

Python Certificate: System Development with Python

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Table of Contents

- 1 Class Overview
- 2 Testing
- 3 Unit Testing
- 4 Profiling

Class Structure

- Lecture
- Exercises
- Lab Time

Testing

I don't need to tell you why testing is important

I'll focus on testing your Python code – not another system

Mostly unit testing

And an introduction to some of the tools

Python Testing Taxonomy

- 1 Unit Testing Tools
- 2 Mock Testing Tools
- 3 Fuzz Testing Tools
- 4 Web Testing Tools
- 5 Acceptance/Business Logic Testing Tools
- 6 GUI Testing Tools
- 7 Source Code Checking Tools
- 8 Code Coverage Tools
- 9 Continuous Integration Tools
- 10 Automatic Test Runners
- 11 Test Fixtures
- 12 Miscellaneous Python Testing Tools

http:

[//wiki.python.org/moin/PythonTestingToolsTaxonomy](http://wiki.python.org/moin/PythonTestingToolsTaxonomy)

Testing

Regression Testing:

Regression testing is any type of software testing that seeks to uncover new software bugs, or regressions, in existing functional and non-functional areas of a system after changes

Unit Testing:

unit testing is a method by which individual units of source code, sets of one or more computer program modules together with associated control data, usage procedures, and operating procedures, are tested to determine if they are fit for use

doctest

doctest

- In the stdlib
- Unique to Python?
- Literate programming
- Verify examples in docs

<http://docs.python.org/library/doctest.html>

doctest example

```
def get(self):  
    """ get() -> return TestClass's associated value.  
    >>> x = _TestClass(-42)  
    >>> print x.get()  
    -42  
    """  
    return self.val
```

An exact dump of the command line output

<http://docs.python.org/library/doctest.html>

doctest uses

“ As mentioned in the introduction, doctest has grown to have three primary uses:

- Checking examples in docstrings.
- Regression testing.
- Executable documentation / literate testing.

These uses have different requirements, and it is important to distinguish them. In particular, filling your docstrings with obscure test cases makes for bad documentation.”

<http://docs.python.org/library/doctest.html>

running doctests

```
In __name__ == "__main__" block:
```

```
if __name__ == "__main__":  
    import doctest  
    doctest.testmod()
```

(Tests the current module)

<http://docs.python.org/library/doctest.html>

running doctests

In an external file:

```
doctest.testfile("name_of_test_file.txt")
```

Tests the docs in the file (ReSt format...)

With a test runner: more on that later

running doctests

In a documentation system:

Sphinx:

Sphinx is a tool that makes it easy to create intelligent and beautiful documentation

- Lots of output options: html, pdf, epub, etc, etc...
- Can auto-generate docs from docstrings
- And run the doctests

<http://sphinx.pocoo.org/>

Code Checking

Not Really testing, but...

pychecker

<http://pychecker.sourceforge.net/>

pylint

<http://www.logilab.org/857/>

pyflakes

<http://pypi.python.org/pypi/pyflakes>

Will help you keep your source code clean

pep8

`pep8.py`

tells you when you have code which doesn't follow
PEP 8

```
pip install pep8
```

<http://pypi.python.org/pypi/pep8/>

LAB

doctests:

- doctest
 - Add doctests to the Circle class we've done in the last few classes.
 - Run it from the `if __name__ == "__main__":` clause (`code\circle.py`)
- Code style
 - Try running `pylint`, `pychecker` or `pyflakes` on it
 - Run `pep8` on it
 - (or your own code)

Lightning Talk

Lightning Talk:

Chris

Unit Testing

Gaining Traction

You need to test your code when you write it – why not preserve those tests?

And allow you to auto-run them later?

Test-Driven development:

Write the tests before the code

Unit Testing

My thoughts:

Unit testing encourages clean, decoupled design

If it's hard to write unit tests for – it's not well designed

but...

“complete” test coverage is a fantasy

PyUnit

PyUnit: the stdlib unit testing framework

```
import unittest
```

More or less a port of Junit from Java

A bit verbose: you have to write classes & methods

unittest example

```
import random
import unittest

class TestSequenceFunctions(unittest.TestCase):

    def setUp(self):
        self.seq = range(10)

    def test_shuffle(self):
        # make sure the shuffled sequence does not lose any elements
        random.shuffle(self.seq)
        self.seq.sort()
        self.assertEqual(self.seq, range(10))

        # should raise an exception for an immutable sequence
        self.assertRaises(TypeError, random.shuffle, (1,2,3))
```

unittest example (cont)

```
def test_choice(self):
    element = random.choice(self.seq)
    self.assertTrue(element in self.seq)

def test_sample(self):
    with self.assertRaises(ValueError):
        random.sample(self.seq, 20)
    for element in random.sample(self.seq, 5):
        self.assertTrue(element in self.seq)

if __name__ == '__main__':
    unittest.main()
```

(code/unitest_example.py)

<http://docs.python.org/library/unittest.html>

unittest

Lots of good tutorials out there:

Google: “python unittest tutorial”

I learned from this one:

http://www.diveintopython.net/unit_testing/index.html

nose and pytest

Due to its Java heritage, unittest is kind of verbose

also no test discovery
(though unittest2 does add that...)

So folks invented nose and pytest

nose

nose

Is nicer testing for python

nose extends unittest to make testing easier.

```
$ pip install nose
```

```
$ nosetests unittest_example.py
```

<http://nose.readthedocs.org/en/latest/>

nose example

The same example – with nose

```
import random
import nose.tools

seq = range(10)

def test_shuffle():
    # make sure the shuffled sequence does not lose any elements
    random.shuffle(seq)
    seq.sort()
    assert seq == range(10)

@nose.tools.raises(TypeError)
def test_shuffle_immutable():
    # should raise an exception for an immutable sequence
    random.shuffle( (1,2,3) )
```

nose example (cont)

```
def test_choice():  
    element = random.choice(seq)  
    assert (element in seq)  
  
def test_sample():  
    for element in random.sample(seq, 5):  
        assert element in seq  
  
@nose.tools.raises(ValueError)  
def test_sample_too_large():  
    random.sample(seq, 20)
```

(code/test_random_nose.py)

pytest

pytest

A mature full-featured testing tool

Provides no-boilerplate testing

Integrates many common testing methods

```
$ pip install pytest
```

```
$ py.test unittest_example.py
```

<http://pytest.org/latest/>

pytest example

The same example – with pytest

```
import random
import pytest

seq = range(10)

def test_shuffle():
    # make sure the shuffled sequence does not lose any elements
    random.shuffle(seq)
    seq.sort()
    assert seq == range(10)

def test_shuffle_immutable():
    pytest.raises(TypeError, random.shuffle, (1,2,3) )
```

pytest example (cont)

```
def test_choice():  
    element = random.choice(seq)  
    assert (element in seq)  
  
def test_sample():  
    for element in random.sample(seq, 5):  
        assert element in seq  
  
def test_sample_too_large():  
    with pytest.raises(ValueError):  
        random.sample(seq, 20)
```

(code/test_random_pytest.py)

A Diversion:

Context Managers: the `with` statement

A class with `__enter__()` and `__exit__()` methods.

`__enter__()` is run before your block of code

`__exit__()` is run after your block of code

Can be used to setup/cleanup before and after:
open/closing files, db connections, etc

A Diversion

“PEP 343: the with statement”

– A.M. Kuchling

[http://docs.python.org/dev/whatsnew/2.6.html#
pep-343-the-with-statement](http://docs.python.org/dev/whatsnew/2.6.html#pep-343-the-with-statement)

“Understanding Python’s with statement”

– Fredrik Lundh

<http://effbot.org/zone/python-with-statement.htm>

“The Python with Statement by Example”

– Jeff Preshing

[http://preshing.com/20110920/
the-python-with-statement-by-example](http://preshing.com/20110920/the-python-with-statement-by-example)

Parameterized Tests

A whole set of inputs and outputs to test?
pytest has a nice way to do that (so does nose...)

```
import pytest
@pytest.mark.parametrize(("input", "expected"), [
    ("3+5", 8),
    ("2+4", 6),
    ("6*9", 42),
])
def test_eval(input, expected):
    assert eval(input) == expected
```

<http://pytest.org/latest/example/parametrize.html>
(code/test_pytest_parameter.py)

Test Coverage

`coverage.py`

Uses debugging hook to see which lines of code are actually executed – plugins exist for most (all?) test runners

```
pip install coverage
```

```
nosetests --with-coverage test_codingbat.py
```

<http://nedbatchelder.com/code/coverage/>

LAB

Unit Testing:

- unittest
 - Pick a codingbat.com example
 - Write a set of unit tests using unittest
(code\codingbat.py codingbat_unittest.py)
- pytest / nose
 - Test a codingbat.com with nose or pytest
 - Try doing test-driven development
(code\test_codingbat.py)
- try running your circle doctest with nose, pytest
(and/or add doctests to your codingbat example)
- try running coverage on your tests

Performance Testing

“Premature optimization is the root of all evil”

– Donald Knuth

Profiling/timing

You can't optimize your code without knowing where the bottlenecks are.

Smarter people than me have said they they are almost always wrong when they try to logically determine where the slow code is. (I know I am)

... and how to speed it up

time.clock()

The really easy way:

```
import time

start = time.clock()
... do_some_stuff ...
print "It took %f seconds to run"%(time.clock - start)
```

It works, it's easy, and it gives a gross approximation

(use `time.clock()`, rather than `time.time()`)

timeit

The good way:

```
import timeit
```

```
timeit.timeit( statement, setup=some_stuff)
```

It's kind of a pain, but gives meaningful results.
(can also be called on the command line)

<http://docs.python.org/library/timeit.html>

(code/timing.py)

%timeit

The easy and good way:
ipython:

```
In [52]: import timing
```

```
In [53]: %timeit timing.primes_stupid(5)  
100000 loops, best of 3: 10.9 us per loop
```

Takes care of the setup/namespace stuff for you

<http://ipython.org/ipython-doc/dev/interactive/tutorial.html>

profiling

A profiler is a tool that describes the run time performance of a program, providing a variety of statistics

Helpful when you don't yet know where your bottlenecks are

The python profiler

```
python -m cProfile profile_example.py
```

spews some stats
(demo)

<http://docs.python.org/library/profile.html>

python profiler

What you get:

ncalls the number of calls.

tottime the total time spent in the given function (and excluding time made in calls to sub-functions),

percall the quotient of tottime divided by ncalls

cumtime the total time spent in this and all subfunctions (from invocation till exit). This figure is accurate even for recursive functions.

percall the quotient of cumtime divided by primitive calls

(demo: `python -m cProfile profile_example.py`)

python profiler

You can also dump to a file:

```
$ python -m cProfile -o profile_dump profile_example.py
```

This gives you a binary file you can examine with pstats:

```
demo: $ python -m pstats
```

pstats

Running pstats

```
$
```

```
$ python -m pstats
```

```
Welcome to the profile statistics browser.
```

```
% read profile_dump
```

```
profile_dump% stats
```

```
Wed Aug 29 16:21:39 2012      profile_dump
```

```
51403 function calls in 0.032 seconds
```

```
Random listing order was used
```

ncalls	tottime	percall	cumtime	percall	filename:lineno(func)
51200	0.006	0.000	0.006	0.000	{method 'append' o
1	0.000	0.000	0.032	0.032	profile_example.py
1	0.001	0.001	0.032	0.032	profile_example.py
100	0.022	0.000	0.027	0.000	profile_example.py

pstats commands

Commands:

- `help` help on pstats or particular command
- `stats` print the profile statistics
- `sort` sort by various data fields
- `strip` strips the leading path info from file names
- `callers` Print callers statistics
- `callees` Print callees statistics
- `quit` quits

Each has options to customize output

automating profile stats

cProfile and pstats are also modules

So you can script collection of profiles and stats

<http://docs.python.org/library/profile.html>

performance tips

Some common python performance issues:

`http:
//wiki.python.org/moin/PythonSpeed/PerformanceTips/`

(some nifty profiling tools described there, too)

LAB

Profiling lab

- run `timeit` on some code of yours (or `timing.py`, or..)
- run iPython's `%timeit` on the same code.
- try to make the factorial code in `timing.py` faster, and time the difference.
- write some code that tests one of the performance issues in:
`http://wiki.python.org/moin/PythonSpeed/PerformanceTips`
use one of the `timeits` to see if you can make a difference.
- try the profile tutorial at:
`http://pysnippet.blogspot.com/2009/12/`
`profiling-your-python-code.html`

Wrap up

Hopefully you've got a bit of an idea how to do unit testing in Python.

And will now start doing it.

And have an idea how to profile and time your code.

Next Week:

Votes are in:

4 votes for persistence: 0 for Extending with C

Persistence it is

Homework

- Write some unit tests for your class project
- Profile your class project
- Use timeit to test the slowest part of the code
- See if you can make it faster