

Week 8 - Testing

Testing your code is essential to verify that it correctly behaves as you meant it to.

Creating tests early in the development of a project usually saves you a lot of time looking for hidden bugs once things get more complicated – we strongly recommend this!

This is particularly important when you collaborate with other developers, but also when you use or refactor your code, which you might have written a long time before. Writing tests for specific functionalities ensures that changes to the code do not affect the correctness of the results.

This week we will focus on *unit testing*, a form of automated testing aimed at assessing the behavior of individual parts of your code in isolation (i.e., units). We will use the python libraries `doctest` and `pytest` to write tests.

Make sure to release your code by 10am next Monday (release notes)!

doctest

Doctest¹ is a library which allows you to write tests together with the documentation of a function or of a class. Doctest allows you to write examples about the function that you document, by specifying which output it should return, when provided with a specific input. To understand how doctest works, you can have a look at the doctests in the demo-project (<https://github.com/EPFL-BIO-210/demo-project>). For example, the function `dX_dt` in `LotkaVolterraModel.py`:

```
def dX_dt(X, a=1.0, b=0.1, c=1.5, d=0.75):
    """
    Computes the growth rate of fox and rabbit populations based on system state (X)
    and parameters (a,b,c,d)
    Parameters
    -----
    X : array or tuple
        [prey_count, predator_count]
    a : float, optional
        natural growth rate of the prey (rabbit)
    b : float, optional
        natural dying rate of the prey
    c : float, optional
        natural growth rate of the predator (fox)
    d : float, optional
        natural dying rate of the predator
    Returns
    -----
    numpy array
        [change of prey_count, change of predator_count]

    Examples
    -----
    >>> dX_dt(np.ones(2),1,0.1,1.5,.75)
    array([ 0.9 , -1.425])
    >>> dX_dt(np.zeros(2),1,0.1,1.5,.75)      # zero is a fixpoint
    array([0., 0.]
```

¹Source: <https://docs.python.org/3/library/doctest.html>

```
"""
return np.array([a * X[0] - b * X[0] * X[1], -c * X[1] + d * b * X[0] * X[1]])
```

Here, when doctest runs, it parses the docstrings and tests the function after the symbols ">>>". The subsequent row defines the expected output.

You can then test the functions in the `LotkaVolterraModel` module with the command `python LotkaVolterraModel.py`. It should have no output (if all tests pass). You can also print a report running by specifying the flag `-v` (verbose): `python LotkaVolterraModel.py -v`.

Doctests are a convenient solution to test your code while writing the documentation, especially for simple function calls. When testing more complex functionalities, `pytest` might be more appropriate!

pytest

Pytest is a popular unit testing library for Python. It makes use of the core concept of unit testing: the *assertion*. Writing an assertion is like asking the code to verify that a certain condition (equality, inequality, truth, ...) is verified. We can have a look at the following example from the `pytest` docs², where the function `func` is tested:

```
# content of test_sample.py
def func(x):
    return x + 1

def test_answer():
    assert func(3) == 5
```

You can then run the example by writing on the terminal `pytest test_sample.py`. You should observe a failure report as `func(3)` does not return 5.

Remark: Test functions must begin with the prefix `test_`.

Remark: The scripts that contain your test function(s) have to begin with `test_` or end with `_test`.

If you want to test more than one function, you can add multiple tests (as functions) in a single script (as a reference, look at the file `test_LVM.py` of the demo project: https://github.com/EPFL-BIO-210/demo-project/blob/feature/doctests/test_LVM.py). When writing unit-tests, each test should focus on one aspect of the function to test, so we expect to have just one or few assertions per test.

In the demo project, we also show how to integrate doctests in `pytest`. For example, to execute the doctests of the file `LotkaVolterraModel.py`, it is possible to add in a `pytest` executing the command `assert doctest.testmod(LotkaVolterraModel, raise_on_error=True)`. This is useful to run your whole test-suite with a single call.

You can run the demo-project tests by writing on the terminal `pytest test_LVM.py`. Tests can both verify regular behavior (`assert`) or exceptional behavior (`pytest.raises(Exception)`).

Line coverage

A widely used metric to evaluate how exhaustively your code is tested is the *line coverage*, which counts the lines of code of your project which are executed at least once by a test.

Importantly, even with 100% line coverage you cannot be sure that your code is bug-free. It is up to you to design tests which perform important checks in your code, at least guaranteeing that

²Source: <https://docs.pytest.org/en/6.2.x/getting-started.html#create-your-first-test>

it behaves as expected in those particular circumstances. Python provides the coverage module³ to compute the line coverage of your tests. When running the test you created with pytest, you can also get the coverage by running in the terminal `coverage run -m pytest test_LVM.py` to collect the results. To see the coverage report, simply type `coverage report -m`.

This week your task is to develop unit tests for your project, to get as close as possible to 100% line coverage. We list some reasonable tests for the two projects below, but you can design many more.

1 Hopfield network

- Test that the weights matrix with the Hebbian weights has the correct size, that its values are in the correct range, that it is symmetric, that the elements on the diagonal are 0
- Consider the following patterns:

$$\begin{pmatrix} 1 & 1 & -1 & -1 \end{pmatrix} \quad \begin{pmatrix} 1 & 1 & -1 & 1 \end{pmatrix} \quad \begin{pmatrix} -1 & 1 & -1 & 1 \end{pmatrix}$$

create a test to check that the Hebbian weights are (for doctests, be careful at using the correct number of digits)

$$\begin{pmatrix} 0. & 0.33333333 & -0.33333333 & -0.33333333 \\ 0.33333333 & 0. & -1. & 0.33333333 \\ -0.33333333 & -1. & 0. & -0.33333333 \\ -0.33333333 & 0.33333333 & -0.33333333 & 0. \end{pmatrix} \quad (1)$$

while the Storkey weights are:

$$\begin{pmatrix} 1.125 & 0.25 & -0.25 & -0.5 \\ 0.25 & 0.625 & -1. & 0.25 \\ -0.25 & -1. & 0.625 & -0.25 \\ -0.5 & 0.25 & -0.25 & 1.125 \end{pmatrix}$$

Hint: the function `allclose` of the library `numpy` might be useful to compare matrices of floats.

- Test the evolution of the system, for example, by checking that it converges to the correct pattern and that the energy function is non-increasing
- Devise additional tests to reach as high line coverage as possible. You can also improve your code to deal with violations of the requirements of the input (e.g., raise an exception if the patterns are non-binary or if they are not unique)

2 Turing pattern formation

- Create tests that check the value of the functions f , g and h .

Use the following parameters: $u = 1$, $v = 2$, $\alpha = 1.5$, $K = 0.125$, $\rho = 13$, $a = 103$, $b = 77$,

the corresponding values must be: $h = 12.235294117647058$, $f = 89.76470588235294$, $g = 100.26470588235294$

³Source: <https://coverage.readthedocs.io>

- Test that the function `diffusion` applied to the matrix

$$\begin{pmatrix} 1 & 2 & 3 & 4 & 5 \\ 2 & 3 & 4 & 5 & 6 \\ 3 & 4 & 5 & 6 & 7 \\ 4 & 5 & 6 & 7 & 8 \\ 5 & 6 & 7 & 8 & 9 \end{pmatrix}$$

returns the following matrix:

$$\begin{pmatrix} 2 & 1 & 1 & 1 & 0 \\ 1 & 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 & -1 \\ 1 & 0 & 0 & 0 & -1 \\ 0 & -1 & -1 & -1 & -2 \end{pmatrix}$$

- Use the following parameters: $M = 3$, $N = 3$, $\Delta x = 0.1$, $\Delta t = 0.0001$, $\alpha = 1.5$, $K = 0.125$, $\rho = 13$, $a = 103$, $b = 77$, $d = 7$, $\gamma = 0.5$ and start from the initial states

$$u_0 = \begin{pmatrix} 0 & 1 & 2 \\ 3 & 4 & 5 \\ 6 & 7 & 8 \end{pmatrix} \quad v_0 = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

Check that the values of u and v after one iteration are:

$$u_1 = \begin{pmatrix} 0.04515 & 1.03448824 & 2.02393571 \\ 3.01347805 & 4.00309286 & 4.99276301 \\ 5.98247609 & 6.97222301 & 7.96199706 \end{pmatrix} \quad v_1 = \begin{pmatrix} 1.2857 & 2.21501324 & 3.14443571 \\ 4.07395305 & 5.00354286 & 5.93318801 \\ 6.86287609 & 7.79259801 & 8.72234706 \end{pmatrix}$$

, while after 5 iterations are:

$$u_5 = \begin{pmatrix} 0.22108069 & 1.1686221 & 2.11711012 \\ 3.0661668 & 4.01550672 & 4.96509061 \\ 5.91485142 & 6.86476954 & 7.81479791 \end{pmatrix} \quad v_5 = \begin{pmatrix} 2.24356641 & 2.9363427 & 3.62979046 \\ 4.32391978 & 5.01802638 & 5.7123312 \\ 6.40676441 & 7.10139079 & 7.79610334 \end{pmatrix}$$

- Test the evolution of the system, for example, by checking that the state has always the correct size, or that if you remove the reaction terms the maximum value is non-increasing and the sum is constant.

Hint: the function `allclose` of the library `numpy` might be useful to check if the sum is constant.

- Devise additional tests to reach as high line coverage as possible. You can also improve your code to deal with violations of the requirements of the input (e.g., raise an exception if the patterns are non-binary or if they are not unique)