28 Jupyter Notebook Tips, Tricks, and Shortcuts

October 12, 2016





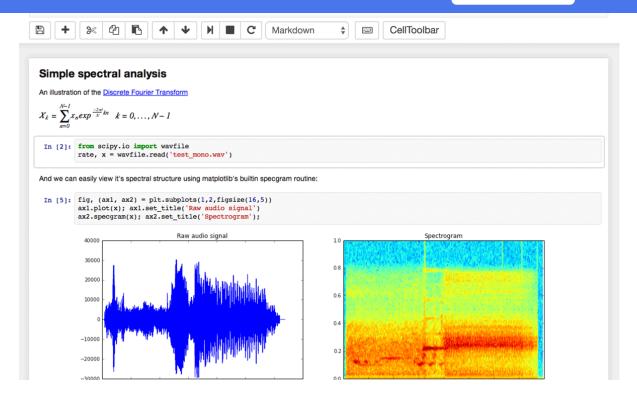


Jupyter Notebook

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. In this post, we've collected some of the top Jupyter notebook tips to quickly turn you into a Jupyter power user!

'Brilliantly wrong'. We nave 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 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. Now let's dive in to our list of 28 (and counting!) Jupyter notebook tips!



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!)

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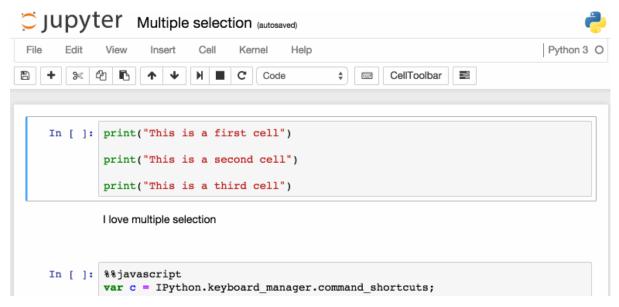
Sign up now!

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

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

- 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 + O Toggle cell output.
- Select Multiple Cells:
 - Shift + Jor 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 cell")
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 its 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()
```

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

GET 50% OFF

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.

?str.replace()

```
Docstring:

S.replace(old, new[, count]) -> str

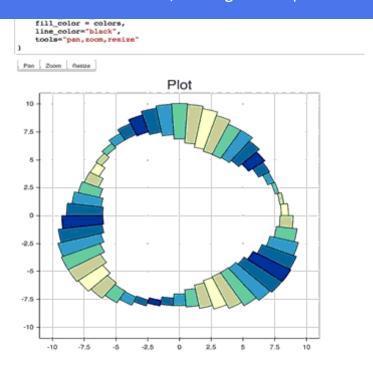
Return a copy of S with all occurrences of substring old replaced by new. If the optional argument count is given, only the first count occurrences are replaced.

Type: method_descriptor
```

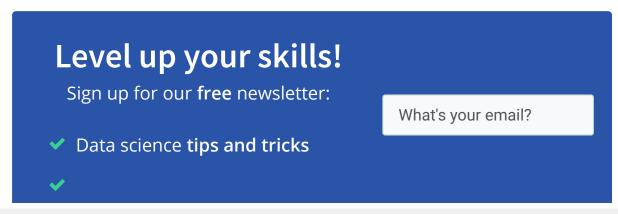
4. Plotting in notebooks

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 is done server-side.
- 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 modification.
- mpld3 provides alternative renderer (using d3) for matplotlib code. Quite nice, though incomplete.
- bokeh is a better option for building interactive plots.
- plot.ly can generate nice plots this used to be a paid service only but was recently open sourced.
- Altair is a relatively new declarative visualization library for Python. It's easy to use
 and makes great looking plots, however the ability to customize those plots is not
 nearly as powerful as in Matplotlib.



The Jupyter interface.



Industry news and updates

5. IPython Magic Commands

The %matplotlib inline you saw above was an example of a *IPython*Magic command. Being based on the IPython kernel, Jupyter has access to all the Magics from the IPython kernel, and they can make your life a lot easier!

```
# This will list all magic commands %lsmagic
```

```
Available line magics:
%alias %alias_magic %autocall %automagic %autosave %bookmark %cat %cd
%clear %colors %config %connect_info %cp %debug %dhist %dirs
%doctest_mode %ed %edit %env %gui %hist %history %killbgscripts %ldir
%less %lf %lk %ll %load %load_ext %loadpy %logoff %logon %logstart
%logstate %logstop %ls %lsmagic %lx %macro %magic %man %matplotlib
%mkdir %more %mv %notebook %page %pastebin %pdb %pdef %pdoc %pfile
```

```
%rehashx %reload_ext %rep %rerun %reset %reset_selective %rm %rmdir
%run %save %sc %set_env %store %sx %system %tb %time %timeit %unalias
%unload_ext %who %who_ls %whos %xdel %xmode
Available cell magics:%%! %%HTML %%SVG %%bash %%capture %%debug %%file
%%html %%javascript %%js %%latex %%perl %%prun %%pypy %%python
%%python2 %%python3 %%ruby %%script %%sh %%svg %%sx %%system %%time
%%timeit %%writefile
Automagic is ON, % prefix IS NOT needed for line magics.
```

I recommend browsing the documentation for all IPython Magic commands as you'll no doubt find some that work for you. A few of my favorites are below:

6. IPython Magic – %env: Set Environment Variables

You can manage environment variables of your notebook without restarting the jupyter server process. Some libraries (like theano) use environment variables to control behavior, %env is the most convenient way.

```
# lists all environment variables# The line below sets the environment
# variable
%env OMP_NUM_THREADS%env OMP_NUM_THREADS=4
```

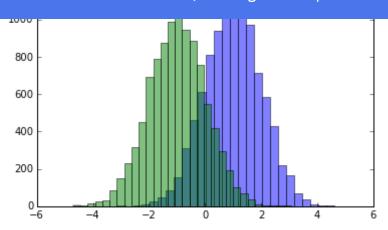
```
env: OMP NUM THREADS=4
```

7. IPython Magic – %run: Execute python code

%run can execute python code from .py files – this is well-documented behavior. Lesser known is the fact that it can also execute other jupyter notebooks, which can quite useful.

Note that using %run is not the same as importing a python module.

```
# this will execute and show the output from
# all code cells of the specified notebook
%run ./two-histograms.ipynb
```



8. IPython Magic – %load: Insert the code from an external script

This will replace the contents of the cell with an external script. You can either use a file on your computer as a source, or alternatively a URL.

```
# Before Running
%load ./hello_world.py
```

```
# %load ./hello_world.py
if __name__ == "__main__":
   print("Hello World!")
```

Hello World!

9. IPython Magic – %store: Pass variables between notebooks.

The %store command lets you pass variables between two different notebooks.

```
data = 'this is the string I want to pass to different notebook'
%store data
del data # This has deleted the variable
```

Now, in a new notebook...

```
%store -r data
print(data)
```

this is the string I want to pass to different notebook

10. IPython Magic – %who: List all variables of global scope.

The <code>%who</code> command without any arguments will list all variables that existing in the global scope. Passing a parameter like <code>str</code> will list only variables of that type.

```
two = "for the show"
three = "to get ready now go cat go"
%who str
```

```
one three two
```

11. IPython Magic – Timing

There are two IPython Magic commands that are useful for timing –

%%time and %timeit. These are especially handy when you have some slow code and you're trying to indentify where the issue is.

%%time will give you information about a single run of the code in your cell.

```
%%time
import time
for _ in range(1000):
   time.sleep(0.01) # sleep for 0.01 seconds
```

S

%%timeit uses the Python timeit module which runs a statement 100,000 times (by default) and then provides the mean of the fastest three times.

```
import numpy
%timeit numpy.random.normal(size=100)
```

The slowest run took 7.29 times longer than the fastest. This could mean that an intermediate result is being cached.

100000 loops, best of 3: 5.5 µs per loop

12. IPython Magic – %writefile and %pycat: Export the contents of a

external script

Using the <code>%%writefile</code> magic saves the contents of that cell to an external file. <code>%pycat</code> does the opposite, and shows you (in a popup) the syntax highlighted contents of an external file.

```
%%writefile 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)
```

```
Writing pythoncode.py
```

```
%pycat pythoncode.py
```

```
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 internal function was called within the statement, the time each call took as well as the cumulative time of all runs of the function.

```
26324 function calls in 0.556 seconds
Ordered by: internal time
ncalls tottime percall cumtime percall filename:lineno(function)
10000 0.527 0.000 0.528 0.000 :2(append_if_not_exists)
10000 0.022 0.000 0.022 0.000 {method 'randint' of
'mtrand.RandomState' objects}
1 0.006 0.006 0.556 0.556 :6(some_useless_slow_function)
6320 0.001 0.000 0.001 0.000 {method 'append' of 'list' objects}
1 0.000 0.000 0.556 0.556 :1()
1 0.000 0.000 0.556 0.556 {built-in method exec}
1 0.000 0.000 0.000 0.000 {method 'disable' of '_lsprof.Profiler' objects}
```

14. IPython Magic – Debugging with %pdb

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

```
7 pick_and_take()
NotImplementedError:
```

```
> (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 – Highresolution plot outputs for Retina notebooks

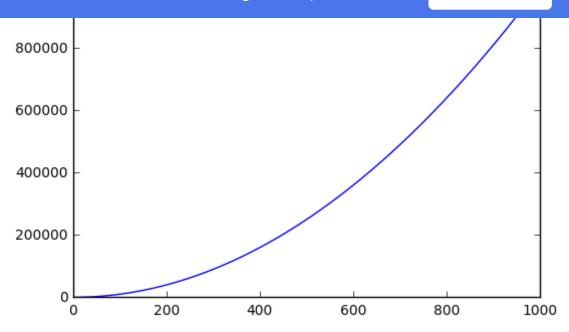
screens

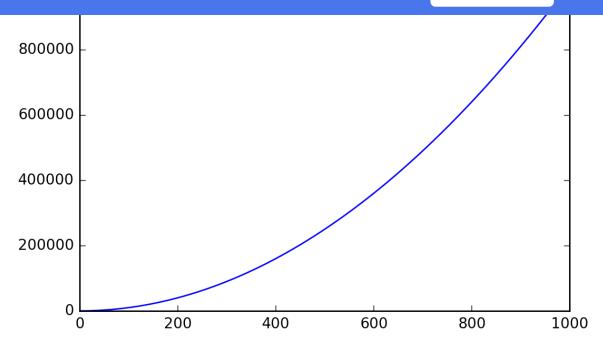
```
x = range(1000)
y = [i ** 2 for i in x]
plt.plot(x,y)
plt.show();
```

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









16. Suppress the output of a final function.

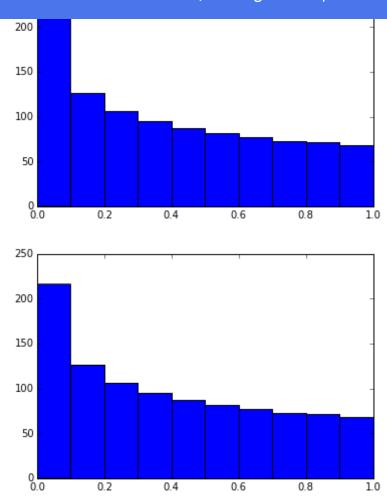
Sometimes it's handy to suppress the output of the function on a final line, for instance when plotting. To do this, you just add a semicolon at the end.

%matplotlib inline
from matplotlib import pyplot as plt

```
# Here you get the output of the function
plt.hist(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 Snell 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.csv
```

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.framework/Versions/3.4/lib/python3.4/site-packages pandas (0.18.1)

18. Using Latex for forumias

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:

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:

- %%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

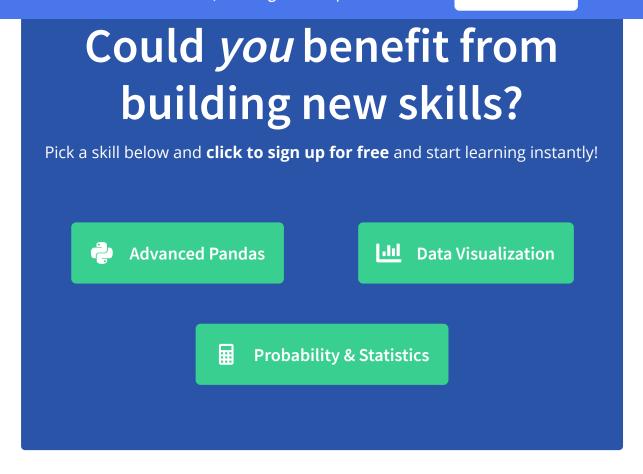
If you used Anaconda to set up your environment, getting R working is extremely easy. Just run the below in your terminal:

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 CRAN if you haven't already.

```
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.

```
pip install rpy2
```

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

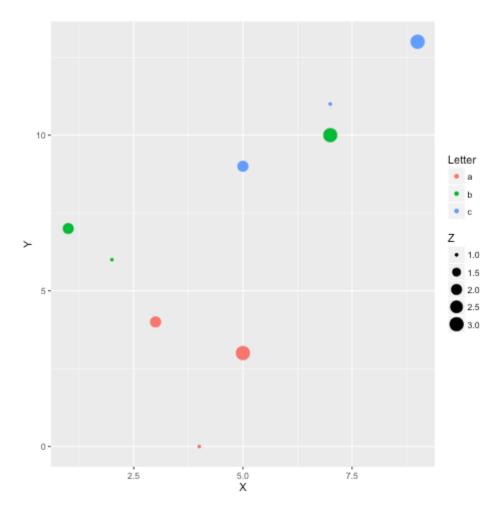
```
%load_ext rpy2.ipython
```

```
%R require(ggplot2)
```

```
array([1], dtype=int32)
```

```
import pandas as pd df = pd.DataFrame({
  'Letter': ['a', 'a', 'a', 'b', 'b', 'b', 'c', 'c', 'c'],
  '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
```

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

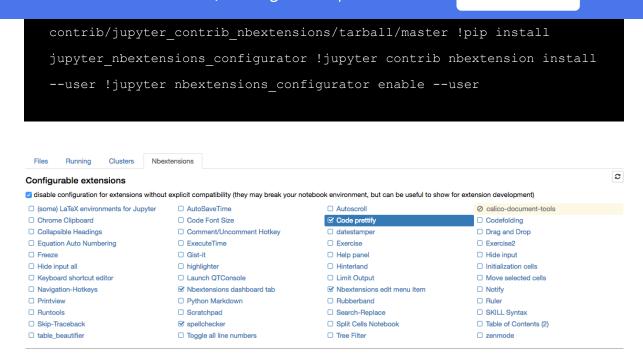
Jupyter supports mutiple cursors, similar to Sublime Text. Simply click and drag your mouse while holding down Alt.

Multicursor support.

24. Jupyter-contrib extensions

Jupyter-contrib extensions 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 RISE allows you to create a powerpoint style presentation from an existing notebook.

conda install -c damianavila82 rise

Or alternatively pip:

pip install RISE

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

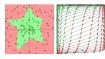
jupyter-nbextension install rise --py --sys-prefix jupyter-nbextension
enable rise --py --sys-prefix

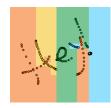
26. The Jupyter output system

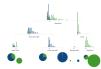
Notebooks are displayed as HTML and the cell output can be HTML, so you can return virtually anything: video/audio/images.

```
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

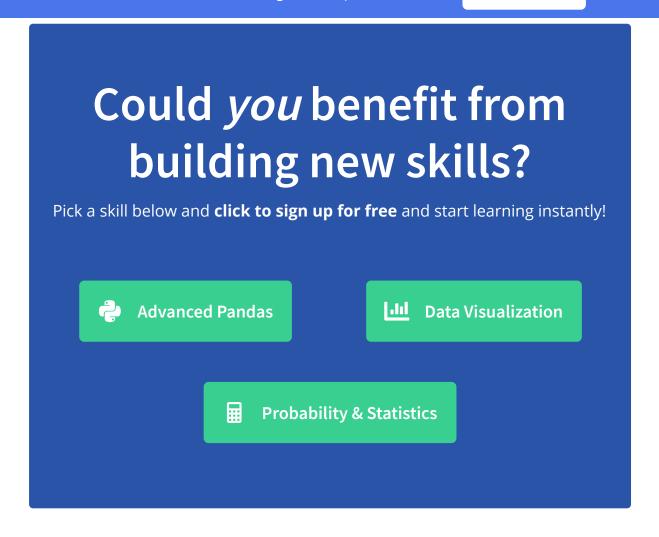
A number of solutions are available for querying/processing large data samples:

 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

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 files using the File > Download as > HTML Menu option.
- Upload your .ipynb file to Google Colab.
- Share your notebook file with gists 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 mybinder 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 mini-course or workshop and don't have time to care about students machines.
- Store your notebook e.g. in dropbox and put the link to nbviewer. 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 Making publication ready Python notebooks.



What are your favorites?

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

I also recommend the links below for further reading:

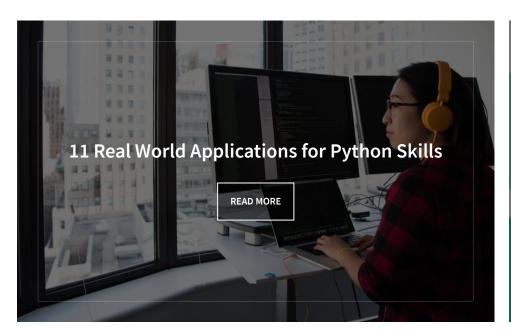
- IPython built-in magics
- Nice interactive presentation about jupyter by Ben Zaitlen
- Advanced notebooks part 1: magics and part 2: widgets
- Profiling in python with jupyter
- 4 ways to extend notebooks
- IPython notebook tricks
- Jupyter vs Zeppelin for big data
- Making publication ready Python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

Josh Devlin

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

TAGS beginner, Jupyter notebook, jupyter notebooks, Learn Python, Learn R, Resources, tips and tricks, tutorial, Tutorials

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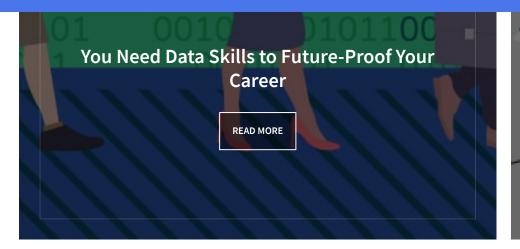








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