11 OCTOBER 2016 / JUPYTER

# 28 Jupyter Notebook tips, tricks, and shortcuts

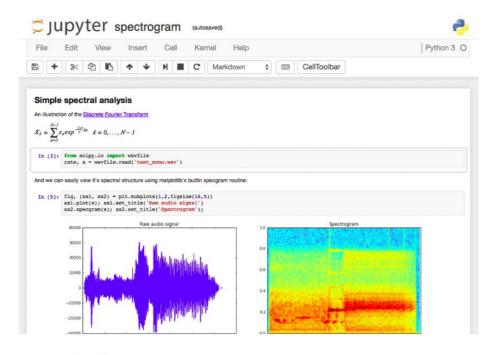
This post is based on a post that originally appeared on <u>Alex Rogozhnikov's</u> blog, 'Brilliantly Wrong'.

We have expanded the post and will continue to do so over time — if you have a suggestion please let us know. Thanks to Alex for graciously letting us republish his work here.

## **Jupyter Notebook**

Jupyter notebook, formerly known as the IPython notebook, is a flexible tool that helps you create readable analyses, as you can keep code, images, comments, formulae and plots together.

Jupyter is quite extensible, supports many programming languages and is easily hosted on your computer or on almost any server — you only need to have ssh or http access. Best of all, it's completely free.



The Jupyter interface.

Project Jupyter was born out of the IPython project as the project evolved to become a notebook that could support multiple languages - hence its historical name as the IPython notebook. The name Jupyter is an indirect acronyum of the three core languages it was designed for: JUlia, PYThon, and R and is inspired by the planet Jupiter.

When working with Python in Jupyter, the IPython kernel is used, which gives us some handy access to IPython features from within our Jupyter notebooks (more on that later!)

## 1. Keyboard Shortcuts

As any power user knows, keyboard shortcuts will save you lots of time. Jupyter stores a list of keybord shortcuts under the menu at the top: Help > Keyboard Shortcuts, or by pressing H in command mode (more on that later). It's worth checking this each time you update Jupyter, as more shortcuts are added all the time.

Another way to access keyboard shortcuts, and a handy way to learn them is to use the command palette: Cmd + Shift + P (or Ctrl + Shift + P on Linux and Windows). This dialog box helps you run any command by name - useful if you don't know the keyboard shortcut for an action or if what you want to do does not have a keyboard shortcut. The functionality is similar to Spotlight search on a Mac, and once you start using it you'll wonder how you lived without it!



The command palette.

Some of my favorites:

- Esc will take you into command mode where you can navigate around your notebook with arrow keys.
- While in command mode:
  - A to insert a new cell above the current cell, B to insert a new cell below.
  - M to change the current cell to Markdown, Y to change it back to code
  - D + D (press the key twice) to delete the current cell
- Enter will take you from command mode back into edit mode for the given cell.
- Shift + Tab will show you the Docstring (documentation) for the the object you
  have just typed in a code cell you can keep pressing this short cut to cycle through a
  few modes of documentation.
- Ctrl + Shift + will split the current cell into two from where your cursor is.
- Esc + F Find and replace on your code but not the outputs.
- Esc + 0 Toggle cell output.
- Select Multiple Cells:
  - Shift + J or Shift + Down selects the next sell in a downwards direction.
     You can also select sells in an upwards direction by using Shift + K or Shift + Up.
  - Once cells are selected, you can then delete / copy / cut / paste / run them as a batch. This is helpful when you need to move parts of a notebook.

You can also use Shift + M to merge multiple cells.

```
In [ ]: print("This is a first off")
In [ ]: print("This is a second cell")
In [ ]: print("This is a third cell")
```

Merging multiple cells.

## 2. Pretty Display of Variables

The first part of this is pretty widely known. By finishing a Jupyter cell with the name of a variable or unassigned output of a statement, Jupyter will display that variable without the need for a print statement. This is especially useful when dealing with Pandas DataFrames, as the output is neatly formatted into a table.

What is known less, is that you can alter a modify the <code>ast\_note\_interactivity</code> kernel option to make jupyter do this for any variable or statement on it's own line, so you can see the value of multiple statements at once.

```
from IPython.core.interactiveshell import InteractiveShell
InteractiveShell.ast_node_interactivity = "all"
```

```
from pydataset import data
quakes = data('quakes')
quakes.head()
quakes.tail()
```

	lat	long	depth	mag	stations
1	-20.42	181.62	562	4.8	41
2	-20.62	181.03	650	4.2	15
3	-26.00	184.10	42	5.4	43
4	-17.97	181.66	626	4.1	19
5	-20.42	181.96	649	4.0	11

	lat	long	depth	mag	stations
996	-25.93	179.54	470	4.4	22
997	-12.28	167.06	248	4.7	35
998	-20.13	184.20	244	4.5	34
999	-17.40	187.80	40	4.5	14
1000	-21.59	170.56	165	6.0	119

If you want to set this behaviour for all instances of Jupyter (Notebook and Console), simply create a file ~/.ipython/profile\_default/ipython\_config.py with the lines below.

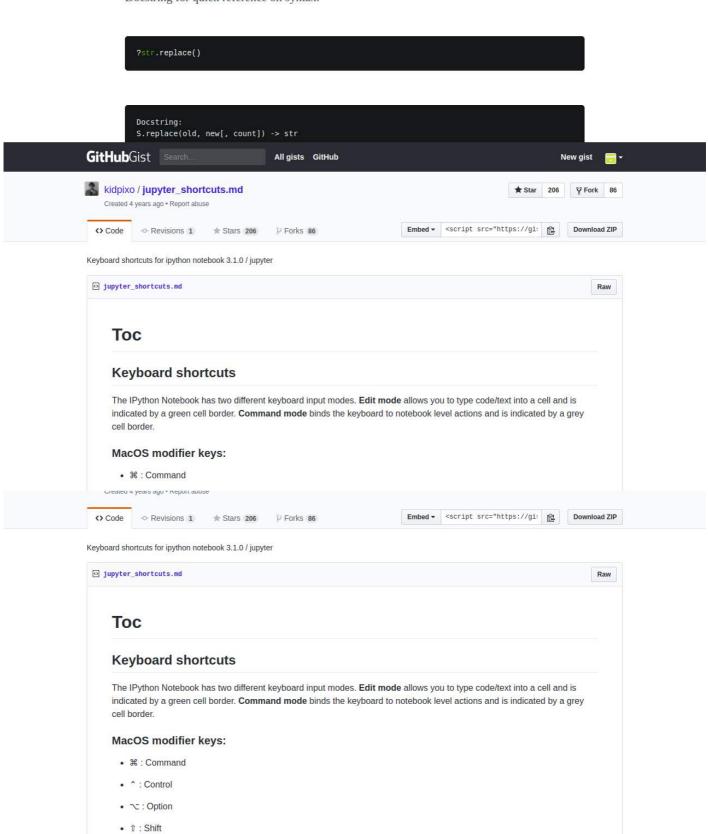
```
c = get_config()

# Run all nodes interactively
c.InteractiveShell.ast_node_interactivity = "all"
```

## 3. Easy links to documentation

Inside the Help menu you'll find handy links to the online documentation for common libraries including NumPy, Pandas, SciPy and Matplotlib.

Don't forget also that by prepending a library, method or variable with ?, you can access the Docstring for quick reference on syntax.



- ∼ : Option
- 1 : Shift
- ←: Return
- 📋 : Space
- → : Tab

#### Command Mode (press Esc to enable)

- ← : enter edit mode
- . û ←: run cell, select below
- ^ ←: run cell
- . ~ cell, insert below
- Y: to code
- M : to markdown
- R: to raw
- 1: to heading 1
- 2: to heading 2

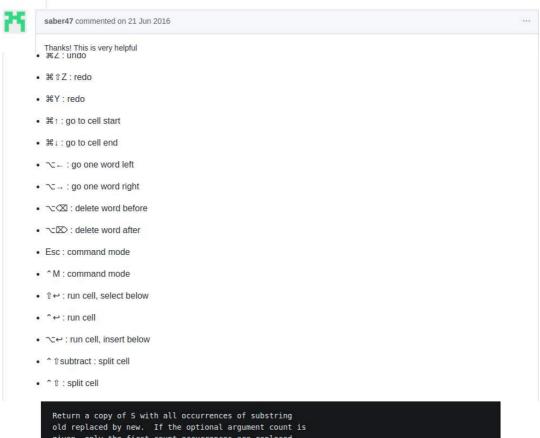
- Y: to code
- · M: to markdown
- R : to raw
- 1: to heading 1
- 2: to heading 2
- 3: to heading 3
- 4: to heading 4
- 5: to heading 5
- 6: to heading 6
- † : select cell above
- . K : select cell above
- 1 : select cell below
- . J : select cell below · A: insert cell above
- · B : insert cell below
- · Z : undo last cell deletion
- D,D : delete selected cell
- S : Save and Checkpoint
- . L : toggle line numbers
- O : toggle output
- 1 O: toggle output scrolling
- Esc : close pager
- Q : close pager
- H : show keyboard shortcut help dialog
- I,I: interrupt kernel
- 0,0 : restart kernel
- 🔒 : scroll down
- Î ⊔ : scroll up



```
· Esc : command mode

    ^M: command mode

. û ←: run cell, select below
· ^ ←: run cell
. ~ : run cell, insert below
• ^ f: split cell
• #S: Save and Checkpoint
• †: move cursor up or previous cell
• 1: move cursor down or next cell
• 1 : ignore
```



given, only the first count occurrences are replaced. method\_descriptor

## 4. Plotting in notebooks

There are many options for generating plots in your notebooks.

- matplotlib (the de-facto standard), activated with %matplotlib inline Here's a Dataquest Matplotlib Tutorial.
- %matplotlib notebook provides interactivity but can be a little slow, since rendering
- Seaborn is built over Matplotlib and makes building more attractive plots easier. Just by importing Seaborn, your matplotlib plots are made 'prettier' without any code
- · mpld3 provides alternative renderer (using d3) for matplotlib code. Quite nice, though incomplete.

%pycat pythoncode.py

```
import numpy
def append_if_not_exists(arr, x):
    if x not in arr:
        arr.append(x)

def some_useless_slow_function():
    arr = list()
    for i in range(10000):
        x = numpy.random.randint(0, 10000)
        append_if_not_exists(arr, x)
```

```
### 13. IPython Magic - %prun: Show how much time your program spent in each function.
Using `%prun statement_name` will give you an ordered table showing you the number of times each
```python
%prun some_useless_slow_function()
```

```
26324 function calls in 0.556 seconds
Ordered by: internal time
ncalls tottime percall cumtime percall filename:lineno(function)
             0.000 0.528 0.000 <ipython-input-46-b52343f1a2d5>:2(append_if_not_exists
10000
       0.527
                             0.000 {method 'randint' of 'mtrand.RandomState' objects}
 10000
        0.022
               0.000
                      0.022
                      0.556 0.556 <ipython-input-46-b52343f1a2d5>:6(some_useless_slow_fu
       0.006
               0.006
                      6320
       0.001
               0.000
       0.000
               0.000
       0.000
               0.000
        0.000
               0.000
```

#### 14. IPython Magic - Debugging with %pdb

Jupyter has own interface for <u>The Python Debugger</u> ( pdb ). This makes it possible to go inside the function and investigate what happens there.

You can view a list of accepted commands for pdb here.

```
%pdb

def pick_and_take():
    picked = numpy.random.randint(0, 1000)
    raise NotImplementedError()

pick_and_take()
```

```
Automatic pdb calling has been turned ON
```

```
NotImplementedError Traceback (most recent call last)
<ipython-input-24-0f6b26649b2e> in <module>()
5 raise NotImplementedError()
6
----> 7 pick and take()
```

```
<ipython-input-24-0f6b26649b2e> in pick_and_take()
    3 def pick_and_take():
    4    picked = numpy.random.randint(0, 1000)
----> 5    raise NotImplementedError()
    6
    7 pick_and_take()
NotImplementedError:
```

```
> <ipython-input-24-0f6b26649b2e>(5)pick_and_take()
      3 def pick_and_take():
      4     picked = numpy.random.randint(0, 1000)
----> 5     raise NotImplementedError()
      6
      7 pick_and_take()
```

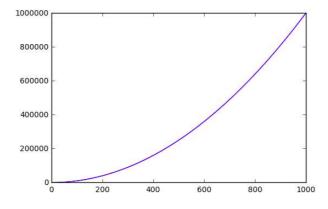
```
ipdb>
```

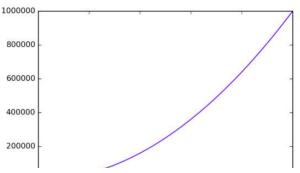
## 15. IPython Magic - High-resolution plot outputs for Retina notebooks

One line of IPython magic will give you double resolution plot output for Retina screens, such as the more recent Macbooks. *Note: the example below won't render on non-retina screens* 

```
x = range(1000)
y = [i ** 2 for i in x]
plt.plot(x,y)
plt.show();
```

```
%config InlineBackend.figure_format = 'retina'
plt.plot(x,y)
plt.show();
```







## 16. Suppress the output of a final function.

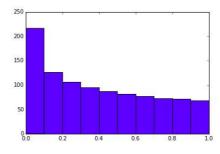
Sometimes it's handy to suppress the output of the function on a final line, for instance when plotting. To do this, you just add a semicolon at the end.

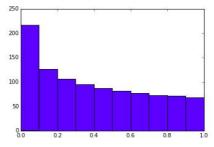
```
%matplotlib inline
from matplotlib import pyplot as plt
import numpy
x = numpy.linspace(0, 1, 1000)**1.5
```

```
# Here you get the output of the function \operatorname{plt.hist}(\mathbf{x})
```

```
(array([ 216., 126., 106., 95., 87., 81., 77., 73., 71., 68.]),
array([ 0. , 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9, 1. ]),
<a list of 10 Patch objects>)
```

```
# By adding a semicolon at the end, the output is suppressed. 
 plt.hist(x);
```





## 17. Executing Shell Commands

It's easy to execute a shell command from inside your notebook. You can use this to check what datasets are in available in your working folder:

```
!ls *.csv
```

nba\_2016.csv titanic.csv

pixar\_movies.csv whitehouse\_employees.c

Or to check and manage packages.

```
!pip install numpy
!pip list | grep pandas
```

```
Requirement already satisfied (use --upgrade to upgrade): numpy in /Library/Frameworks/Python.fra pandas (0.18.1)
```

## 18. Using LaTeX for forumlas

When you write LaTeX in a Markdown cell, it will be rendered as a formula using MathJax.

This:

```
\\( P(A \mid B) = \frac{P(B \mid A) \, P(A)}{P(B)} \\)
```

Becomes this:

$$P(A \mid B) = \frac{P(B|A), P(A)}{P(B)}$$

Markdown is an important part of notebooks, so don't forget to use its expressiveness!

## 19. Run code from a different kernel in a notebook

If you want to, you can combine code from multiple kernels into one notebook.

Just use IPython Magics with the name of your kernel at the start of each cell that you want to use that Kernel for:

- %bash
- %%HTML
- %python2
- %%python3
- %%ruby
- %%perl

```
%%bash
for i in {1..5}
do
echo "i is $i"
done
```

```
i is 1
i is 2
i is 3
i is 4
```

i is 5

### 20. Install other kernels for Jupyter

One of the nice features about Jupyter is ability to run kernels for different languages. As an example, here is how to get and R kernel running.

#### Easy Option: Installing the R Kernel Using Anaconda

If you used Anaconda to set up your environment, getting R working is extremely easy. Just run the below in your terminal:

conda install -c r r-essentials

#### Less Easy Option: Installing the R Kernel Manually

If you are not using Anaconda, the process is a little more complex. Firstly, you'll need to install R from <u>CRAN</u> if you haven't already.

Once that's done, fire up an R console and run the following:

install.packages(c('repr', 'IRdisplay', 'crayon', 'pbdZMQ', 'devtools'))
devtools::install\_github('IRkernel/IRkernel')
IRkernel::installspec() # to register the kernel in the current R installation

## 21. Running R and Python in the same notebook.

The best solution to this is to install  $\underline{rpy2}$  (requires a working version of R as well), which can be easily done with pip:

pip install rpy2

You can then use the two languages together, and even pass variables inbetween:

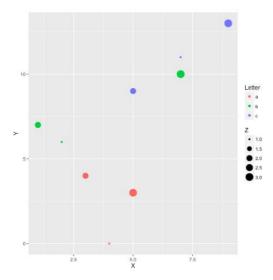
%load\_ext rpy2.ipython

%R require(ggplot2)

array([1], dtype=int32)

```
'X': [4, 3, 5, 2, 1, 7, 7, 5, 9],
'Y': [0, 4, 3, 6, 7, 10, 11, 9, 13],
'Z': [1, 2, 3, 1, 2, 3, 1, 2, 3]
})
```

```
%R -i df
ggplot(data = df) + geom_point(aes(x = X, y= Y, color = Letter, size = Z))
```



Example courtesy Revolutions Blog

## 22. Writing functions in other languages

Sometimes the speed of numpy is not enough and I need to write some fast code. In principle, you can compile function in the dynamic library and write python wrappers...

But it is much better when this boring part is done for you, right?

You can write functions in cython or fortran and use those directly from python code.

First you'll need to install:

```
!pip install cython fortran-magic
```

%load\_ext Cython

```
%cython

def myltiply_by_2(float x):

return 2.0 * x
```

```
myltiply_by_2(23.)
```

Personally I prefer to use fortran, which I found very convenient for writing number-crunching functions. More details of usage can be found here.

```
%load_ext fortranmagic
```

```
%fortran
subroutine compute_fortran(x, y, z)
  real, intent(in) :: x(:), y(:)
  real, intent(out) :: z(size(x, 1))

z = sin(x + y)
end subroutine compute_fortran
```

```
compute_fortran([1, 2, 3], [4, 5, 6])
```

There are also different jitter systems which can speed up your python code. More examples can be found here.

## 23. Multicursor support

Jupyter supports mutiple cursors, similar to Sublime Text. Simply click and drag your mouse while holding down Alt.

```
In []: x = [
    'one'
    'two'
    'three'
    'four'
    'five'
]
```

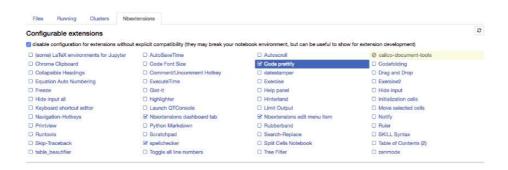
Multicursor support.

## 24. Jupyter-contrib extensions

 $\label{lem:decomposition} \underline{\text{Jupyter-contrib extensions}} \text{ is a family of extensions which give Jupyter a lot more functionality, including e.g. jupyter spell-checker and code-formatter.}$ 

The following commands will install the extensions, as well as a menu based configurator that will help you browse and enable the extensions from the main Jupyter notebook screen.





The nbextension configurator.

## 25. Create a presentation from a Jupyter notebook.

Damian Avila's <u>RISE</u> allows you to create a powerpoint style presentation from an existing notebook.

You can install RISE using conda:

```
conda install -c damianavila82 rise
```

Or alternatively pip:

```
pip install RISE
```

And then run the following code to install and enable the extension:

```
jupyter-nbextension install rise --py --sys-prefix
jupyter-nbextension enable rise --py --sys-prefix
```

## 26. The Jupyter output system

Notebooks are displayed as HTML and the cell output can be HTML, so you can return virtually anything: video/audio/images.

In this example I scan the folder with images in my repository and show thumbnails of the first 5:

```
import os
from IPython.display import display, Image
names = [f for f in os.listdir('../images/ml_demonstrations/') if f.endswith('.png')]
for name in names[:5]:
    display(Image('../images/ml_demonstrations/' + name, width=100))
```









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We can create the same list with a bash command, because magics and bash calls return python variables:

```
names = !ls ../images/ml_demonstrations/*.png
names[:5]
```

```
['../images/ml_demonstrations/colah_embeddings.png',
'../images/ml_demonstrations/convnetjs.png',
'../images/ml_demonstrations/decision_tree.png',
'../images/ml_demonstrations/decision_tree_in_course.png',
'../images/ml_demonstrations/dream_mnist.png']
```

## 27. 'Big data' analysis

A number of solutions are available for querying/processing large data samples:

- ipyparallel (formerly ipython cluster) is a good option for simple map-reduce operations in python. We use it in rep to train many machine learning models in parallel
- pyspark
- · spark-sql magic %%sql

## 28. Sharing notebooks

The easiest way to share your notebook is simply using the notebook file (.ipynb), but for those who don't use Jupyter, you have a few options:

- Convert notebooks to html file using the File > Download as > HTML Menu option.
- Share your notebook file with <u>gists</u> or on github, both of which render the notebooks.
   See this example.
  - If you upload your notebook to a github repository, you can use the handy
     <u>mybinder</u> service to allow someone half an hour of interactive Jupyter access to
     your repository.
- Setup your own system with jupyterhub, this is very handy when you organize minicourse or workshop and don't have time to care about students machines.
- Store your notebook e.g. in dropbox and put the link to <u>nbviewer</u>. nbviewer will render the notebook from whichever source you host it.
- Use the File > Download as > PDF menu to save your notebook as a PDF. If you're
  going this route, I highly recommend reading Julius Schulz's excellent article <u>Making</u>
  publication ready Python notebooks.
- Create a blog using Pelican from your Jupyter notebooks.

## What are your favorites?

Let me know what your favorite Jupyter notebook tips are.

At <u>Dataquest</u>, our interactive guided projects use Jupyter notebooks to building data science projects and get a job in data. If you're interested, you can <u>signup and do our first module for free</u>.

I also recommend the links below for further reading:

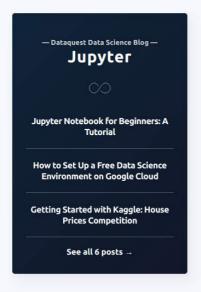
- IPython built-in magics
- Nice interactive presentation about jupyter by Ben Zaitlen
- Advanced notebooks part 1: magics and part 2: widgets
- · Profiling in python with jupyter
- · 4 ways to extend notebooks
- · IPython notebook tricks
- · Jupyter vs Zeppelin for big data
- Making publication ready Python notebooks.

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## Josh Devlin Data Scientist at Dataquest.io. Loves Data and Aussie Rules Football. Australian living in Texas.

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