

Going from notebooks to production code

Catherine Nelson and Robert Masson

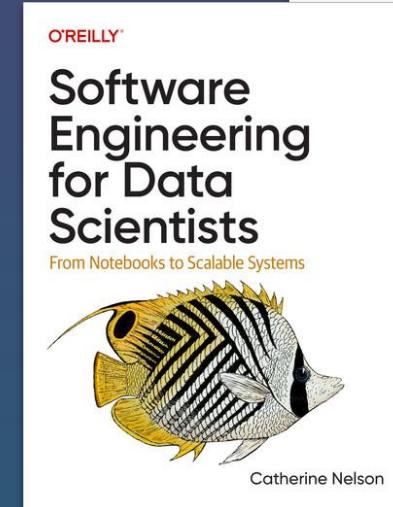
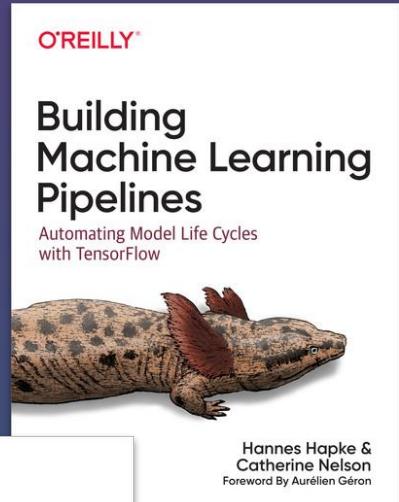
About Catherine

~10 years in data science with a strong focus on ML

Author of two O'Reilly books

Transitioning to startup AI Engineer role

Previously Principal Data Scientist at SAP Concur



<https://github.com/catherinenelson1/notebooks-to-production>

About Robert

12 years in data science with a focus on product analytics

10 years at FB/Meta, now at Atlassian working on Search, Chat and third party integrations

Agenda for today

- We'll give an intro presentation
- *Your turn: download the repo and get set up*
- We'll give a short presentation on modular code
- *Your turn: convert part of a notebook into modular functions*
- We'll give a short presentation on unit tests
- *Your turn: write tests for your new functions*
- We'll wrap up with some final thoughts

Introduction and setup

<https://github.com/catherinenelson1/notebooks-to-production>

Differences between notebooks and production code

- Notebooks are fantastic for exploration and experimentation
- Notebooks mix code, outputs and documentation together
- Production code is modular and reusable - forces you to break things into logical pieces
- This makes it easier to test, and guarantees that your results are what you expect
- Know when to graduate your code from exploration to something reliable and maintainable

Steps to go from notebooks to production

- Separate code out into different tasks
- Organize the code into functions
- Write the tests so that you know your code does what you expect it to do
- Make it shareable by sharing the modules it depends on (the requirements)
- Use CI/CD tools to deploy it

Our example notebook - let's classify penguins!



Penguins can be hard to tell apart

Given data on flipper size, bill length, bill depth, body mass, build a model to predict penguin species

Features: flipper size, bill length, body mass

Labels: 3 different species

Put the model in production

Example notebook [DEMO]

<https://github.com/catherinenelson1/notebooks-to-production>

Your turn:

Clone the repo

Install the requirements in a virtual environment

Ensure that you can run the notebook to the end

Use a virtual environment

```
python -m venv notebooks-to-prod

# Mac or Linux
source notebooks-to-prod/bin/activate

# Windows
notebooks-to-prod\Scripts\activate.bat

pip install -r requirements.txt
```

What you should get

Make a prediction on new data

```
[30]: scaled_data = scaler.transform([[40, 17, 190, 3500]])  
  
[31]: encoder.inverse_transform(lr.predict(scaled_data))[0]  
[31]: 'Adelie'  
  
[32]: encoder.inverse_transform(rf.predict(scaled_data))[0]  
[32]: 'Adelie'  
  
[33]: lr.predict_proba(X_test[:4])  
  
[33]: array([[2.43316874e-03, 1.08422977e-02, 9.86724534e-01],  
           [9.99192006e-01, 7.71119169e-04, 3.68744421e-05],  
           [9.95382486e-01, 4.57623344e-03, 4.12810025e-05],  
           [8.75121168e-01, 1.19155948e-01, 5.72288365e-03]])  
  
[34]: rf.predict_proba(X_test[:4])  
  
[34]: array([[0. , 0. , 1. ],  
           [0.99, 0.01, 0. ],  
           [1. , 0. , 0. ],  
           [0.53, 0.47, 0. ]])
```

Modular code

<https://github.com/catherinenelson1/notebooks-to-production>

Principles of modular code

Modular code means your code is broken into small independent parts – it's the opposite of having all the code for your project in one giant script

As far as possible, make sure that each function or class has one well-defined purpose – one "thing" that it does

You should be able to describe it in a short sentence, like "this function creates a data visualization"

If you start using the word "and" in your description, this may be a clue to break it down into multiple functions

"This function cleans the data and creates a data visualization" should be divided into one function that cleans the data and another that creates the visualization

Identify the steps in your notebook

What are the steps you need to keep?

Write a function stub for each step

For each function, what are the inputs and outputs?

Add in the code

Think about how it could go wrong and write a test

Instructions for exercise 1

- Look at the notebook and try to identify the different tasks
- In the exercises folder, simple_script.py copies (most of) the code from the notebook into a script; *run it and see its output*
- In the exercises folder, choose which file you're going to work on
 - Choose modular_code_exercise if you want to implement 2 functions
 - Choose modular_code_function_stubs.py if you want to implement all the functions yourself
- Implement the functions based on the code in the notebook or in simple_script.py
- Run your code and check it gives you the same answer as simple_script.py

Steps in the notebook



Load penguin
data

Clean missing
values

Scale/encode
features and
labels

Train model

Predict on new
penguins

From steps to function stubs



```
def download_data():
    pass

def clean_data():
    pass

def preprocess_data():
    pass

def train_model():
    pass

def predict_new_data():
    pass
```

Solution for Exercise 1

[DEMO]

<https://github.com/catherinenelson1/notebooks-to-production>

Unit tests

<https://github.com/catherinenelson1/notebooks-to-production>

Why write tests?

A test is code that calls a function and checks that it does what it's supposed to do – it's a safety net for your code

Tests give you evidence and confidence that your code works correctly by verifying expected behavior

For tiny one-off experiments, tests may not be essential

But you should use them if your code is part of a larger system, other people are changing your code, or other code depends on yours to return a certain result

Testing gives you guarantees that your code is working, and you'll know if a change that you or someone else makes breaks it

What goes in a test

1. **Arrange:** Set up everything you'll need to run the function, for example, load some data.
2. **Act:** Run the function you're testing.
3. **Assert:** Check that the result of running the function is what you expect.
4. **Cleanup:** Make sure the test doesn't leave any trace behind. For example, if you have opened a file, make sure to close it.

An example



```
def weighted_mean(num_list, weights):
    running_total = 0
    for i in range(len(num_list)):
        running_total += (num_list[i] * weights[i])
    return (running_total/sum(weights))
```

An example

```
def test_weighted_mean():

    list_a = [1, 2, 4]
    list_b = [1, 2, 4]

    result = weighted_mean(list_a, list_b)

    assert result == 3
```

An example



```
def test_weighted_mean():

    result = weighted_mean([1, 2, 4], [1, 2, 4])
    assert result == 3

    empty_list_result = weighted_mean([], [])
    assert not empty_list_result

    wrong_types_result = weighted_mean(['one', 2, 4], [1, 2, 4])
    assert not wrong_types_result
```

Use a testing framework

- A testing framework handles finding and running your tests automatically
- Without it, you'd need to run every test function manually or write boilerplate code to run them
- This doesn't scale
- We'll use Pytest, a popular Python testing framework

Instructions for exercise 2

- In the exercises folder, go to test_modular_code.py
- Complete the tests for clean_data and preprocess_data
- Use the Arrange/Act/Assert/Cleanup structure
- Test for some things you think might go wrong - poorly formatted data, missing data...
- Run the tests using `pytest test_modular_code.py`

Solution for Exercise 2

[DEMO]

<https://github.com/catherinenelson1/notebooks-to-production>

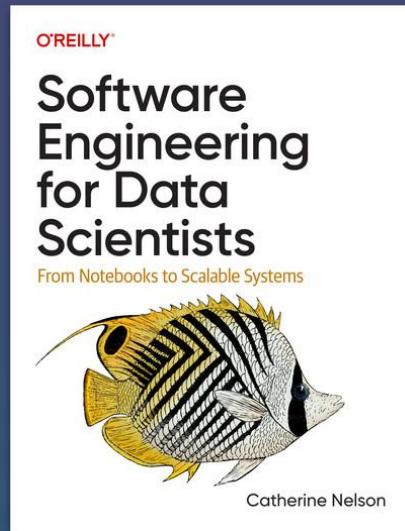
Other topics we didn't cover

- Deployment
- Automation (CI/CD)
- Efficiency
- Logging
- Error handling

But what about AI? [DEMO]

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Further reading



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