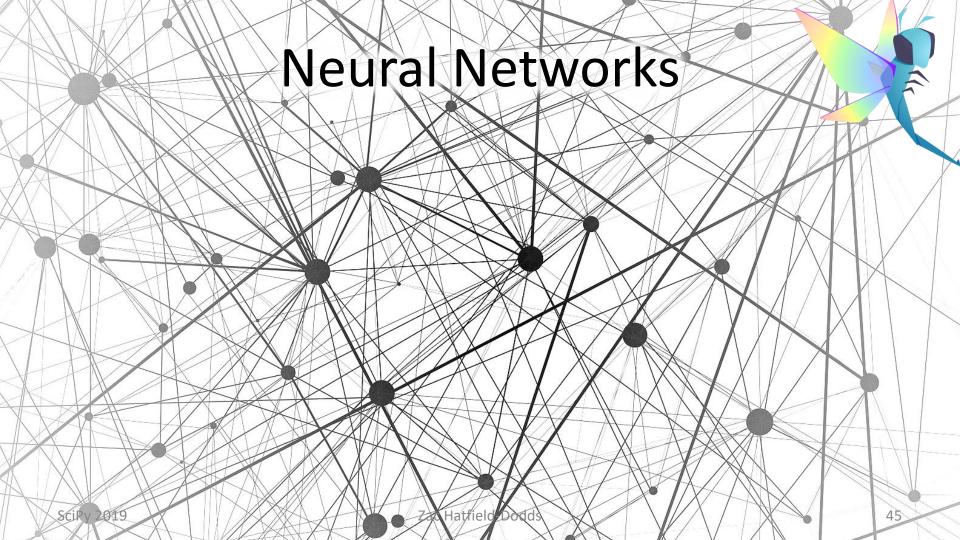
That's it – but this is really, really powerful

Compilers

- Very popular technique for compilers!
 - Generate valid program
 - Use many different compilers and settings
 - Run and compare results
 - Any difference == there's a bug somewhere

A kind of differential testing





Neural Networks

- Testing for implementation errors is rare
 - Embed lots of assertions
 - Use simple properties across single steps
- Testing things like...
 - Training steps change neuron weights
 - Bounds on inputs and outputs
 - Converges when expected to



Computer vision

- Cars should not confidently choose a different action if the camera feed has
 - slight changes in contrast or brightness or scale
 - a small skew or offset or blur added
 - light rain or fog added

- https://deeplearningtest.github.io/deepTest/
 - Luckily there aren't many on the road (yet)

Exercise #3



- Same plan as last time, again:
 - pytest test-the-untestable.py
 - Open, edit, re-test



In which we discuss all the other things that you might want to know.

PERFORMANCE, CONFIGURATION, AND COMMUNITY

Observability

- --hypothesis-show-statistics
 - Shows timing stats, perf breakdown, exit reasons
 - Add custom entries by calling event() in a test

- Use note() if you like print-debugging
 - Only prints for minimal failing example
 - Details controlled by verbosity setting

Performance (generation)

- All pretty obvious in generation phase:
 - Calling slow things or many things is slow
 - Generating larger data takes longer
 - Filter more, and getting output takes longer

Otherwise Hypothesis is pretty fast!

Performance (shrinking)

- Composition of shrinking
 - If any part shrinks, the whole should shrink
 - Order of recursive terms is important!
- Keep things local
 - Put filters (or assume) as far in as possible
 - Avoid drawing a size, then that many things
- Don't waste more tuning than you save!

Configuration

• hypothesis.settings

Per-test decorator or whole-suite profiles

- Lots of options
 - deadline, max_examples, report_multiple_bugs, database, etc.

Reproducing failures

- Hypothesis tests should never be flaky.
 - We detect most user-caused flakiness too

- Failures cached and retried until fixed
 - for local dev, reproducibility is automatic
- Printed seed to re-run failures from CI
- Explicit decorator for really tough cases



Update early & often!



- Hypothesis releases every pull request.
 - All bug fixes are available in ~30 minutes
 - As are features, performance improvements, ...
 - We use strict semver and code review
 - (and have a fantastic test suite ©)

So stay up to date – for your own sake!

Who uses Hypothesis?



- 4% of all Pythonistas (PSF survey)
- Many companies
- ~2000 open source projects (github stats)

Consulting Services

- Want exciting new features?
- Want Hypothesis training for your team?
- Want your tests (and code) reviewed?

- Zac Hatfield-Dodds and David MacIver
 - Say hi via hello@hypothesis.works



About the project

- MPL-2.0 license
- New contributors welcome!
 - most remaining issues are non-trivial
 - using or extending Hypothesis is valued too

- Tries to be *legible*
 - we design APIs and errors to teach users
 - does what you expect; or explains why not

