Testing scientific code

Because you're worth it

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Introduction to testing project

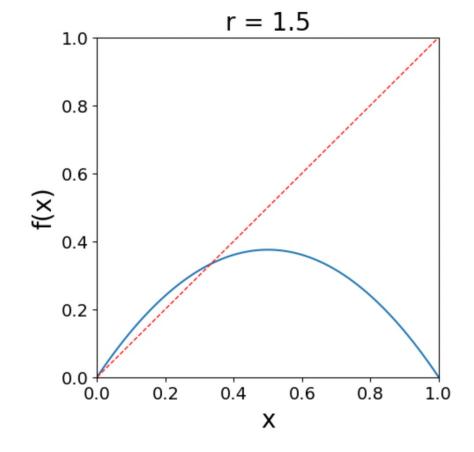
Fork repository testing_project

• Simple, discrete model for population growth

growth rate, 0...4
$$f(x) = r * x * (1 - x)$$
reproduction starvation

current population size, as fraction of maximum possible size, 0...1



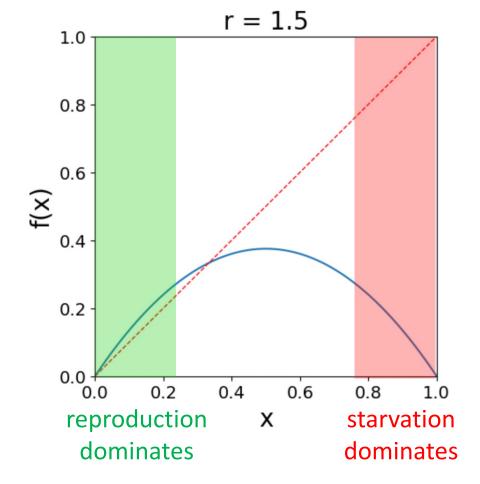


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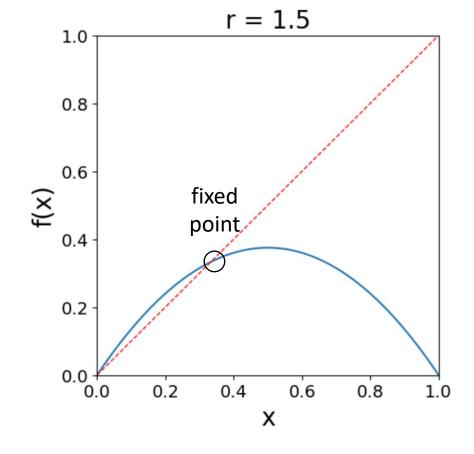


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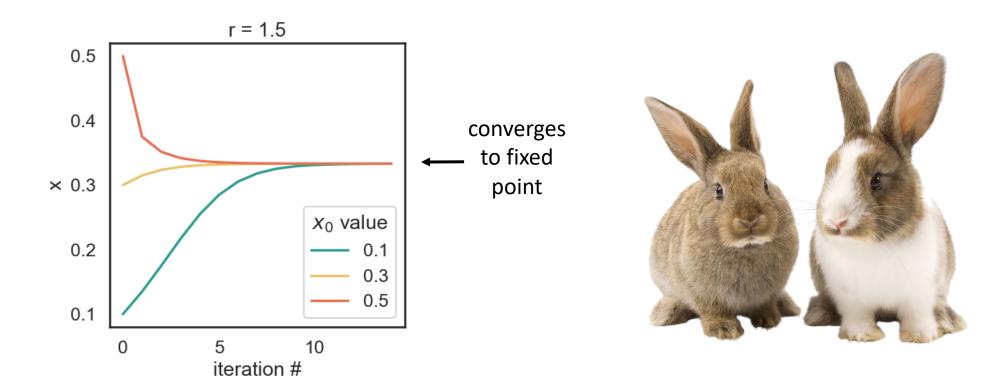
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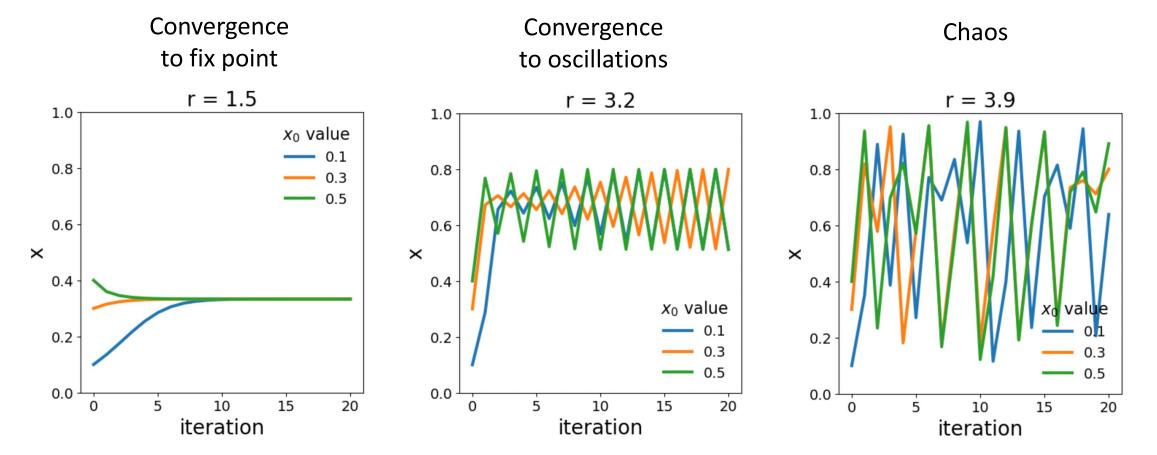




- x_0 : initial population size
- Iterated function: $f(x_0) = x_1 \rightarrow f(x_1) = x_2 \rightarrow f(x_2) = x_3$



Different growth rates lead to a variety of population dynamics





What a good test looks like

- What does a good test look like? What should I test?
- Good:
 - Short and quick to execute
 - Easy to read
 - Tests one thing
- Bad:
 - Relies on data files
 - Messes with "real-life" files, servers, databases

Basic structure of test

- A good test is divided in three parts:
 - Given: Put your system in the right state for testing
 - Create data, initialize parameters, define constants...
 - When: Execute the feature that you are testing
 - Typically, one or two lines of code
 - Then: Compare outcomes with the expected ones
 - Define the expected result of the test
 - Set of *assertions* that check that the new state of your system matches your expectations

Test simple but general cases

- Start with simple, general case
 - Take a realistic scenario for your code, try to reduce it to the simplest example
- Example: Tests for 'lower' method of strings

```
def test_lower():
    # Given
    string = 'HeLlO wOrld'
    expected = 'hello world'

# When
    output = string.lower()

# Then
    assert output == expected
```

Test special cases and boundary conditions

- Code often breaks in corner cases: empty lists, None, NaN, 0.0, lists with repeated elements, non-existing file, ...
- This often involves making design decision: handle corner case with special behavior, or raise a meaningful exception?

```
def test_lower_empty_string():
    # Given
    string = ''
    expected = ''

# When
    output = string.lower()

# Then
    assert output == expected
```

Other good corner cases for string.lower():

```
'do-nothing case': string = 'hi'
symbols: string = '123 (!'
```

Common for-loop pattern for testing multiple cases

Often these cases are collected in a single test:

Hands-on!

- Take a look at the logistic map f(x) = r * x * (1 x)
- or, in Python

```
def f(x, r):
    """ Compute the logistic map for a given value of x and r. """
    return r * x * (1 - x)
```

- What should we test?
 - Generic cases
 - Corner cases

Hands-on! for

- In the testing_project repository, open the file logistic.py and implement the logistic function, f(x, r)
- In test_logistic.py we already added a reference test for these corner cases:
 - x=0, $r=1.1 \Rightarrow f(x, r)=0$
 - x=1, $r=3.7 \Rightarrow f(x, r)=0$
- Add a new test for these generic cases using the for-loop pattern:
 - x=0.1, $r=2.2 \Rightarrow f(x, r)=0.198$
 - x=0.2, $r=3.4 \Rightarrow f(x, r)=0.544$
 - x=0.5, r=2 => f(x, r)=0.5

The for-loop pattern can be improved

- It is repetitive to write the for-loop pattern
- If one of the cases break, it can be complicated to figure out which one

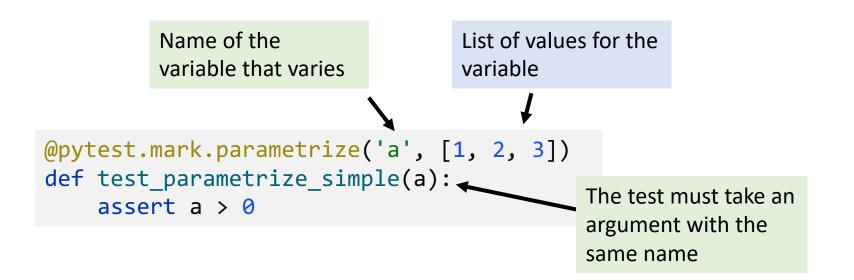
- pytest has many helpers for simplifying common testing cases!
- One of them is the parametrize decorator, that simplifies running the same test with multiple cases

Simple example

```
def test_for_loop_simple():
    cases = [1, 2, 3]
    for a in cases:
        assert a > 0
```

test_for_loop_simple
runs once and loops over
3 test cases

Simple example, with the parametrize decorator



test_parametrize_simple
runs 3 times
with a=1, a=2, and a=3

Simple example, with the parametrize decorator

```
platform darwin — Python 3.11.3, pytest—7.3.1, pluggy—1.0.0 — /Users/pietro.berkes/miniconda3/envs/aspp/bin/python cachedir: .pytest_cache rootdir: /Users/pietro.berkes/o/ASPP/testing_project/demos plugins: anyio—3.5.0 collected 3 items

test_parametrize.py::test_parametrize_simple[1] PASSED
test_parametrize.py::test_parametrize_simple[2] PASSED
test_parametrize.py::test_parametrize_simple[3] PASSED
[100%]
```

== **3 passed** in 0.00s ==

pytest automatically creates one separate test for each test case

Example with multiple values

 This is a more typical case with several input values and the expected result of the test

```
def test_for_loop_multiple():
    cases = [
        (1, 'hi', 'hi'),
        (2, 'no', 'nono')
    ]
    for a, b, expected in cases:
        result = b * a
        assert result == expected
```

test_for_loop_multiple
runs once and loops over
2 test cases

Same example, with the parametrize decorator

```
Name of all the variables, separated by commas in one string

Opytest.mark.parametrize('a, b, expected', [(1, 'hi', 'hi'), (2, 'no', 'nono')])

def test_parametrize_multiple(a, b, expected):
    result = b * a
    assert result == expected

The test must take arguments with the same names as in the string
```

```
test_parametrize_multiple
runs 2 times with
1) a=1 b='hi' expected='hi'
and
2) a=2 b='no', expected='nono'
```

Same example, with the parametrize decorator

```
Name of all the variables, separated by commas in one string

Opytest.mark.parametrize('a, b, expected', [(1, 'hi', 'hi'), (2, 'no', 'nono')])

def test_parametrize_multiple(a, b, expected):
    result = b * a
    assert result == expected

The test must take arguments with the same names as in the string
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pytest automatically creates one separate test for each test case

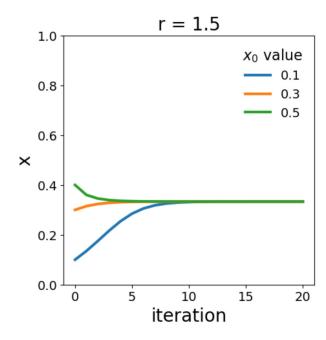
Hands-on!

• Rewrite the test with the generic cases for the logistic map using parametrize

• Reference example for the corner cases test:

Hands-on! Simulate a population over time

- 1. Implement a function iterate_f that runs f for it iterations. Write tests for the following cases:
 - x=0.1, r=2.2, it=1
 => iterate f(it, x, r)=[0.1, 0.198]
 - x=0.2, r=3.4, it=4
 => iterate_f(it, x, r)=[0.2, 0.544, 0.843418, 0.449019, 0.841163]
 - x=0.5, r=2, it=3
 => iterate_f(it, x, r)=[0.5, 0.5, 0.5]
- (Bonus) Import the plot_trajectory function from the plot_logistic module and use it to visualize the trajectories generated by your code.
 - Try with values r<3, and 3<r<4 to get an intuition for how the function behaves differently with different parameters.



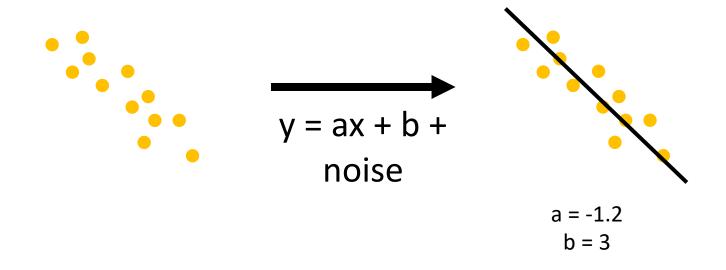


Strategies for testing scientific code

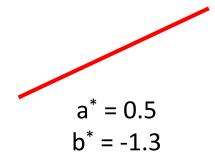
Strategies for testing learning algorithms

- Learning algorithms can get stuck in local maxima, the solution for general cases might not be known (e.g., unsupervised learning)
- Turn your validation cases into tests
- Stability tests:
 - Start from final solution; verify that the algorithm stays there
 - Start from solution and add a small amount of noise to the parameters; verify that the algorithm converges back to the solution
- Parameter Recovery: Generate synthetic data from the model with known parameters, then test that the code can learn the parameters back

Learning algorithms fit the parameters of a model to observed data



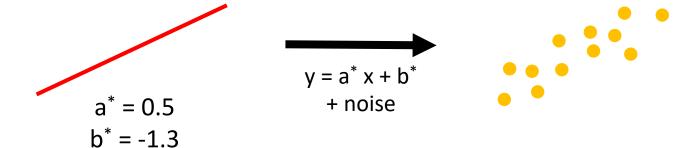
1) Fix initial parameters



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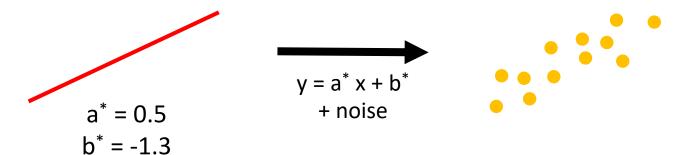
1) Fix initial parameters

2) Generate synthetic data



1) Fix initial parameters

2) Generate synthetic data



a =
$$0.5098$$

b = -1.287

3) Run the algorithm

$$y = ax + b$$
+ noise

2) Generate synthetic data 1) Fix initial parameters $y = a^* x + b^*$ $a^* = 0.5$ + noise $b^* = -1.3$ 4) Compare 3) Run the algorithm a = 0.5098b = -1.287y = ax + b+ noise

Hands-on! Recover the population growth, r

In the module logistic_fit, we implemented a function fit_r
that, given a population trajectory, finds the value of r that
generated it

• For example:

```
In [1]: from logistic import iterate_f
In [2]: from logistic_fit import fit_r
In [3]: xs = iterate_f(it=23, x0=0.3, r=3.421)

In [4]: fit_r(xs)
Out[4]: 3.4210000000000000
```

Hands-on!

- Write a test for the function fit_r using the parameters recovery method
- The test should
 - 1. Set a initial value for x0 and r
 - 2. Use iterate_f to generate a population trajectory
 - 3. Pass the population trajectory to fit_r and collect the result parameters
 - 4. Check that the fitted r is close enough to the original r

```
In [1]: from logistic import iterate_f
In [2]: from logistic_fit import fit_r
In [3]: xs = iterate_f(it=23, x0=0.3, r=3.421)

In [4]: fit_r(xs)
Out[4]: 3.42100000000000003
```

Randomness in Testing

- Using randomness in testing can be useful
 - To check that the code is stable and works correctly in many different cases
 - To find corner cases or numerical problems

```
def test_logistic_fit_randomized():
    random_state = np.random.RandomState(SEED)
    for _ in range(100):
        x0 = random_state.uniform(0.0001, 0.9999)
        r = round(random_state.uniform(0.001, 3.999), 3)

        xs = iterate_f(it=17, x0=x0, r=r)
        recovered_r = fit_r(xs)

        assert_allclose(r, recovered_r, atol=1e-3)
```



Random Seeds and Reproducibility

- When running tests that involve radomness and some test doesn't pass it is vital to be able to reproduce that test exactly!
- Computers produce pseudo-random numbers: setting a seed resets the basis for the random number generator
- This is essential for reproducibility
- At a minimum, you should manually set the seed for each of your random tests

```
SEED = 42
random_state = np.random.RandomState(SEED)
random_state.rand()
```

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Hands On!

- a) Write a randomized test that checks that for r=1.5, for any random starting point x0, the logistic equation converges to 1/3
 - Write a for loop of 100 iterations, in each iteration create a random x0
 - For each value of x0, test that after many iterations in iterate_f the value of x is equal to 1/3

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A Pytest Solution

- Non-scientific coding uses random testing more rarely, so there is no helper tools for that in pytest
- However, in scientific coding it is quite common
- What do we want?
 - For each (random) test there should be a seed
 - For each run of the test, the seed should be different
 - That seed should be printed with the test result
 - It needs to be possible to explicitely run the test again with that seed!

Fixtures (minimal solution)

Fixtures are functions that are run before the tests are executed

```
import numpy as np
import pytest
# set the random seed for once here
SEED = np.random.randint(0, 2**31)
@pytest.fixture
def random_state():
   print(f'Using seed {SEED}')
   random state = np.random.RandomState(SEED)
   return random state
def test_something(random_state):
   random state.rand()
```

If an input argument of a test matches the name of a fixture, then the fixture is called and the return value assigned to the argument.

pytest handles that automatically as part of running the test suite

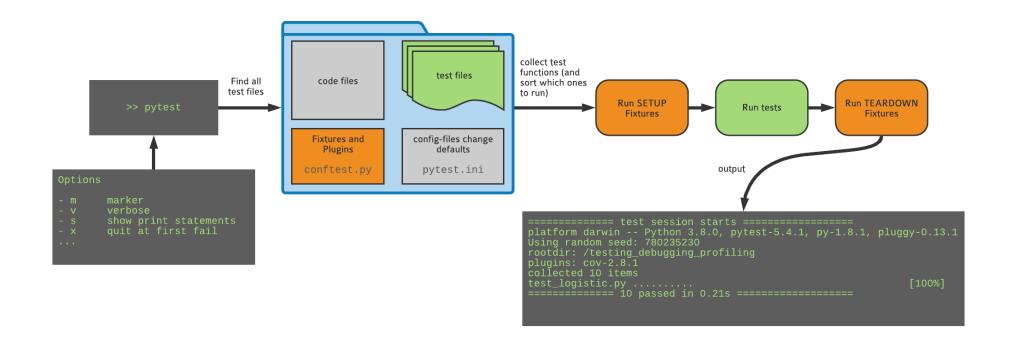
Hands On!

- a) Write a randomized test that checks that fit_r can recover r for any random value of x0 and r
- b) Add a fixture at the top of your test file, that lets you print the seed to the console.

```
| test session starts | test session starts
```

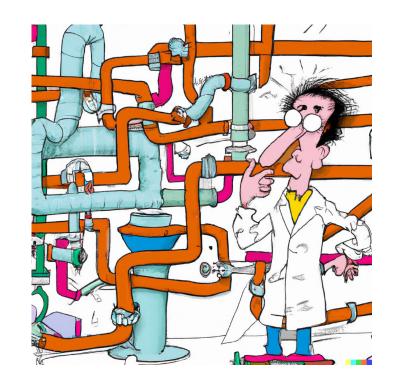
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What happens when you run pytest



Fixtures (real solution)

- conftest.py is a special pytest config file (don't import it!)
- conftest.py can be used to define custom behavior or plugins. Fixtures can also be defined here, so that they can be used by all tests.
- See the file demos/conftest_example.py in the repo you forked. If you move it to the main folder and rename it, the functions defined there select a seed for each test and allow you to pass a seed on the command line using --seed 123



Hands On!

- a) Write a randomized test that checks that fit_r can recover r for any random value of x0 and r
- b) Add a fixture at the top of your test file, that lets you print the seed to the console.
- c) Add the conftest.py file the root directory of the project (hint: it is hiding in the demos folder!). It sets a random seed before each run and makes it possible to reproduce failures in random tests
- d) conftest.py defines a new random_state fixture, modify your test accordingly
- e) Check that the console output of pytest now includes the seed!



Other commonly used helpers in pytest

Decorating "special" tests

• @xfail: Expected failure, outputs an "x" (or "X") in the report

```
@pytest.mark.xfail
def test_something():
    ...
```

• @skip: Skip test, useful e.g. when the feature doesn't exist yet

```
@pytest.mark.skip(reason="functionality not yet
implemented")
def test_something():
...
```

 @skipif: Skip the test if a condition is met, useful for tests that only works on a specific platform, or for a specific version of Python

Marking tests with custom markers

- If you have lots of tests, you can categorize them with your own markers
 - although for custom mark names you need to register the marks "pytest.ini"
 - https://docs.pytest.org/en/7.1.x/example/markers.html#registering-markers
- Example:
 - Smoke tests check for really basic features: run these frequently
 - Other tests may be many or too slow to run every time and test for more edge cases

```
@pytest.mark.smoke
def test_something_basic():
    ...
```

```
> pytest -m smoke
> pytest -m "smoke and not slow"
```

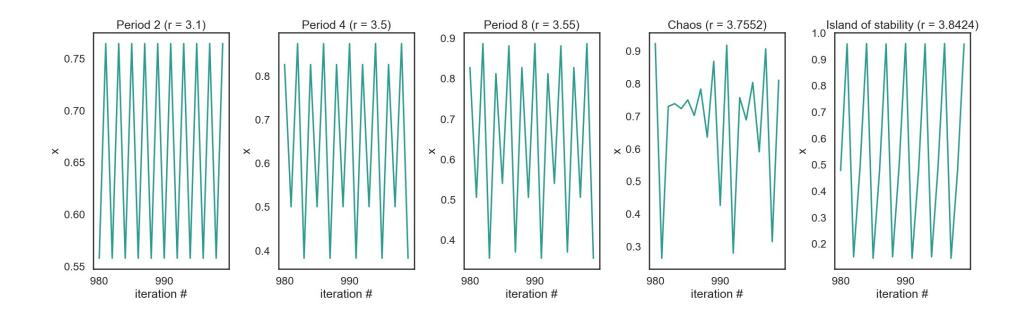
Writing temporary files: tmp_path

- To test functions that write to disk without leaving around the files when the test is finished, use the tmp_path fixture
- The value of tmp_path is a pathlib.Path object
- The directory is created at the start of the test, and removed at the end

```
def test_create_file(tmp_path):
    d = tmp_path / "sub"
    d.mkdir()
    p = d / "hello.txt"
    content = "some random text"
    p.write_text(content)
    assert p.read_text() == content
    assert len(list(tmp_path.iterdir())) == 1
All you need to do is add an argument with this exact name
```

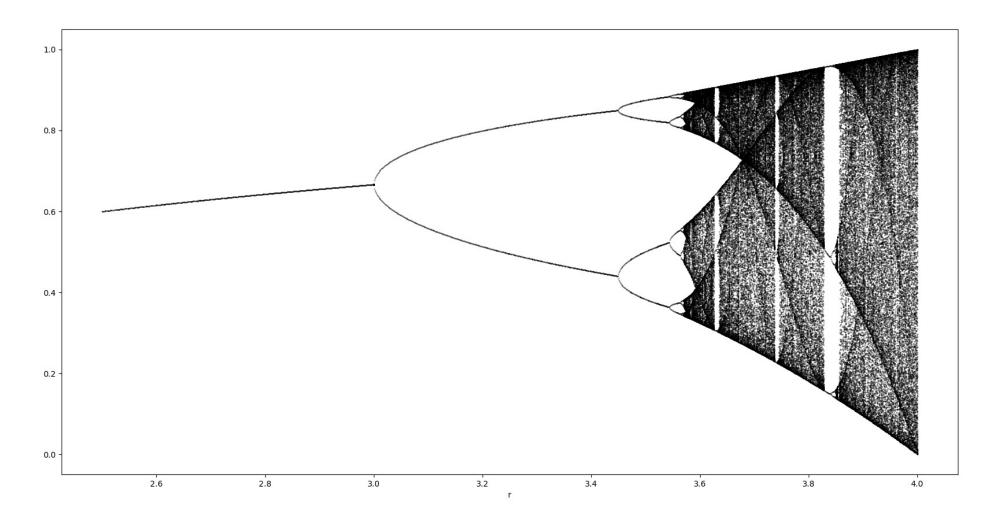
Final exercise

Excursion: Logistic map



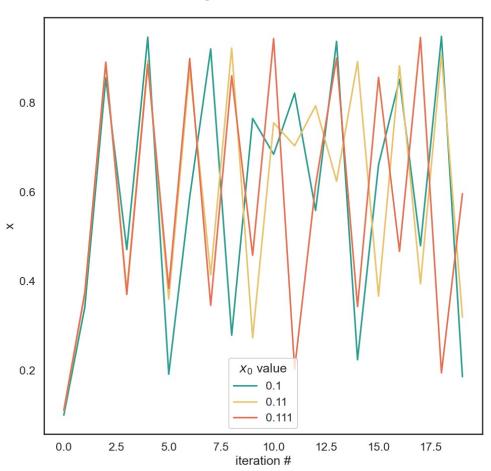
- Between r=3 and r=4 the logistic map has a range of behaviors
- Periodic vs. chaotic

Excursion: Logistic map



Excursion: Logistic map and chaos





- Sensitive Dependence on Initial Conditions (SDIC)
- Even starting points that are very close quickly diverge to completely different itineraries
- This is called the "Butterfly effect"







Hands on!

Some r values for 3 < r < 4 have some interesting properties: a chaotic trajectory neither diverges nor converges.

- a) Use the plot_bifurcation function from the plot_logfun module using your implementation of f and iterate to look at the bifurcation diagram. The function generates an output image, bifurcation_diagram.png
- b) Write a test that checks for chaotic behavior when r=3.8. Run the logistic map for 100'000 iterations and verify the conditions for chaotic behavior:
 - 1) The function is deterministic: this does not need to be tested in this case
 - 2) Orbits must be bounded: check that all values are between 0 and 1
 - 3) Orbits must be aperiodic: check that the last 1000 values are all different
 - 4) Sensitive dependence on initial conditions: this is the bonus exercise (in readme)

The test should check conditions 2) and 3)!

Testing is good for your self-esteem

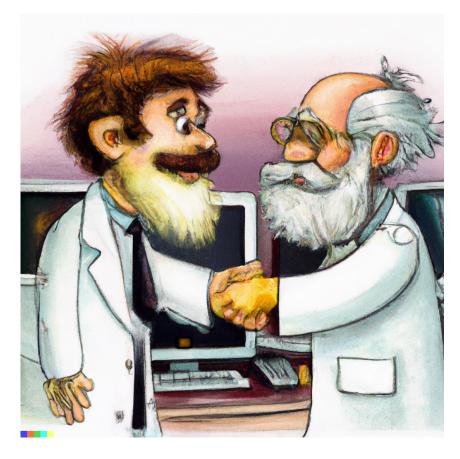
- Immediately: Always be confident that your results are correct, whether your approach works of not
- In the future: save your future self some trouble!
- If you are left thinking "it's cool but I cannot test my code because XYZ", talk to us during the week and we'll show you how to do it;-)

You, in 2023



You, in 2024

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Up next:







