Beyond assertion: setup and teardown

UNIT TESTING FOR DATA SCIENCE IN PYTHON

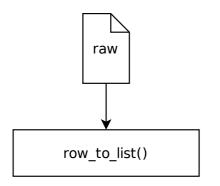


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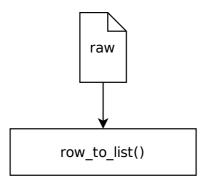


```
1,801 201,411
1,767565,112
2,002 333,209
1990 782,911
1,285 389129
```



```
1,801 201,411
1,767565,112 # dirty row, no tab
2,002 333,209
1990 782,911
1,285 389129
```

```
1,801 201,411
2,002 333,209
1990 782,911
1,285 389129
```



row_to_list()

convert_to_int()

```
1990 782,911 # dirty row, no comma
1,285 389129 # dirty row, no comma
```

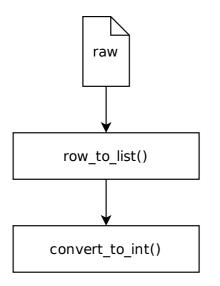
1,801

2,002

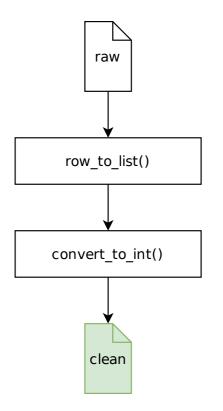
201,411

333,209





```
1,801 201,411
2,002 333,209
```

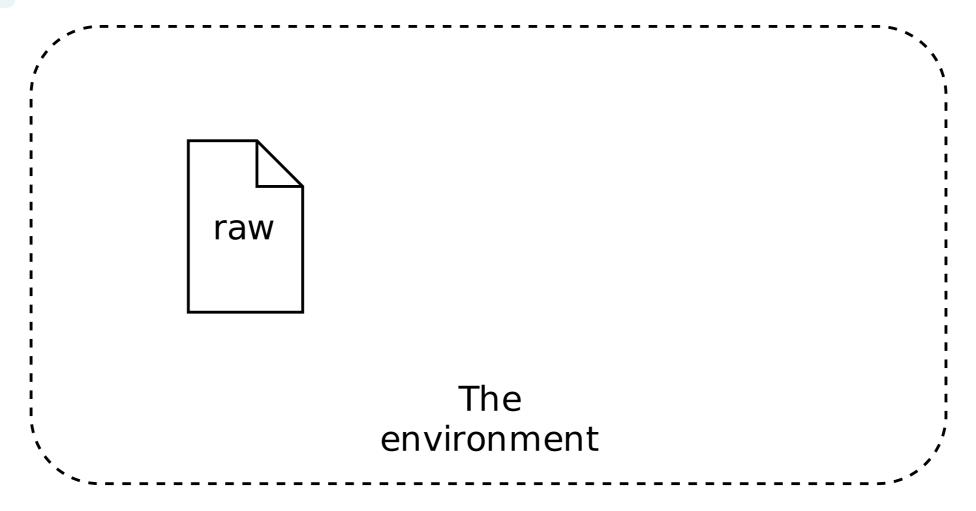


 1801
 201411

 2002
 333209

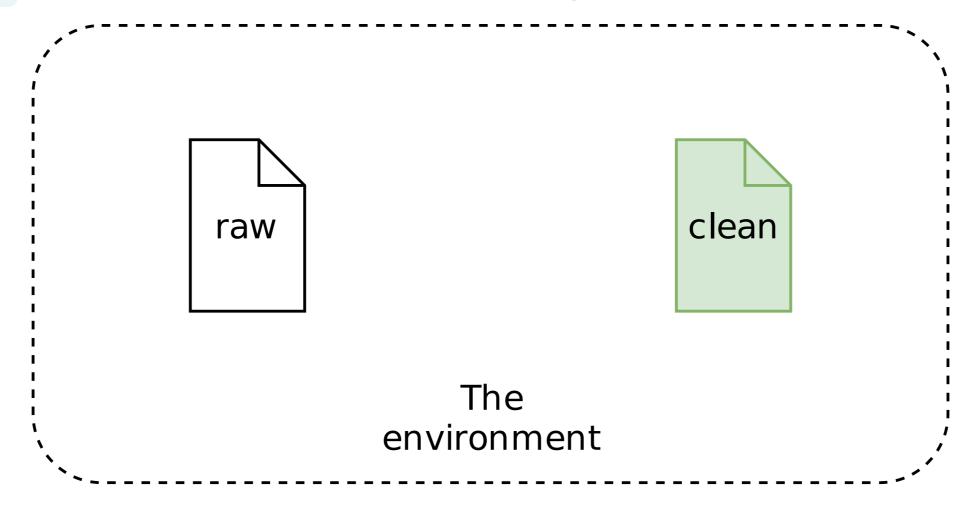
Environment preconditions

• preprocess() needs a raw data file in the environment to run.



Environment modification

• preprocess() modifies the environment by creating a clean data file.



Testing the preprocessing function

```
def test_on_raw_data():
```

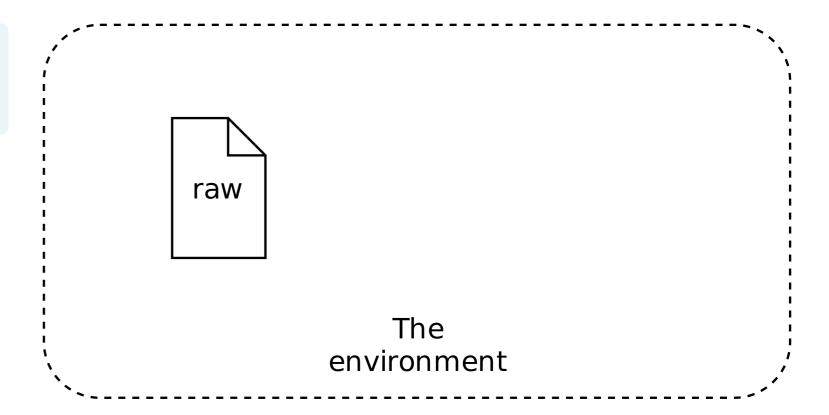
The environment



Step 1: Setup

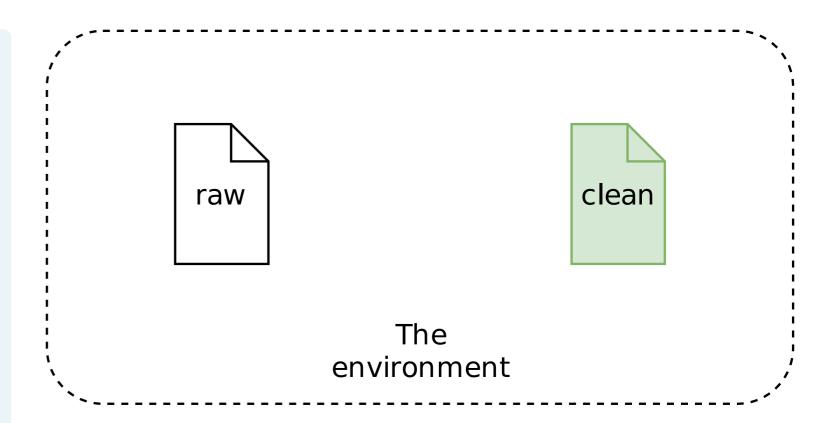
```
def test_on_raw_data():
    # Setup: create the raw data file
```

• Setup brings the environment to a state where testing can begin.



Step 2: Assert

```
def test_on_raw_data():
   # Setup: create the raw data file
    preprocess(raw_data_file_path,
               clean_data_file_path
   with open(clean_data_file_path) as f:
       lines = f.readlines()
   first_line = lines[0]
    assert first_line == "1801\t201411\n"
    second_line = lines[1]
   assert second_line == "2002\t333209\n"
```



Step 3: Teardown

```
def test_on_raw_data():
    # Setup: create the raw data file
    preprocess(raw_data_file_path,
               clean_data_file_path
    with open(clean_data_file_path) as f:
        lines = f.readlines()
    first_line = lines[0]
    assert first_line == "1801\t201411\n"
    second_line = lines[1]
    assert second_line == "2002\t333209\n"
    # Teardown: remove raw and clean data file
```

• Teardown brings environment to initial state.



The new workflow

Old workflow

assert

New workflow

• setup \rightarrow assert \rightarrow teardown

Fixture

```
import pytest

@pytest.fixture

def my_fixture():
    # Do setup here
    return data
```

```
def test_something(my_fixture):
    ...
    data = my_fixture
    ...
```

Fixture

```
import pytest
@pytest.fixture
def my_fixture():
   # Do setup here
    yield data  # Use yield instead of return
   # Do teardown here
def test_something(my_fixture):
    data = my_fixture
```

. . .

Test

```
import os
import pytest

def test_on_raw_data():
```

Fixture

@pytest.fixture def raw_and_clean_data_file(): raw_data_file_path = "raw.txt" clean_data_file_path = "clean.txt" with open(raw_data_file_path, "w") as f: f.write("1,801\t201,411\n" "1,767565,112\n" "2,002\t333,209\n" "1990\t782,911\n" "1,285\t389129\n" yield raw_data_file_path, clean_data_file_pa os.remove(raw_data_file_path) os.remove(clean_data_file_path)

Test

```
import os
import pytest
def test_on_raw_data(raw_and_clean_data_file):
    raw_path, clean_path = raw_and_clean_data_fil
    preprocess(raw_path, clean_path)
    with open(clean_data_file_path) as f:
        lines = f.readlines()
    first_line = lines[0]
    assert first line == "1801\t201411\n"
    second_line = lines[1]
    assert second line == "2002\t333209\n"
```

The built-in tmpdir fixture

- **Setup**: create a temporary directory.
- **Teardown**: delete the temporary directory along with contents.

tmpdir and fixture chaining

• setup of tmpdir() \rightarrow Setup of raw_and_clean_data_file() \rightarrow test \rightarrow teardown of raw_and_clean_data_file() \rightarrow teardown of tmpdir().

```
@pytest.fixture
def raw_and_clean_data_file(tmpdir):
    raw_data_file_path = tmpdir.join("raw.txt")
    clean_data_file_path = tmpdir.join("clean.txt")
    with open(raw_data_file_path, "w") as f:
        f.write("1,801\t201,411\n"
                "1,767565,112\n"
                "2,002\t333,209\n"
                "1990\t782,911\n"
                "1,285\t389129\n"
   yield raw_data_file_path, clean_data_file_path
   # No teardown code necessary
```

Let's practice setup and teardown using fixtures!

UNIT TESTING FOR DATA SCIENCE IN PYTHON



Mocking

UNIT TESTING FOR DATA SCIENCE IN PYTHON

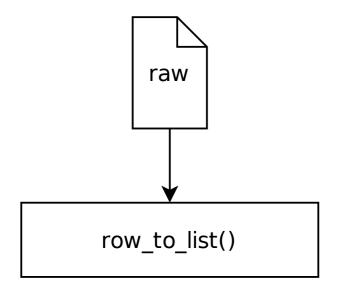


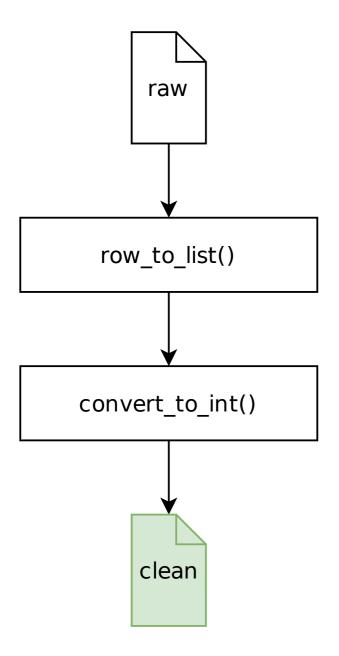
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Test Automation Engineer





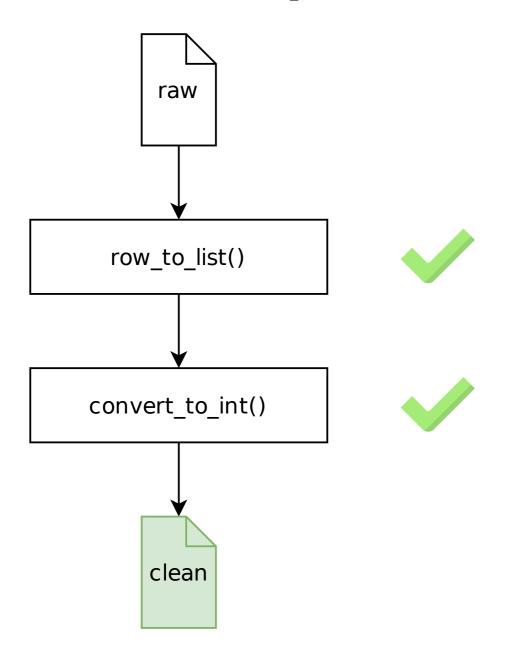






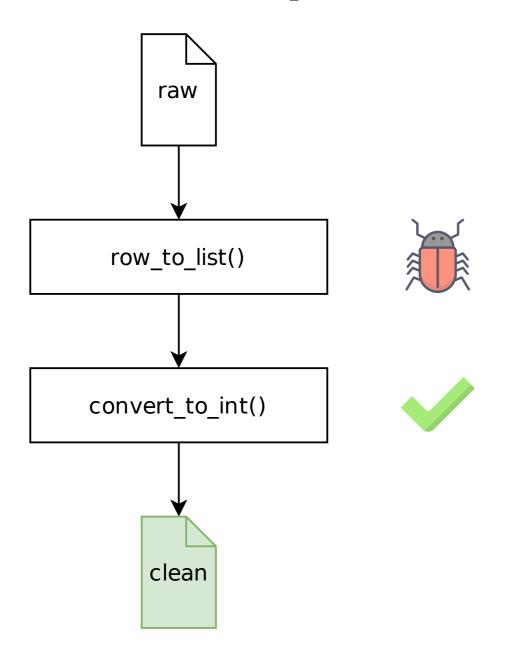
```
pytest -k "TestPreprocess"
```

Test result depend on dependencies



===== 1 passed, 20 deselected in 0.61 seconds ======

Test result depend on dependencies



```
pytest -k "TestPreprocess"
```

```
========= test session starts ===========
collected 21 items / 20 deselected / 1 selected
data/test_preprocessing_helpers.py F
                                       [100%]
______ TestPreprocess.test_on_raw_data ______
   def test_on_raw_data(self, raw_and_clean_data_file):
       raw_path, clean_path = raw_and_clean_data_file
      preprocess(raw_path, clean_path)
      with open(clean_path, "r") as f:
          lines = f.readlines()
      first_line = lines[0]
      IndexError: list index out of range
data/test_preprocessing_helpers.py:121: IndexError
```

Test result depends on dependencies

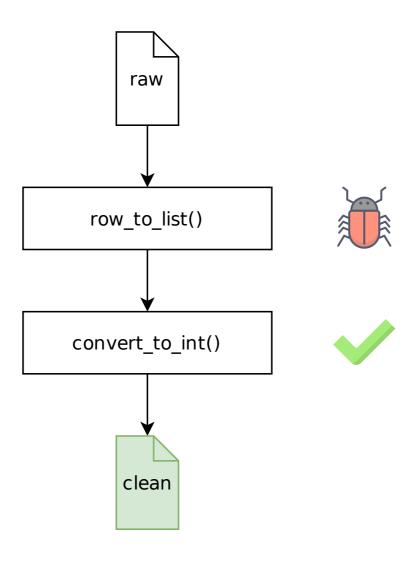
Test result should indicate bugs in

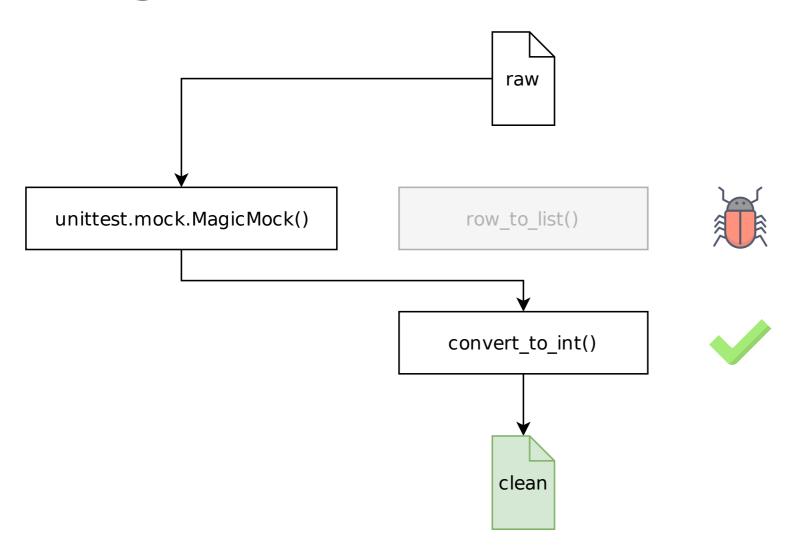
- function under test i.e. preprocess() .
- not dependencies e.g. row_to_list() or convert_to_int().

Mocking: testing functions independently of dependencies

Packages for mocking in pytest

- pytest-mock : Install using pip install pytest-mock .
- unittest.mock: Python standard library package.





Theoretical structure of mocker.patch()

```
mocker.patch("<dependency name with module name>")
```

Theoretical structure of mocker.patch()

```
mocker.patch("data.preprocessing_helpers.row_to_list")
```

```
unittest.mock.MagicMock()
```

Making the MagicMock() bug-free

Raw data

```
1,801 201,411
1,767565,112
2,002 333,209
1990 782,911
1,285 389129
```

```
def row_to_list_bug_free(row):
    return_values = {
        "1,801\t201,411\n": ["1,801", "201,411"],
        "1,767565,112\n": None,
        "2,002\t333,209\n": ["2,002", "333,209"],
        "1990\t782,911\n": ["1990", "782,911"],
        "1,285\t389129\n": ["1,285", "389129"],
     }
    return return_values[row]
```

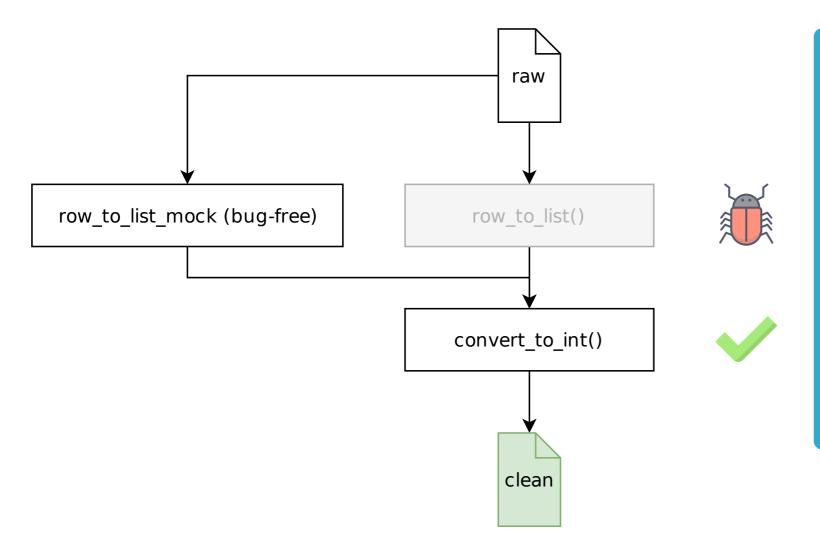
Side effect

Raw data

```
1,801 201,411
1,767565,112
2,002 333,209
1990 782,911
1,285 389129
```

```
def row_to_list_bug_free():
    return_values = {
        "1,801\t201,411\n": ["1,801", "201,411"],
        "1,767565,112\n": None,
        "2,002\t333,209\n": ["2,002", "333,209"],
        "1990\t782,911\n": ["1990", "782,911"],
        "1,285\t389129\n": ["1,285", "389129"],
        }
    return return_values[row]
```

Bug free replacement of dependency



Checking the arguments

• call_args_list attribute returns a list of arguments that the mock was called with

```
row_to_list_mock.call_args_list
```

```
[call("1,801\t201,411\n"),
  call("1,767565,112\n"),
  call("2,002\t333,209\n"),
  call("1990\t782,911\n"),
  call("1,285\t389129\n")
]
```

Checking the arguments

• call_args_list attribute returns a list of arguments that the mock was called with

```
row_to_list_mock.call_args_list
```

```
[call("1,801\t201,411\n"),
  call("1,767565,112\n"),
  call("2,002\t333,209\n"),
  call("1990\t782,911\n"),
  call("1,285\t389129\n")
]
```

```
from unittest.mock import call
def test_on_raw_data(raw_and_clean_data_file,
                     mocker,
    raw_path, clean_path = raw_and_clean_data_file
    row_to_list_mock = mocker.patch(
        "data.preprocessing_helpers.row_to_list",
        side_effect = row_to_list_bug_free
    preprocess(raw_path, clean_path)
    assert row_to_list_mock.call_args_list == [
        call("1,801\t201,411\n"),
        call("1,767565,112\n"),
        call("2,002\t333,209\n"), call("1990\t782,911\n"
        call("1,285\t389129\n")
```

Dependency buggy, function bug-free, test still passes!

```
pytest -k "TestRowToList"
```

```
collected 21 items / 14 deselected / 7 selected
                                  [100%]
data/test_preprocessing_helpers.py .....FF
   _____ TestRowToList.test_on_normal_argument_1 _________
    ======== 2 failed, 5 passed, 14 deselected in 0.70 seconds =========
```



Dependency buggy, function bug-free, test still passes!



Let's practice mocking!

UNIT TESTING FOR DATA SCIENCE IN PYTHON



Testing models

UNIT TESTING FOR DATA SCIENCE IN PYTHON



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Functions we have tested so far

- preprocess()
- get_data_as_numpy_array()
- split_into_training_and_testing_sets()

Raw data to clean data

data/raw/housing_data.txt

```
2,081 314,942
1,059 186,606
293,410 <-- row with missing area
1,148 206,186
...
```

Raw data to clean data

data/clean/clean_housing_data.txt

```
2081 314942
1059 186606
1148 206186
...
```

Clean data to NumPy array

```
get_data_as_numpy_array(
    "data/clean/clean_housing_data.txt", 2
)
```

Splitting into training and testing sets

```
from data.preprocessing_helpers import preprocess
from features.as_numpy import get_data_as_numpy_array
from models.train import (
  split_into_training_and_testing_sets
preprocess("data/raw/housing_data.txt",
           "data/clean/clean_housing_data.txt"
data = get_data_as_numpy_array(
    "data/clean/clean_housing_data.txt", 2
training_set, testing_set = (
    split_into_training_and_testing_sets(data)
```

```
split_into_training_and_testing_sets(data)
```

Functions are well tested - thanks to you!



The linear regression model

```
def train_model(training_set):
```



The linear regression model

```
from scipy.stats import linregress

def train_model(training_set):
    slope, intercept, _, _, _ = linregress(training_set[:, 0], training_set[:, 1])
    return slope, intercept
```

Return values difficult to compute manually



Return values difficult to compute manually

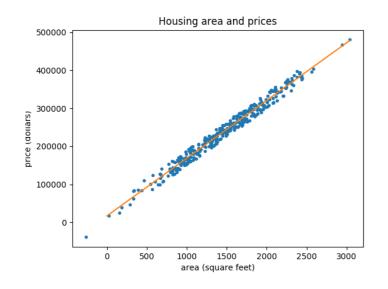


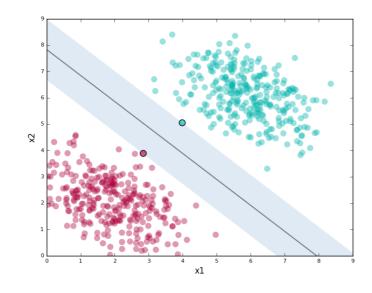
Return values difficult to compute manually

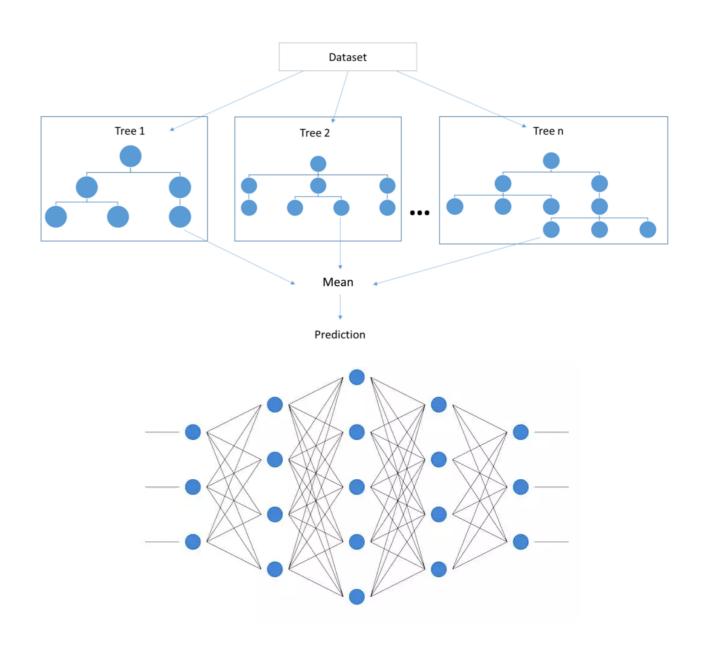


• Cannot test train_model() without knowing expected return values.

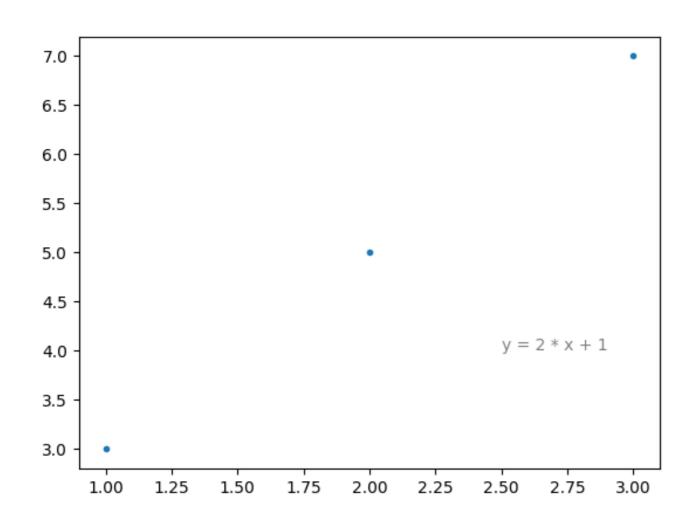
True for all data science models



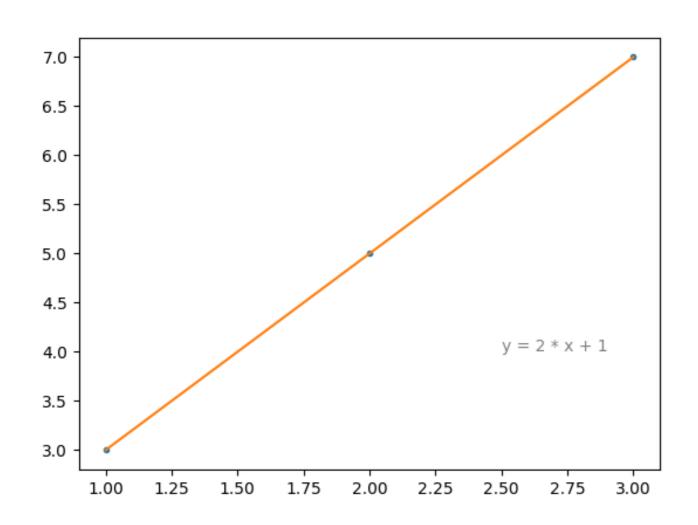




Trick 1: Use dataset where return value is known

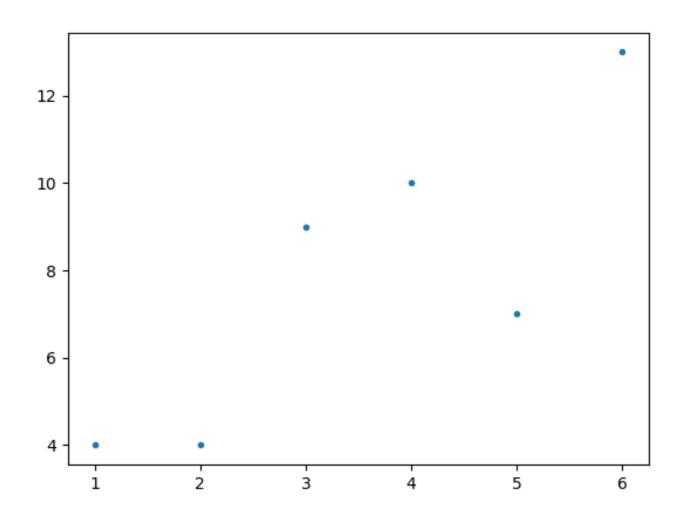


Trick 1: Use dataset where return value is known

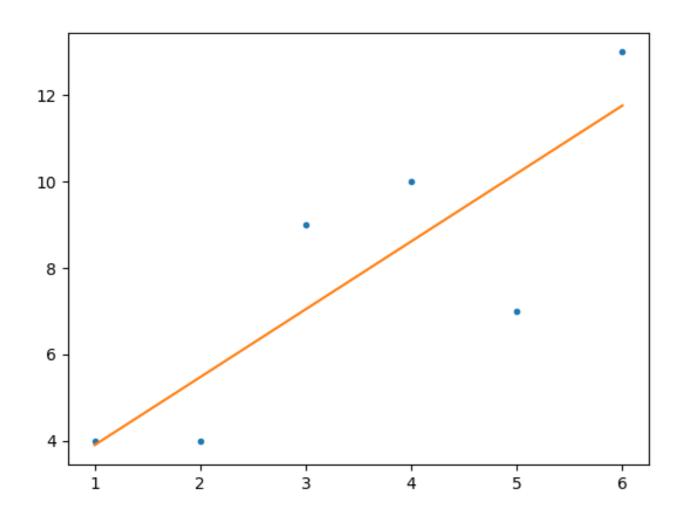


```
import pytest
import numpy as np
from models.train import train_model
def test_on_linear_data():
   test_argument = np.array([[1.0, 3.0],
                              [2.0, 5.0],
                              [3.0, 7.0]
   expected_slope = 2.0
   expected_intercept = 1.0
    slope, intercept = train_model(test_argument)
    assert slope == pytest.approx(expected_slope)
   assert intercept == pytest.approx(
        expected_intercept
```

Trick 2: Use inequalities



Trick 2: Use inequalities



Recommendations

- Do not leave models untested just because they are complex.
- Perform as many sanity checks as possible.

Using the model

```
from data.preprocessing_helpers import preprocess
from features.as_numpy import get_data_as_numpy_array
from models.train import (
  split_into_training_and_testing_sets, train_model
preprocess("data/raw/housing_data.txt",
           "data/clean/clean_housing_data.txt"
data = get_data_as_numpy_array(
    "data/clean/clean_housing_data.txt", 2
training_set, testing_set = (
    split_into_training_and_testing_sets(data)
slope, intercept = train_model(training_set)
```

```
train_model(training_set)
```

151.78430060614986 17140.77537937442

Testing model performance

```
def model_test(testing_set, slope, intercept):
    """Return r^2 of fit"""
```

- Returns a quantity r^2 .
- Indicates how well the model performs on unseen data.
- Usually, $0 \le r^2 \le 1$.
- $r^2 = 1$ indicates perfect fit.
- $r^2 = 0$ indicates no fit.
- Complicated to compute r^2 manually.

Let's practice writing sanity tests!

UNIT TESTING FOR DATA SCIENCE IN PYTHON



Testing plots

UNIT TESTING FOR DATA SCIENCE IN PYTHON



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Pizza without cheese!



This lesson: testing matplotlib visualizations



```
data/
src/
|-- data/
|-- features/
|-- models/
|-- visualization
| |-- __init__.py
tests/
```

```
data/
src/
|-- data/
|-- features/
|-- models/
|-- visualization
| |-- __init__.py
| |-- plots.py
```

```
def get_plot_for_best_fit_line(slope,
                                intercept,
                                x_array,
                                y_array,
                                title
    H \cap H
    slope: slope of best fit line
    intercept: intercept of best fit line
    x_array: array containing housing areas
    y_array: array containing housing prices
```

```
data/
src/
|-- data/
|-- features/
|-- models/
|-- visualization
| |-- __init__.py
| |-- plots.py
tests/
```

```
def get_plot_for_best_fit_line(slope,
                                intercept,
                                x_array,
                                y_array,
                                title
                                ):
    H \cap H
    slope: slope of best fit line
    intercept: intercept of best fit line
    x_array: array containing housing areas
    y_array: array containing housing prices
    title: title of the plot
    0.00
```

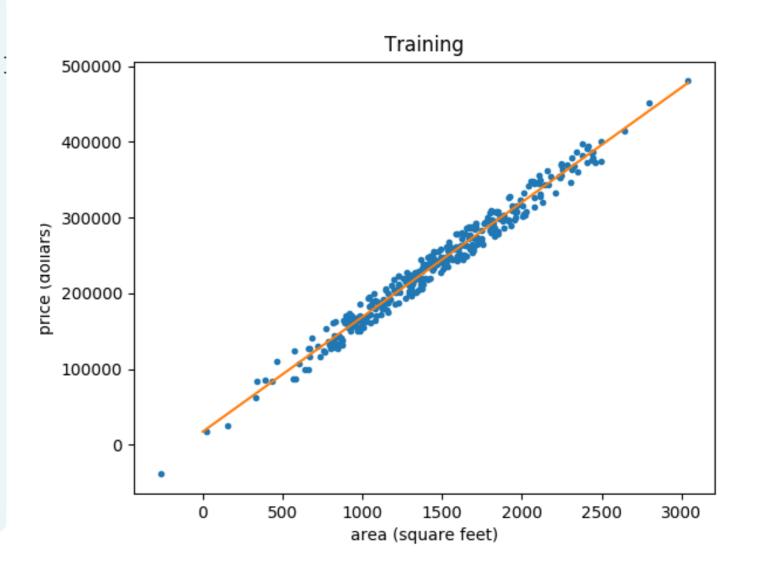
```
data/
src/
|-- data/
|-- features/
|-- models/
|-- visualization
| |-- __init__.py
| |-- plots.py
```

```
def get_plot_for_best_fit_line(slope,
                                intercept,
                                x_array,
                                y_array,
                                title
    H \cap H
    slope: slope of best fit line
    intercept: intercept of best fit line
    x_array: array containing housing areas
    y_array: array containing housing prices
    title: title of the plot
    Returns: matplotlib.figure.Figure()
    0.00
```

```
data/
src/
|-- data/
|-- features/
|-- models/
|-- visualization
| |-- __init__.py
| |-- plots.py
tests/
```

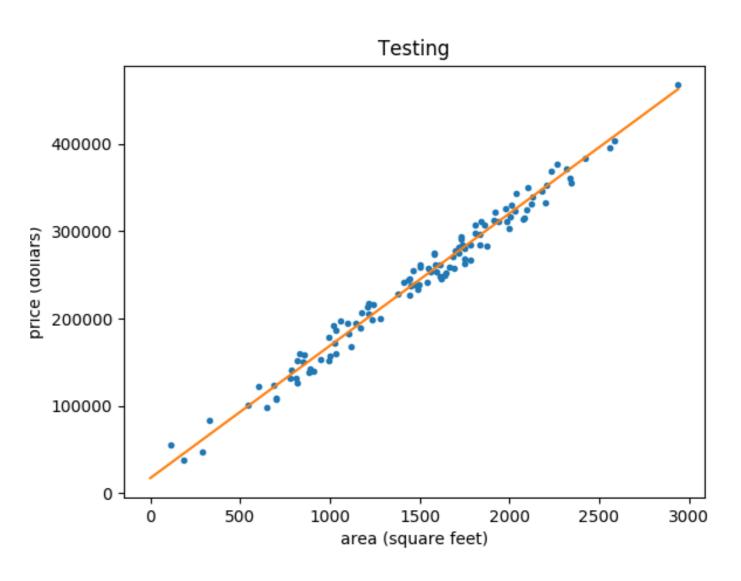
Training plot

```
from visualization import get_plot_for_best_fit_!
preprocess(...)
data = get_data_as_numpy_array(...)
training_set, testing_set = (
    split_into_training_and_testing_sets(data)
slope, intercept = train_model(training_set)
get_plot_for_best_fit_line(slope, intercept,
    training_set[:, 0], training_set[:, 1],
    "Training"
```



Testing plot

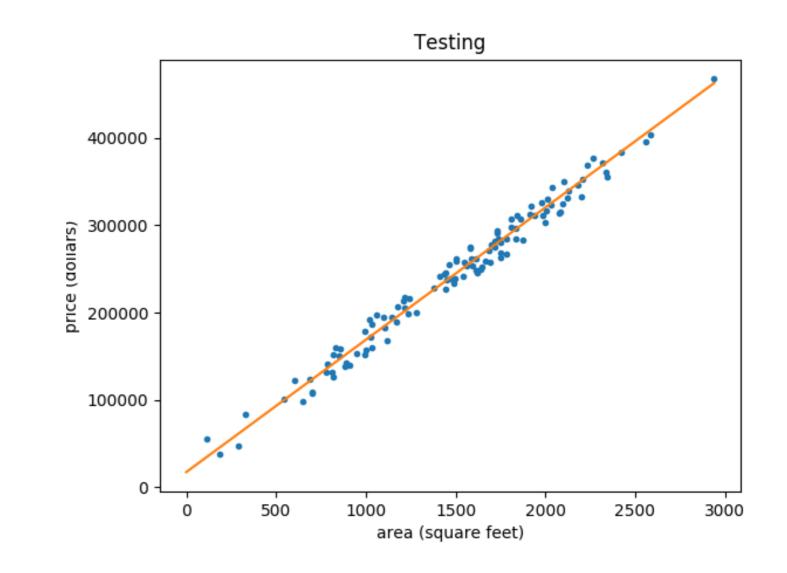
```
from visualization import get_plot_for_best_fit_!
preprocess(...)
data = get_data_as_numpy_array(...)
training_set, testing_set = (
                        split_into_training_and_testing_sets(data)
slope, intercept = train_model(training_set)
get_plot_for_best_fit_line(slope, intercept,
                       training_set[:, 0], training_set[:, 1],
                         "Training"
get_plot_for_best_fit_line(slope, intercept,
                       testing_set[:, 0], testing_set[:, 1], "Testing_set[:, 1], "Testing
```



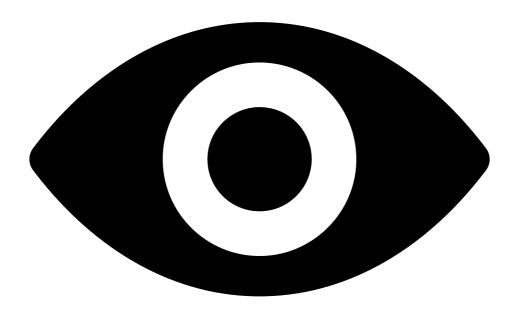
Don't test properties individually

matplotlib.figure.Figure()

- Axes
 - configuration
 - style
- Data
 - style
- Annotations
 - style
- ...



Testing strategy for plots



Testing strategy for plots

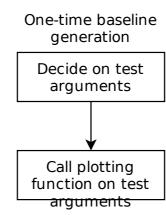
One-time baseline generation

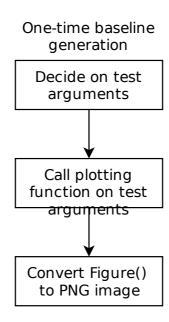


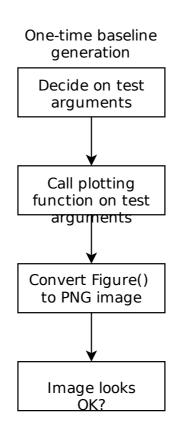
One-time baseline generation

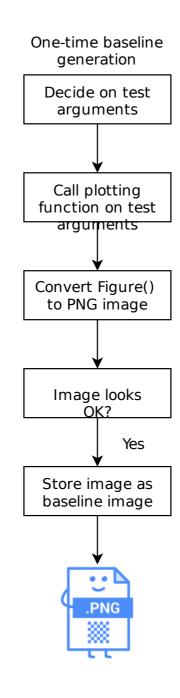
Decide on test arguments

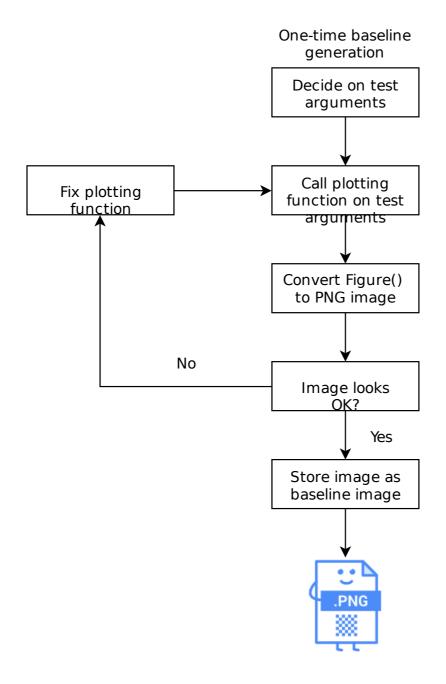


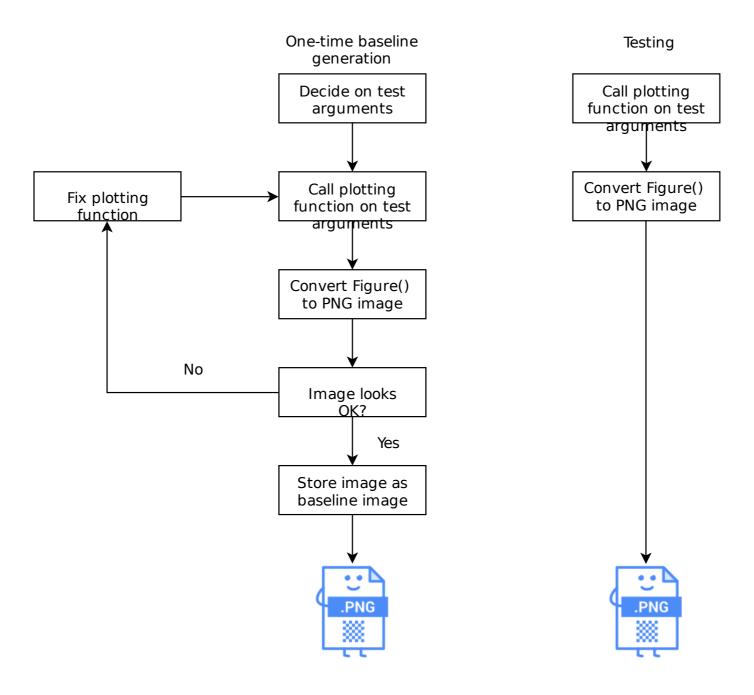


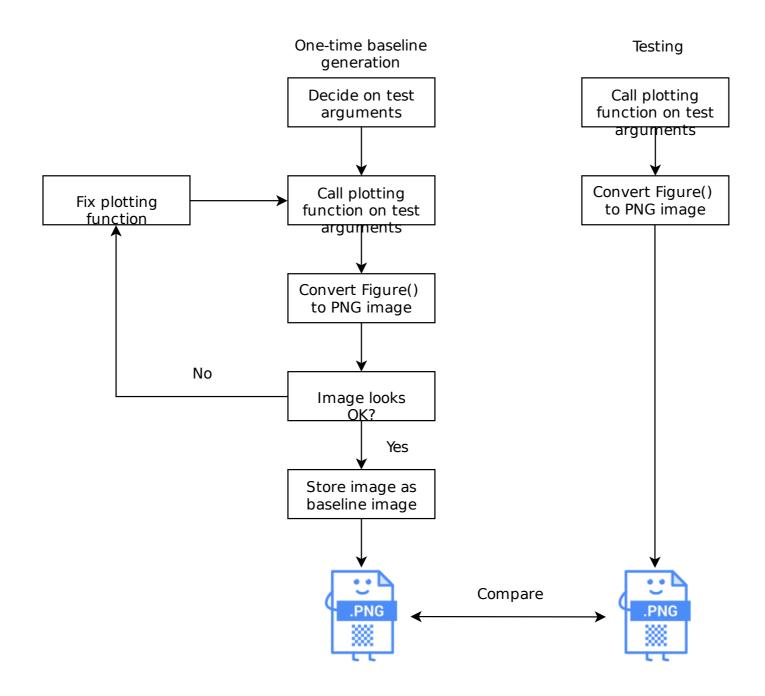












pytest-mpl

- Knows how to ignore OS related differences.
- Makes it easy to generate baseline images.

pip install pytest-mpl

An example test

```
import pytest
import numpy as np
from visualization import get_plot_for_best_fit_line
def test_plot_for_linear_data():
    slope = 2.0
   intercept = 1.0
   x_{array} = np.array([1.0, 2.0, 3.0]) # Linear data set
   y_{array} = np.array([3.0, 5.0, 7.0])
   title = "Test plot for linear data"
    return get_plot_for_best_fit_line(slope, intercept, x_array, y_array, title)
```

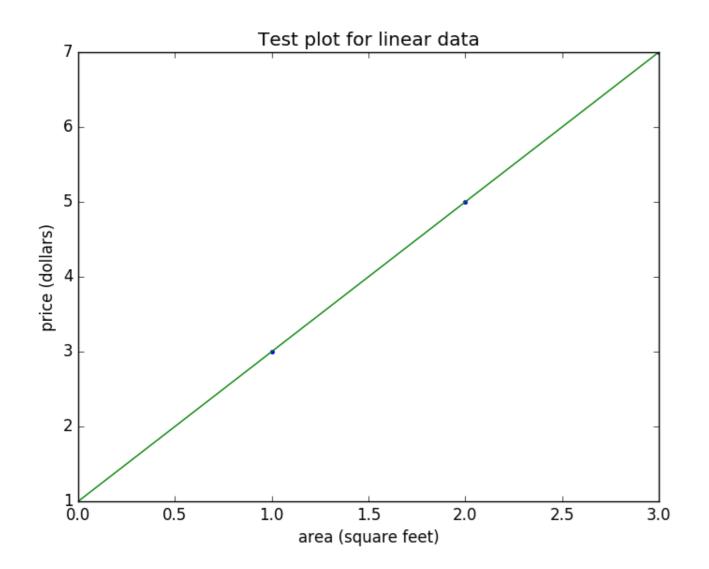
An example test

```
import pytest
import numpy as np
from visualization import get_plot_for_best_fit_line
@pytest.mark.mpl_image_compare # Under the hood baseline generation and comparison
def test_plot_for_linear_data():
   slope = 2.0
   intercept = 1.0
   x_{array} = np.array([1.0, 2.0, 3.0]) # Linear data set
   y_{array} = np.array([3.0, 5.0, 7.0])
   title = "Test plot for linear data"
    return get_plot_for_best_fit_line(slope, intercept, x_array, y_array, title)
```

Generating the baseline image

Generate baseline image

Verify the baseline image



```
data/
src/
tests/
 -- data/
 -- features/
 -- models/
 -- visualization
    -- __init__.py
     -- test_plots.py
                        # Test module
    -- baseline
                         # Contains baselines
        |-- test_plot_for_linear_data.png
```

Run the test

```
!pytest -k "test_plot_for_linear_data" --mpl
```



Reading failure reports

```
!pytest -k "test_plot_for_linear_data" --mpl
```

```
_____ TestGetPlotForBestFitLine.test_plot_for_linear_data ______
Error: Image files did not match.
 RMS Value: 11.191347848524174
 Expected:
   /tmp/tmplcbtsb10/baseline-test_plot_for_linear_data.png
 Actual:
   /tmp/tmplcbtsb10/test_plot_for_linear_data.png
 Difference:
   /tmp/tmplcbtsb10/test_plot_for_linear_data-failed-diff.png
 Tolerance:
======== 1 failed, 36 deselected in 1.13 seconds =========
```



Yummy!



Let's test plots!

UNIT TESTING FOR DATA SCIENCE IN PYTHON



Congratulations

UNIT TESTING FOR DATA SCIENCE IN PYTHON



Dibya Chakravorty
Test Automation Engineer









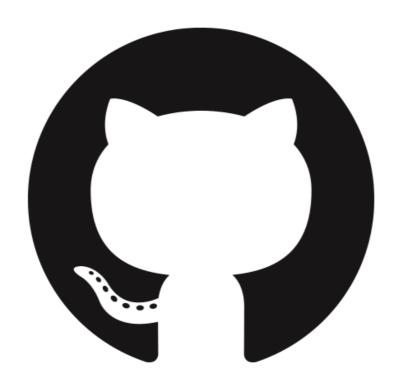
You learned a lot



- Testing saves time and effort.
- pytest
 - Testing return values and exceptions.
 - Running tests and reading the test result report.
- Best practices
 - Well tested function using normal, special and bad arguments.
 - TDD, where tests get written before implementation.
 - Test organization and management.
- Advanced skills
 - Setup and teardown with fixtures, mocking.
 - Sanity tests for data science models.
 - Plot testing.

Code for this course

https://github.com/gutfeeling/univariate-linear-regression



Icon sources

Icons made by the following authors from flaticon.com.

- Freepik
- Smashicons
- Vectors Market
- Kiranshastry
- Dimitry Miroliubov
- Creatica Creative Agency
- Gregor Cresnar

Image sources

- 1. https://chibird.com/post/20998191414/i-make-a-lot-of-procrastination-drawings-theyre
- 2. http://www.dekoleidenschaft.de/ratgeber/10-tipps-fuer-mehr-ordnung-im-kleiderschrank/
- 3. http://me-monaco.me/paper-storage-box-with-lid/
- 4. https://towardsdatascience.com/random-forests-and-decision-trees-from-scratch-in-python-3e4fa5ae4249
- 5. https://towardsdatascience.com/demystifying-support-vector-machines-8453b39f7368
- 6. https://www.bbc.co.uk/bbcthree/article/b290ff0e-1d75-43b1-8ff1-a9ac80d4d842

I wish you all the best!

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