

# How can we teaching coding to scientists

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# Plan of the talk

- ▶ why should we teach scientists to code?
- ▶ what should we teach them?
- ▶ how should we teach them?
- ▶ how can I help?

# Why should we teach scientists to code

- ▶ Algorithms are at the heart of science. All scientists need to use and understand algorithms;
- ▶ Many branches of science use extensive computation.

# Is it reasonable to teach scientists about algorithms?

- ▶ understand what algorithms are, how they are implemented as programs on digital devices, and that programs execute by following precise and unambiguous instructions;
- ▶ create and debug simple programs;
- ▶ use logical reasoning to predict the behaviour of simple programs
- ▶ use technology purposefully to create, organise, store, manipulate and retrieve digital content

Does anyone recognize this?

# Understanding algorithms

- ▶ understand what algorithms are, how they are implemented as programs on digital devices, and that programs execute by following precise and unambiguous instructions;
- ▶ create and debug simple programs;
- ▶ use logical reasoning to predict the behaviour of simple programs
- ▶ use technology purposefully to create, organise, store, manipulate and retrieve digital content

National curriculum key stage 1: ages 5 to 7.

# Algorithms for data analysis

It's very common to have the following situation:

- ▶ we have some somewhat messy data;
- ▶ we have some hypothesis about these data we would like to test;
- ▶ the hypothesis can often be formulated as a model.

The combination is - data science, and statistics.

# Teaching data science, and all of statistics

Loading data, reviewing data, plotting data:

- ▶ Course page at <https://matthew-brett.github.io/les-pilot>;
- ▶ Our notebooks at <https://notebooks.azure.com/matthewbrett/libraries/dev-ac-uk>

With thanks and a recommendation for:

- ▶ Statistics without the agonizing pain:
- ▶ <https://www.youtube.com/watch?v=5Dnw46eC-0o>

## Setup for the demo - running on Azure

- ▶ Go to <https://notebooks.azure.com/matthewbrett/libraries/dev-ac-uk>;
- ▶ Clone that site (you'll need a free Microsoft login), and click on the `getting_started` notebook.



## Setup of the demo - running on your laptop

You can also run this on your own machine if you have Python and pip set up:

```
pip install pandas  
pip install jupyter[notebook]
```

Download <https://github.com/matthew-brett/dev.ac.uk> (via 'Clone or Download', 'Download zip').

```
cd directory_containing_download  
unzip dev.ac.uk-master.zip  
cd dev.ac.uk-master  
jupyter notebook
```

# Code for everyone

We have seen code as a way of:

- ▶ expressing operations on data;
- ▶ doing (and explaining) statistical inference.

Superficially, this is a course for scientists. But all academic fields are now analyzing data, as are the media. So, how should we teach?

# Berkeley Fundamentals of Data Science course

- ▶ All subjects (arts, sciences ...).
- ▶ Written as interactive Jupyter Notebooks.
- ▶ Textbook at <https://www.inferentialthinking.com>

## Back to the demo

- ▶ <https://notebooks.azure.com/matthewbrett/libraries/dev-ac-uk>
- ▶ `Plotting the classics.ipynb`.
- ▶ `Literary characters.ipynb`.

See the README for the site for the license of these files.

The Berkeley course is substantial

See :

<https://www.inferentialthinking.com/chapters/16/classification.html>

## Beyond the basics

- ▶ Many academic fields need a lot of computation.
- ▶ Computation is often done in a very sloppy and disorganized way.
- ▶ It's easy to make errors.
- ▶ People usually don't share their analyses.

## Easy to make errors in computation

*In my own experience, error is ubiquitous in scientific computing, and one needs to work very diligently and energetically to eliminate it. One needs a very clear idea of what has been done in order to know where to look for likely sources of error. I often cannot really be sure what a student or colleague has done from his/her own presentation, and in fact often his/her description does not agree with my own understanding of what has been done, once I look carefully at the scripts.*

David L. Donoho (2010) "An invitation to reproducible computational research" *Biostatistics* 11(3) 385

## Sharing analyses

*An article about computational science in a scientific publication is **not** the scholarship itself, it is merely **advertising** of the scholarship. The actual scholarship is the complete software development environment and the complete set of instructions which generated the figures.*

Jonathan B. Buckheit and David L. Donoho (1995) "Wavelab and reproducible research"

[http://statweb.stanford.edu/~wavelab/Wavelab\\_850/wavelab.pdf](http://statweb.stanford.edu/~wavelab/Wavelab_850/wavelab.pdf)



What tools do we need to teach for this endeavor?

# What tools do we need to teach for this endeavor?

- ▶ Unix command line;
- ▶ Text editor for everything (including presentations);
- ▶ Git for version control;
- ▶ A scientific programming language such as Python or R;
- ▶ Unit testing, code coverage, continuous integration;
- ▶ make / Makefiles;

# Can it be taught?

- ▶ <http://www.jarrodmillman.com/rcsds>
- ▶ <https://github.com/berkeley-stat159>

# How can I help?

- ▶ It's a culture; spread the culture;
- ▶ Organize meetups, interest groups, talks;
- ▶ Find like-minded lecturers to support you, and for you to support.
- ▶ Consider forming or joining a Hacker Within group:  
<http://www.thehackerwithin.org>.

Is this the end?

Yes, it's the end of the talk.