

PHYS4038/MLiS and ASI/MPAGS

Scientific Programming in



mpags-python.github.io

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An introduction to scientific programming with



Session 3:
Staying organised

Session 3

In this session:

- Solutions to exercises 1 & 2
 - Organising your python installation
 - Version control
 - GitHub tools and workflow
 - How to submit coursework
-
- Help with conda, git, etc.
 - Advice on coursework

Questions

- Talk to me:
 - During teaching sessions (*preferred*)
 - Specific questions, clarifications – just ask
 - Bigger issues – wait until end of lecture / start of examples class
 - Remote students connect via skype group:
 - <https://join.skype.com/KpW5oCLNNiJt>
 - text during lecture, video during examples class
 - Via email: steven.bamford@nottingham.ac.uk
 - Arrange a meeting
 - email me
 - office: CAPT A112b
 - or in skype group

Exercise solutions

<https://github.com/mpags-python/exercises>

<https://nbviewer.jupyter.org/github/mpags-python/exercises/blob/master/Exercises2.ipynb>

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

- **Some extra details for using on UoN computers**

- Anaconda is installed on hard drive (C:) of each machine
- User can write to C:\Anaconda3 – but will not be available from other machines, and probably wiped periodically
- User folder is Z: – available from all machines (but 4Gb limit)
- Either recreate environment in C: each time, or create in Z:

- specify environment by directory (-p option):

```
$ conda create -p Z:\envs\python_course
```

- set as default (every time):

```
$ conda config --prepend envs_dirs Z:\envs
```

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:

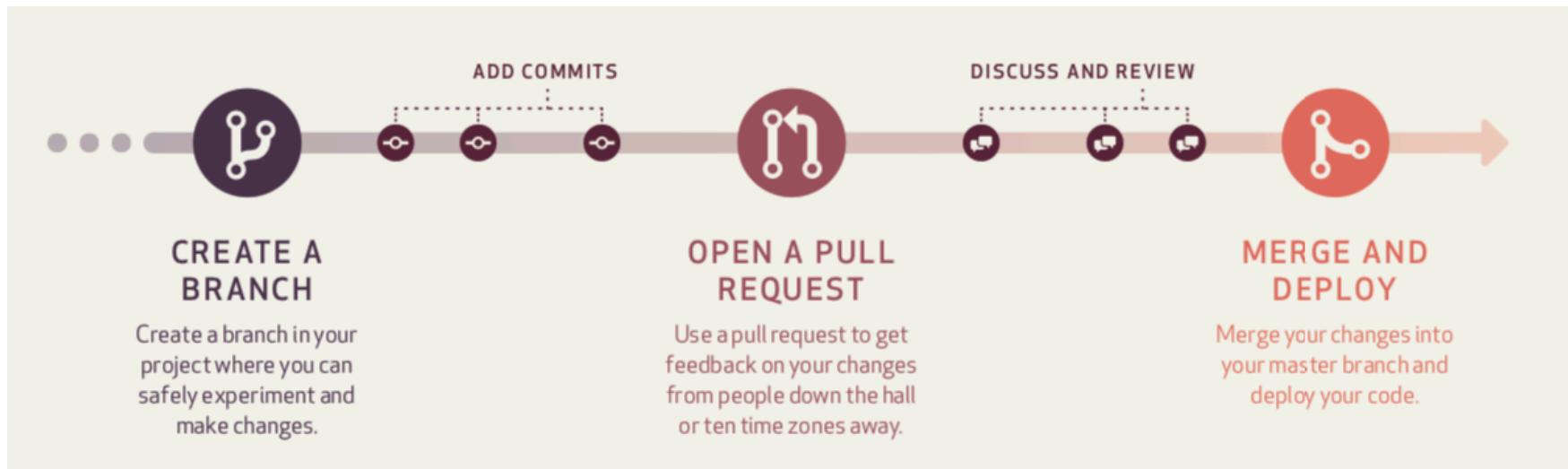


Getting started with version control

- Create a GitHub account
- Join assignment to create a new repository
<https://classroom.github.com/a/K4tUSkL0>
- Create README in the browser
- Brief intro to Markdown
<https://guides.github.com/features/mastering-markdown/>
- Installing git (with conda)
- Cloning your repo locally
- Editing locally, status and diff
- Add, commit and push
- 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>

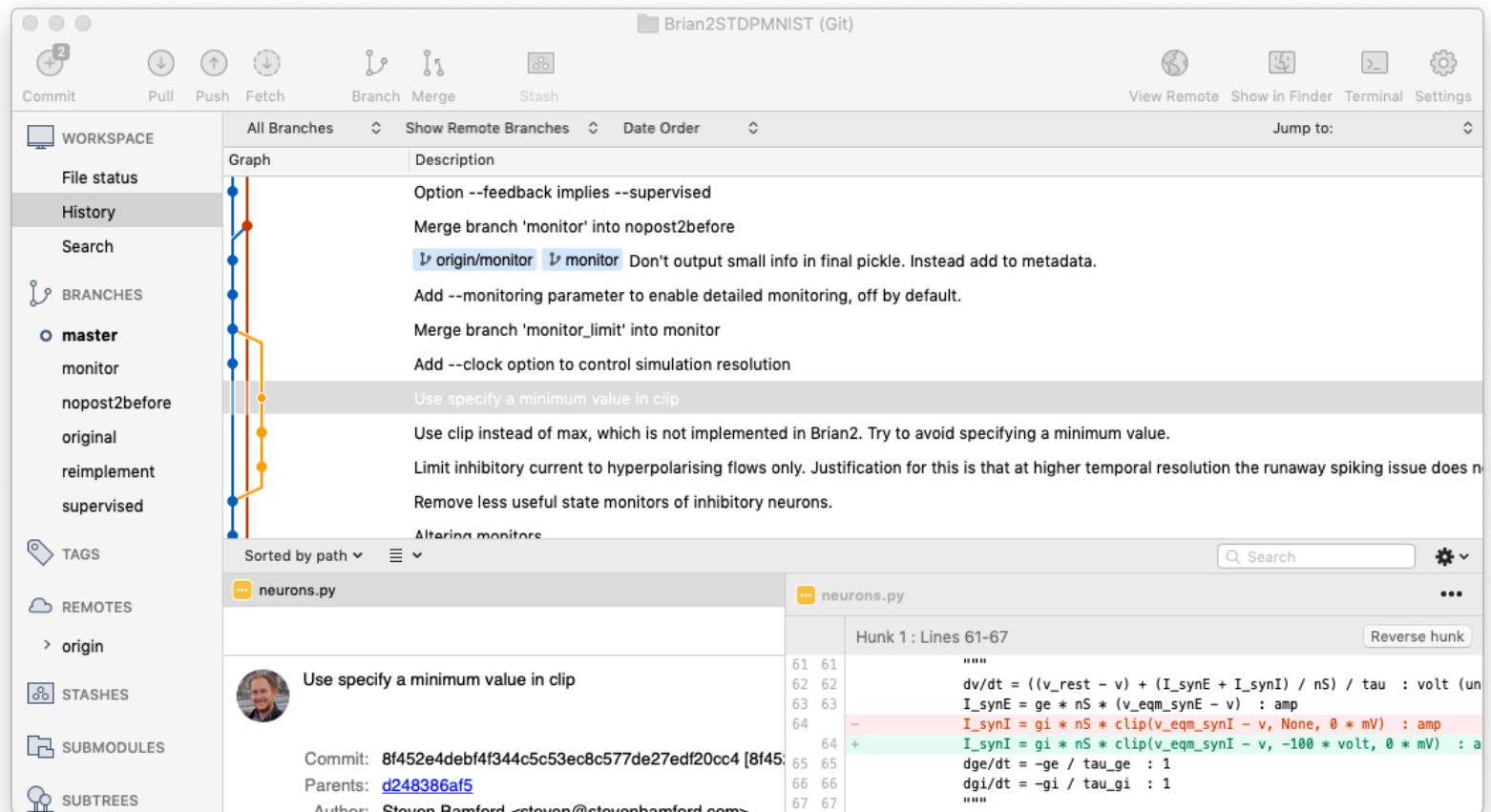
Git GUIs



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 **1st November** (optional for feedback)
 - README describing what you intend your code to do
 - Rough outline of the code (classes, functions, snippets, comments, pseudocode)
 2. hand-in by **15th November** (optional for feedback)
 - Rough version of your code, may be incomplete, have bugs, although try to make it reasonable and easy to understand!
 3. hand-in by **13th December** (required for MPAGS credits)
 - Complete working version of your code

Deadlines are 3pm on Fridays.

Questions and exercises

Now...

Work on your coursework README

Practise with conda and git

Any questions?

- shout and wave
- skype (spbamford)
 - <https://join.skype.com/KpW5oCLNNijt>
- email steven.bamford@nottingham.ac.uk