

# The SciPy Stack

# Data Analytics in Python

# The SciPy Stack

SciPy is a Python-based ecosystem of libraries and tools for scientific computing and data analytics

- ▶ iPython
- ▶ Jupyter notebooks
- ▶ Numpy
- ▶ Pandas
- ▶ Matplotlib

iPython is the primary way of interacting with the SciPy stack – whether through the shell or a Jupyter notebook – so we'll learn iPython first

# iPython

Two modes:

- ▶ Interactive shell
  - ▶ Replacement for `python` REPL
- ▶ Jupyter notebook
  - ▶ Interactive web-based documents mixing text, executable code, graphics

Before we proceed, make sure your computer is ready (OS shell):

```
$ conda update conda
$ conda update python ipython jupyter numpy pandas matplotlib
```

# iPython Shell History

```
In [1]: ['Sage', 'Thyme', 'Oragano', 'Posh']
Out[1]: ['Sage', 'Thyme', 'Oragano', 'Posh']

In [2]: type(In[1])
Out[2]: str

In [3]: type(Out[1])
Out[3]: list

In [4]: spices = Out[1]

In [5]: spices
Out[5]: ['Sage', 'Thyme', 'Oragano', 'Posh']

In [6]: spices is Out[1]
Out[6]: True
```

In is a list, Out is a dict.

Single ? gives abbreviated version of python's help

```
In [7]: def add(a, b):
...:     """Return the result of + operation on a and b"""
...:     return a + b
...:
In [8]: add?
Signature: add(a, b)
Docstring: Return the result of + operation on a and b
File:      ~/cs2316/<ipython-input-7-af5293282e78>
Type:      function
```

Double ?? gives source code, if available.

```
In [9]: add??
Signature: add(a, b)
Source:
def add(a, b):
    """Return the result of + operation on a and b"""
    return a + b
File:      ~/cs2316/<ipython-input-7-af5293282e78>
Type:      function
```

# iPython Magic Commands

Special commands provided by iPython, prepended by %.

- ▶ Run a Python script from within iPython:

```
In [35]: %run people.py
[<Stan, 2008-08-13, 150cm, 45kg>,
 <Kyle, 2008-02-25, 160cm, 50kg>,
 <Cartman, 2008-05-26, 140cm, 100kg>,
 <Kenny, 2009-07-30, 130cm, 40kg>]
```

- ▶ Get help with a magic command with ?

```
In [2]: %cd?
Docstring:
Change the current working directory.

This command automatically maintains an internal list of directories
you visit during your IPython session, in the variable _dh. The
command %dhist shows this history nicely formatted. You can also
do 'cd -<tab>' to see directory history conveniently.

Usage:

    cd 'dir': changes to directory 'dir'.
(additional output elided)
```

Get a list of all magic commands with %lsmagic

# iPython Shell Commands

Run shell commands by prepending with a !

```
In [27]: !ls *.py
fun.py    grades.py  maths.py  people.py  pp.py

In [28]: pyscripts = !ls *.py

In [29]: pyscripts
Out[29]: ['fun.py', 'grades.py', 'maths.py', 'people.py', 'pp.py']
```

iPython provides magic commands for most common shell commands.

# iPython Direcotry Bookmarking

Great timesaving feature: bookmark directories

```
In [3]: %pwd
Out[3]: '/home/chris/vcs/github.com/cs2316/cs2316.github.io/code'

In [4]: %cd
/home/chris

In [5]: %bookmark cs2316code ~chris/vcs/github.com/cs2316/cs2316.github.io/code

In [6]: cd cs2316code
(bookmark:cs2316code) -> ~chris/vcs/github.com/cs2316/cs2316.github.io/code
/home/chris/vcs/github.com/cs2316/cs2316.github.io/code
```



# iPython Automagic commands

With `automagic` turned on, some shell commands can be run as if they were built into iPython:

```
In [22]: pwd
Out[22]: '/Users/chris/cs2316'

In [23]: ls *.py
fun.py    grades.py maths.py  people.py pp.py
```

- ▶ Toggle automagic on and off with `%automagic`.
- ▶ These commands work with automagic:
  - ▶ `%cd`, `%cat`, `%cp`, `%env`, `%ls`, `%man`, `%mkdir`, `%more`, `%mv`, `%pwd`, `%rm`, and `%rmdir`

# Timing Code in iPython

```
In [23]: import numpy as np

In [24]: pylist = list(range(1, 100000))

In [25]: nparray = np.arange(1, 1000000)

In [35]: %timeit _ = [x * 2 for x in pylist]
100 loops, best of 3: 7.89 ms per loop

In [37]: %timeit _ = nparray.copy() * 2
100 loops, best of 3: 3.76 ms per loop
```

Notice that I copied the Numpy array before applying the  $\ast 2$  operation to make the comparison to the Python list comprehension fair. You'll learn why when we discuss Numpy in the next lecture.

# Profiling a Script

```
In [7]: %run -p -l 10 -s cumulative funccalc.py
        2673375 function calls (1147466 primitive calls) in 1.691 seconds

Ordered by: cumulative time
List reduced from 56 to 10 due to restriction <10>

ncalls  tottime  percall  cumtime  percall  filename:lineno(function)
    2/1    0.000    0.000    1.691    1.691 {built-in method builtins.exec}
    1    0.000    0.000    1.691    1.691 <string>:1(<module>)
    1    0.000    0.000    1.691    1.691 interactiveshell.py:2431(safe_execfile)
    1    0.000    0.000    1.691    1.691 py3compat.py:182(execfile)
    1    0.000    0.000    1.690    1.690 funccalc.py:1(<module>)
    1    0.000    0.000    1.689    1.689 funccalc.py:46(main)
    1    0.039    0.039    1.689    1.689 funccalc.py:34(profile)
510961/10000  0.510    0.000    0.603    0.000 funccalc.py:14(sub)
510961/10000  0.514    0.000    0.598    0.000 funccalc.py:6(add)
510961/10000  0.340    0.000    0.340    0.000 funccalc.py:22(mult)
```

- ▶ -p means profile
- ▶ -l 10 means show only 10 lines
- ▶ -s cumulative means sort by cumulative time

# Profiling a Function

`%prun` profiles a function. Uses same options as `% run -p`.

```
In [10]: %prun -l 10 -s cumulative funcalc.profile()
         2673429 function calls (1148052 primitive calls) in 1.726 seconds
```

```
Ordered by: cumulative time
```

```
List reduced from 15 to 10 due to restriction <10>
```

	ncalls	tottime	percall	cumtime	percall	filename:lineno(function)
	1	0.000	0.000	1.726	1.726	{built-in method builtins.exec}
	1	0.000	0.000	1.726	1.726	<string>:1(<module>)
	1	0.042	0.042	1.726	1.726	funcalc.py:34(profile)
511231/10000		0.537	0.000	0.620	0.000	funcalc.py:6(add)
511231/10000		0.523	0.000	0.615	0.000	funcalc.py:14(sub)
511231/10000		0.336	0.000	0.336	0.000	funcalc.py:22(mult)
20000		0.019	0.000	0.097	0.000	random.py:223(randint)
501231		0.092	0.000	0.092	0.000	funcalc.py:15(dec)
501231		0.082	0.000	0.082	0.000	funcalc.py:7(inc)
20000		0.036	0.000	0.078	0.000	random.py:179(randrange)

# Interactive Debugging in iPython

Enter a debug session with `%debug` ...

# A Taste of Data Analytics in iPython Shell

```
In [1]: cd analytics/  
/home/chris/vcs/github.com/cs2316/cs2316.github.io/code/analytics
```

```
In [3]: exam1grades = np.loadtxt('exam1grades.txt')
```

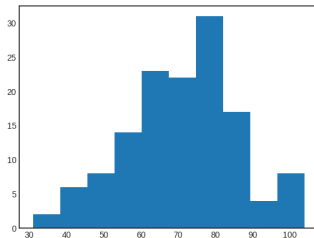
```
In [4]: import matplotlib.pyplot as plt
```

```
In [5]: %matplotlib qt5
```

```
In [6]: plt.hist(exam1grades)
```

```
Out[6]:
```

```
(array([ 2.,  6.,  8., 14., 23., 22., 31., 17.,  4.,  8.]),  
array([ 31. , 38.3, 45.6, 52.9, 60.2, 67.5, 74.8, 82.1,  
       89.4, 96.7, 104. ]),  
<a list of 10 Patch objects>)
```



# Jupyter Notebooks

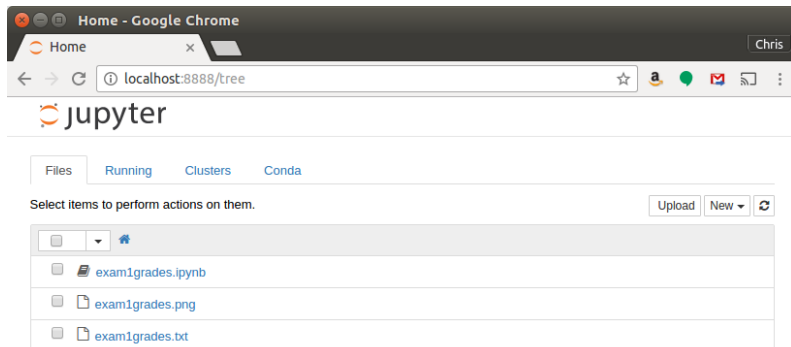
Go to the directory that holds your notebooks, or the class web site repo's code/analytics directory for this example and enter `jupyter notebook`.

```
[chris@bolshoi ~/vcs/github.com/cs2316/cs2316.github.io/code/analytics]
$ jupyter notebook
[I 15:06:15.705 NotebookApp] Serving notebooks from local directory:
    /home/chris/vcs/github.com/cs2316/cs2316.github.io/code/analytics
[I 15:06:15.705 NotebookApp] 0 active kernels
[I 15:06:15.705 NotebookApp] The Jupyter Notebook is running at:
    http://localhost:8888/
[I 15:06:15.705 NotebookApp] Use Control-C to stop this server and shut down all
    kernels (twice to skip confirmation).
Created new window in existing browser session.
```

Now a Jupyter Notebook server is running and you're ready to use iPython from the Jupyter Notebook web interface.

# Jupyter Web Interface

After running `jupyter notebook` from your OS command shell, open a browser and navigate to `localhost:8888`. You'll see a screen that looks like this:



Notice the listing of files in the directory in which you started the Jupyter notebook server.



# A Taste of Data Analytics in Jupyter Notebook

Select the `exam1grades.ipynb` file and you'll get this:

