

PHYS4038/MLiS and ASI/MPAGS

Scientific Programming in



mpags-python.github.io

Steven Bamford



**University of
Nottingham**
UK | CHINA | MALAYSIA

An introduction to scientific programming with



Session 3:
Staying organised

Session 3

In this session:

- Organising your python installation
- Version control
- GitHub tools and workflow
- How to submit coursework

Managing your environment

- **Some good things about Python**
 - lots of modules from many sources
 - ongoing development of Python and modules
- **Some bad things about Python**
 - lots of modules from many sources
 - ongoing development of Python and modules
- **A solution**
 - Maintain (or have option to create) separate environments (or manifests) for different projects

Managing your environment

- **Desirable**

- long term stability of your programs
- help others easily install same dependencies
- benefit from latest features and bugfixes

- **Solution**

- maintain separate environments for different projects
 - Anaconda: `conda`
 - native Python: `pip` and `virtualenv`

Managing your environment

- **conda** – <http://conda.pydata.org>
 - specific to the Anaconda Python distribution
 - install modules
 - automatically manage dependencies and compatibility
 - similar to 'pip', but can install binaries and not just for python
 - can use pip within a conda environment (but try conda first)
 - create and switch between environments
 - specific collections of compatible modules and executables
- Windows: use Anaconda Prompt
- Linux/Mac: use any terminal

Managing your environment

- conda basic usage

```
$ conda create -n python_course # -n <name> or -p <path>
$ conda activate python_course # <name> or <path>
$ conda install scipy matplotlib
$ ipython # use the environment
$ conda deactivate
```

Managing your environment

- Saving your environment (to use on another machine or distribute)

```
$ conda env export -n python_course > environment.yml  
$ conda create -n new_env -f environment.yml
```

- environment.yml contains all dependencies and versions
- maybe neater to manually maintain your own environment.yml

```
name: myenv  
dependencies:  
  - python  
  - numpy  
  - matplotlib
```

- to make your environment match an environment.yml file:

```
$ conda env update -n myenv -f myenv.yml --prune
```


Managing your environment

- **virtualenv**

- general Python solution – <http://virtualenv.pypa.io>
- modules are installed with pip – <https://pip.pypa.io>

```
$ pip install virtualenv      # install virtualenv
$ virtualenv ENV1             # create a new environment ENV1
$ source ENV/bin/activate    # set PATH to our environment
(ENV1)$ pip install emcee    # install modules into ENV1
(ENV1)$ pip install numpy==1.8.2 # install specific version
(ENV1)$ python               # use our custom environment
(ENV1)$ deactivate           # return our PATH to normal
```

Managing your environment

- **virtualenv**
 - can record current state of modules to a 'requirements' file

```
(ENV1)$ pip freeze > requirements.txt
$ cat requirements.txt
emcee==2.1.0
numpy==1.8.2
$ deactivate
$ virtualenv ENV2
$ sourceENV2/bin/activate
(ENV2)$ pip install -r requirements.txt
```

Managing your environment

- **Updating packages**

```
$ conda update --all
```

```
$ conda update scipy emcee
```

OR

```
$ pip install --upgrade
```

```
$ pip install --upgrade scipy emcee
```

Jupyter kernel discovery

- Can install and run Jupyter notebook in an environment, but better to run from base environment and then select kernel within notebook
- Jupyter can autodiscover conda environments
- Just need to install nb_conda_kernels in notebook environment

```
$ conda install -n base nb_conda_kernels
```

- and ipykernel in any environments you want to use in notebook

```
$ conda install -n myenv ipykernel
```

Version control

- Keep a secure backup of your work
- Maintain a record of significant changes
- Undo mistakes
- Undo undone mistakes that turned out to not be mistakes
- Log the reasons why you made particular changes
- Separate your work on different features
- Collaborate more easily



- Distributed version control
 - everyone has a full copy of history

GitHub



- Where many projects keep and share code
 - particularly open-source projects
- Unlimited private repos for education and research:
 - <https://education.github.com>

Similar alternative:



Getting started with version control

- Create a GitHub account
- Join assignment to create a new repository
<https://classroom.github.com/a/bsgUSS2H>
- Create README in the browser
- Brief intro to Markdown
<https://guides.github.com/features/mastering-markdown/>
- Installing git (with conda)

```
$ conda install git
```

Getting started with version control

- Clone your repo locally

```
$ git clone <link_to_your_repo>
```

- Edit README.md locally, then check status and diff

```
$ git status
```

```
$ git diff # show changes
```

- Add files to commit, perform commit and push commit to GitHub

```
$ git add README.md
```

```
$ git commit -m"Edited the readme"
```

```
$ git push
```

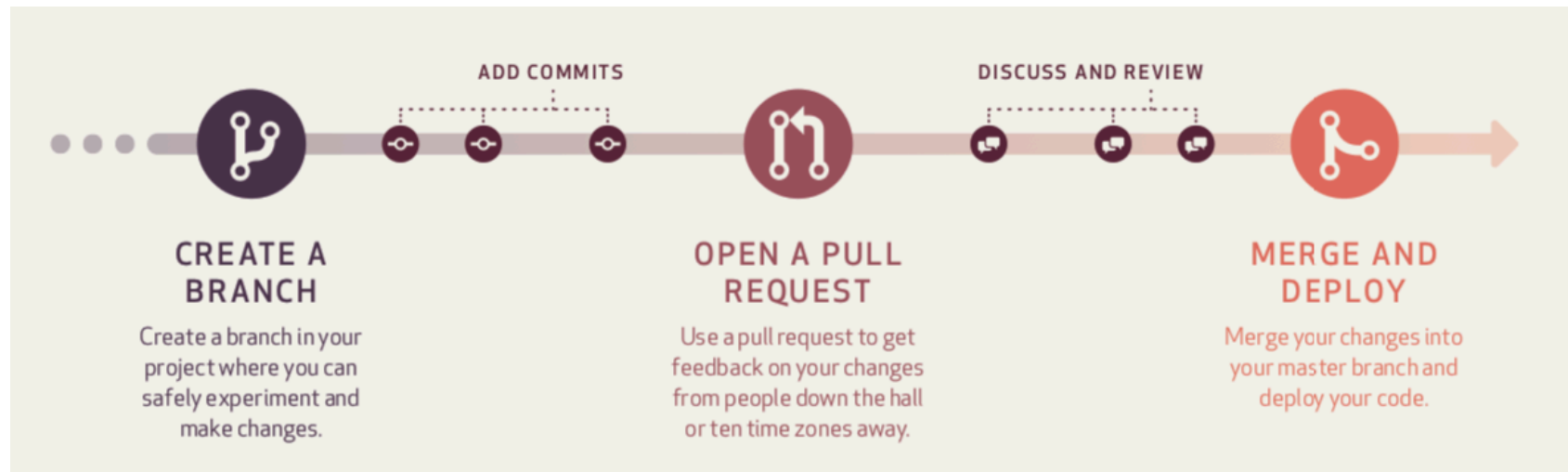
- If files changed on GitHub, fetch and merge the changes

```
$ git pull
```

<https://guides.github.com/introduction/git-handbook/>

Good practice and GitHub extras

- Using branches and tags
- Issues
- Pull requests



For more information:

- <https://guides.github.com>
- <https://www.atlassian.com/git/tutorials>
- <https://lab.github.com>

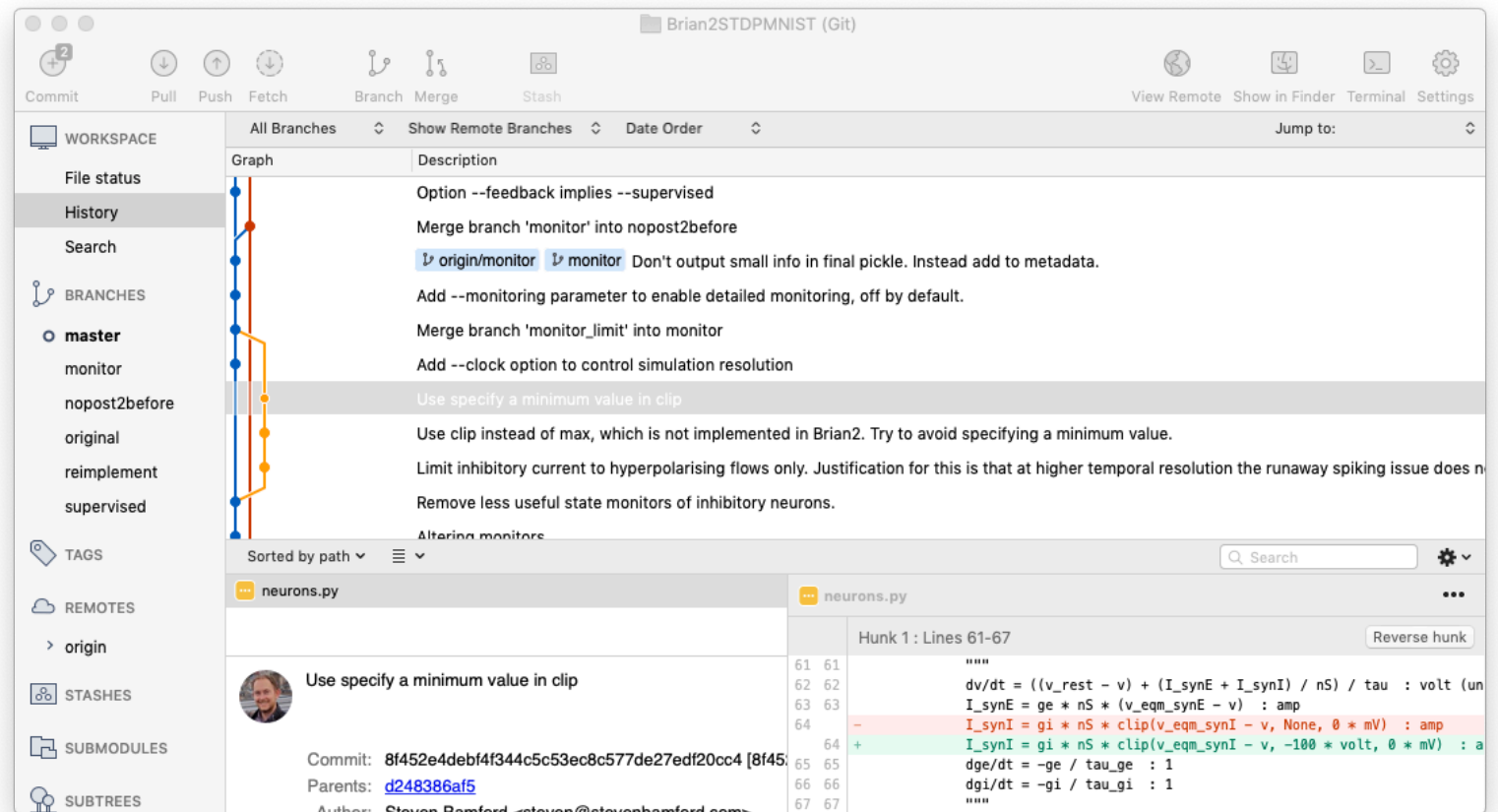
Git GUIs



GUI for Windows & Mac



GUI for Windows, Linux, Mac



Assessment

For those taking this module for MPAGS credits

- Assessed by development of a Python program relevant to your interests
 - put course material into practice
 - opportunity to become familiar with Python
 - get feedback on your coding
- Your code should...
 - be written as an executable module (.py file) or Jupyter notebook (.ipynb)
 - do something meaningful: analyse real data or perform a simulation
 - define at least two user functions (but typically more)
 - make use of appropriate specialist modules
 - produce at least one informative plot
 - comprise $>\sim 50$ lines of actual code
 - excluding comments, imports and other 'boilerplate'
 - contain no more than 1000 lines in total
 - if you have written more, please isolate an individual element

Code development

- Three stages (first two optional for MPAGS students)
 1. hand-in by **28th October**
 - README describing what you intend your code to do
 - Rough outline of the code (classes, functions, snippets, comments, pseudocode)
 2. hand-in by **18th November**
 - Rough version of your code, may be incomplete, have bugs, although try to make it reasonable and easy to understand!
 3. hand-in by **16th December**
 - Complete working version of your code

Deadlines are 3pm on Wednesdays.

Coursework submission

- Submission and feedback via your GitHub repository
- Mandatory for MLiS, optional for MPAGS
- **Create a branch called sub1**
- Should contain a README file including:
 - your full name and university
 - possibly some background (basic explanation, references, ...)
 - an overview of the intended functionality of your program
 - ideas of the modules you plan to use
 - ideas of the structure of your code (functions, etc.)
 - possibly snippets or pseudocode
 - any remaining uncertainties or questions

The screenshot shows the GitHub interface for the repository 'mpags-python / coursework-bamford', which is marked as 'Private'. The 'Code' tab is selected, and the repository is noted as 'created by GitHub Classroom'. It shows '1 commit' and '1 branch'. A modal window titled 'Switch branches/tags' is open, with 'sub1' entered in the search field. Below the search field, the 'Create branch: sub1 from 'master'' option is highlighted. The repository's README file is partially visible, showing the title 'Steven's cool project' and the author 'Steven Bamford'.

Questions and exercises

Any questions?

- ask on the Slack channel (@Steven Bamford)
- email steven.bamford@nottingham.ac.uk
- ask in the next synchronous session

Exercises

Practice using conda and git