Exploring Science on Twitter

with IPython Notebook and Python Pandas

Brenda Moon

@brendam

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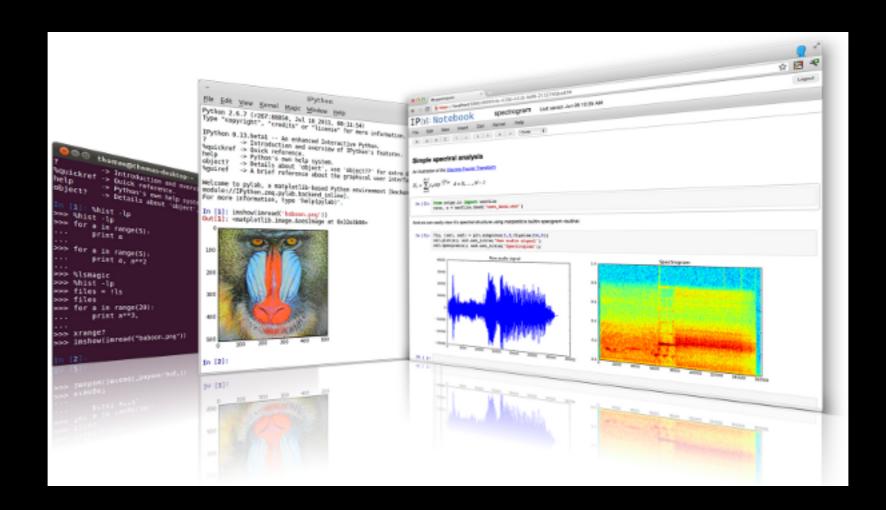
@brendam

Python Environment

- virtualenv keep environment isolated
- virtualenvwrapper manage virtualenv easily
- cpvirtualenv copy virtualenv
- setvirtualenvproject set work directory

IPython Notebook

http://www.ipython.org/



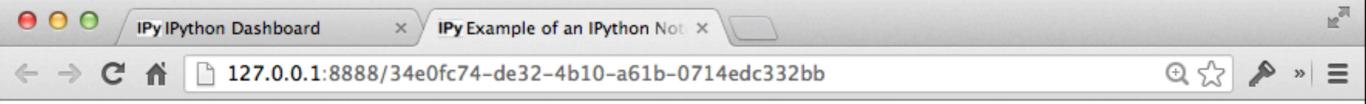


ruby:~ brenda\$ workon pyconau2013

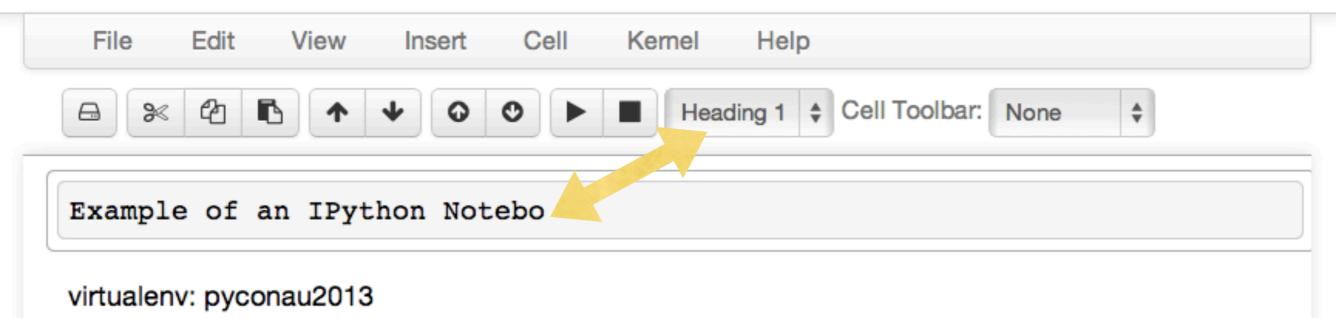
(pyconau2013)ruby:pyconau2013 brenda\$ ipython notebook --pylab inline

- \$ ipython notebook --pylab inline
-] Using existing profile dir: u'/Users/brenda/.ipython/profile_default'
- Using MathJax from CDN: http://cdn.mathjax.org/mathjax/latest/MathJax.js
- Serving notebooks from local directory: /Volumes/HDD/Documents/pyconau2013
- The IPython Notebook is running at: http://127.0.0.1:8888/
-] Use Control-C to stop this server and shut down all kernels.





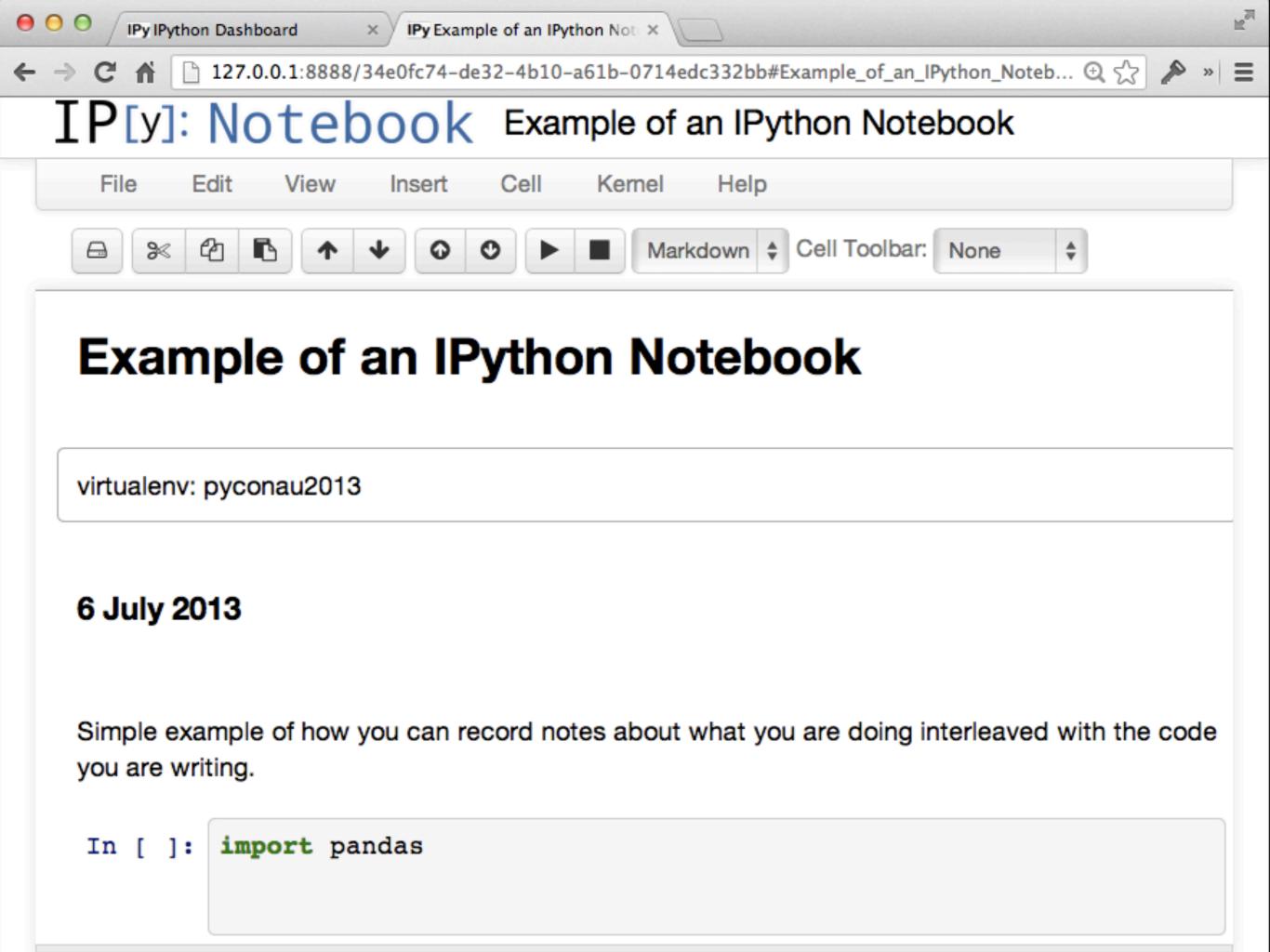
IP[y]: Notebook Example of an IPython Notebook



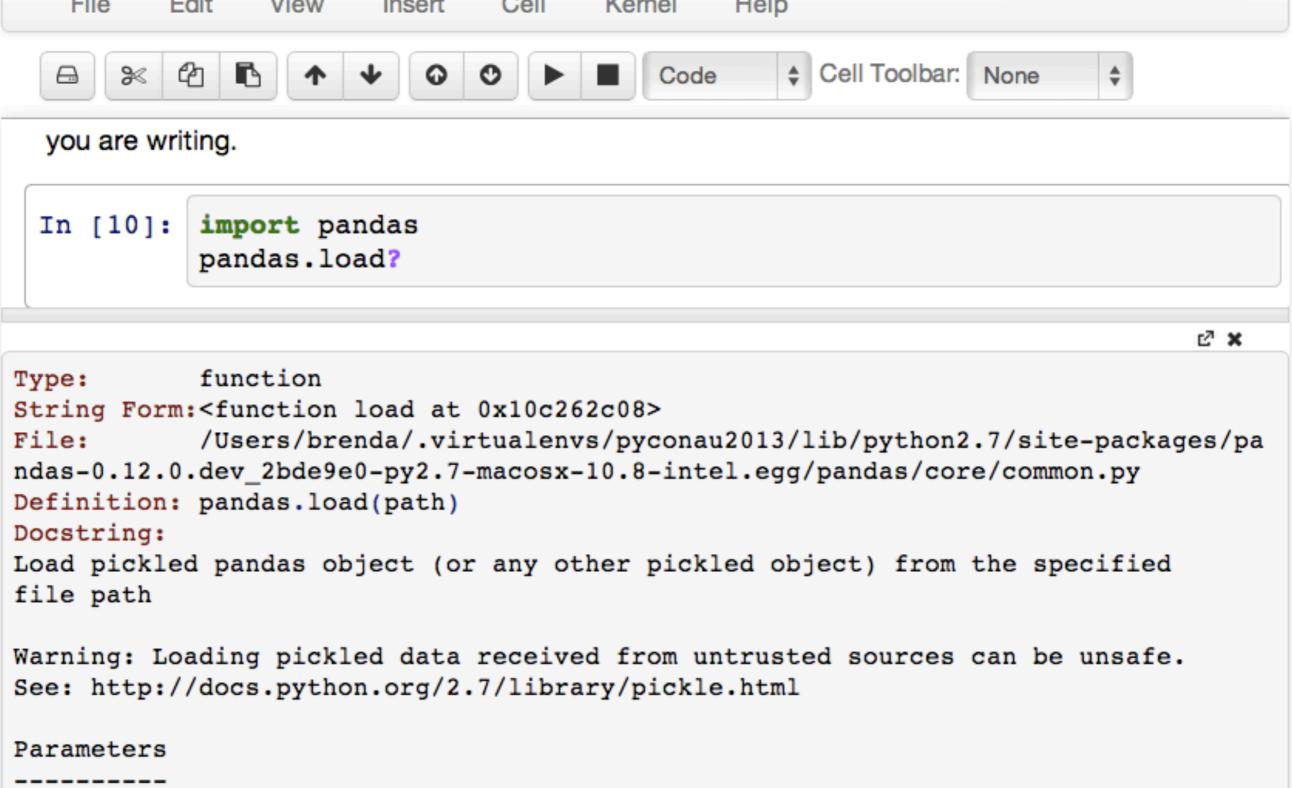
6 July 2013

Simple example of how you can record notes about what you are doing interleaved with the code you are writing.

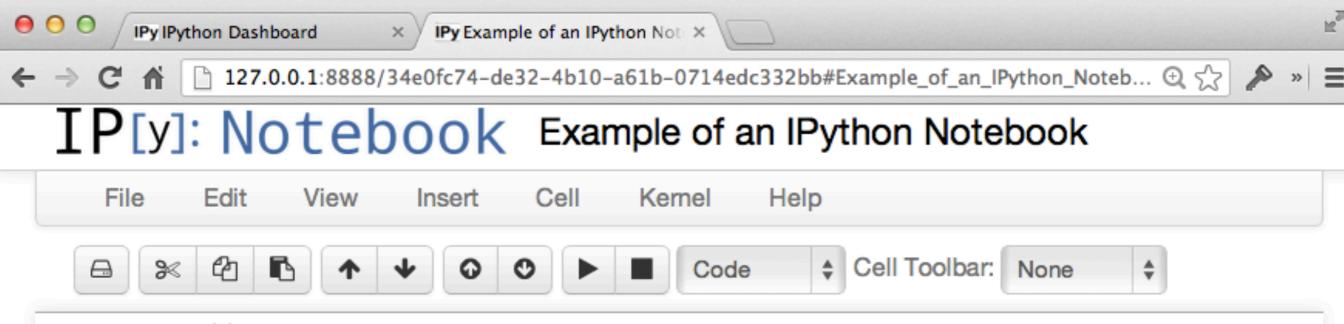
```
In [ ]: import pandas
```

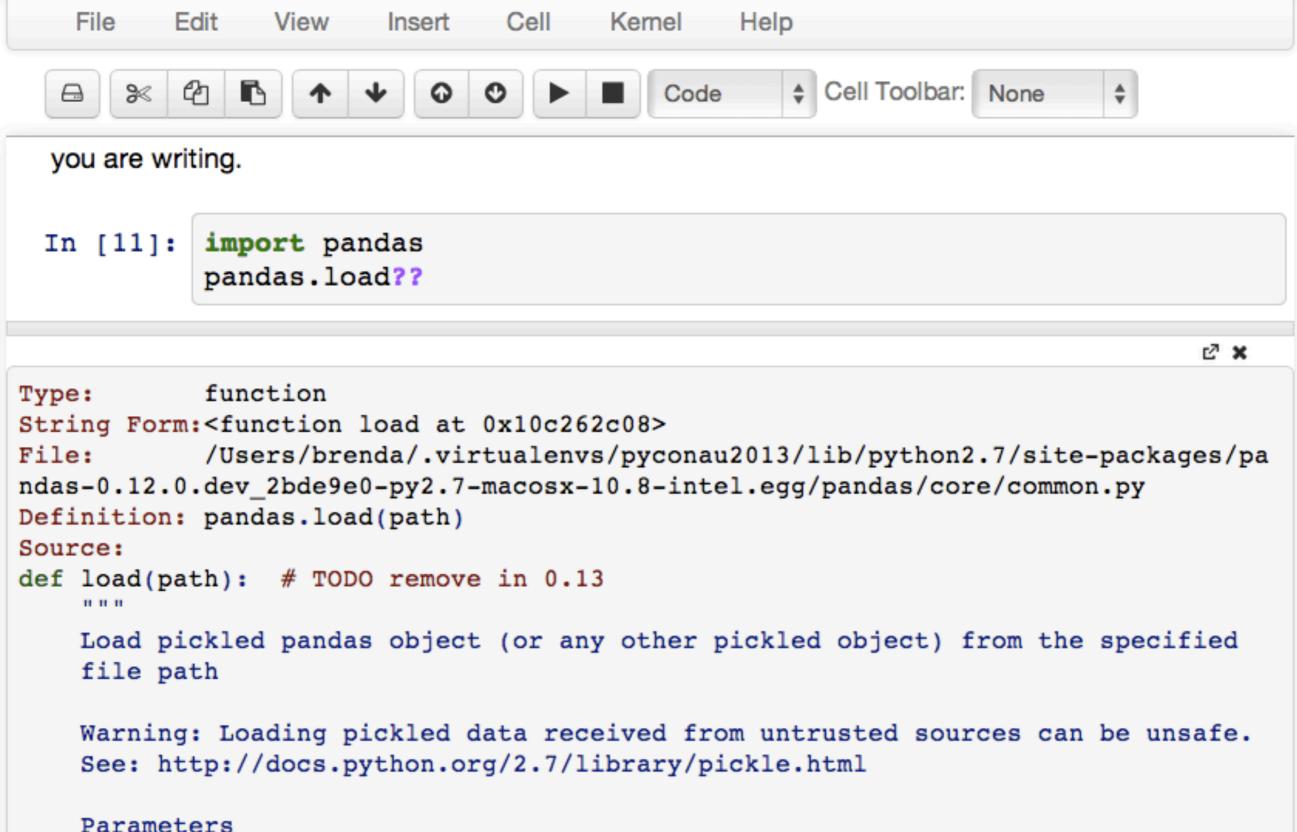






path : string







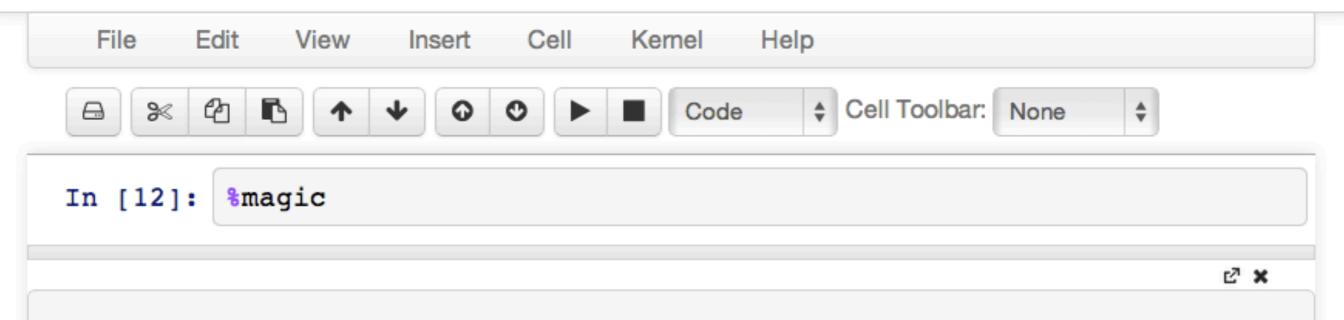


← → C ↑ 127.0.0.1:8888/34e0fc74-de32-4b10-a61b-0714edc332bb#Example_of_an_IPython_Noteb... ⊖ ☆





IP[y]: Notebook Example of an IPython Notebook



IPython's 'magic' functions

The magic function system provides a series of functions which allow you to control the behavior of IPython itself, plus a lot of system-type features. There are two kinds of magics, line-oriented and cell-oriented.

Line magics are prefixed with the % character and work much like OS command-line calls: they get as an argument the rest of the line, where arguments are passed without parentheses or quotes. For example, this will time the given statement::

%timeit range(1000)

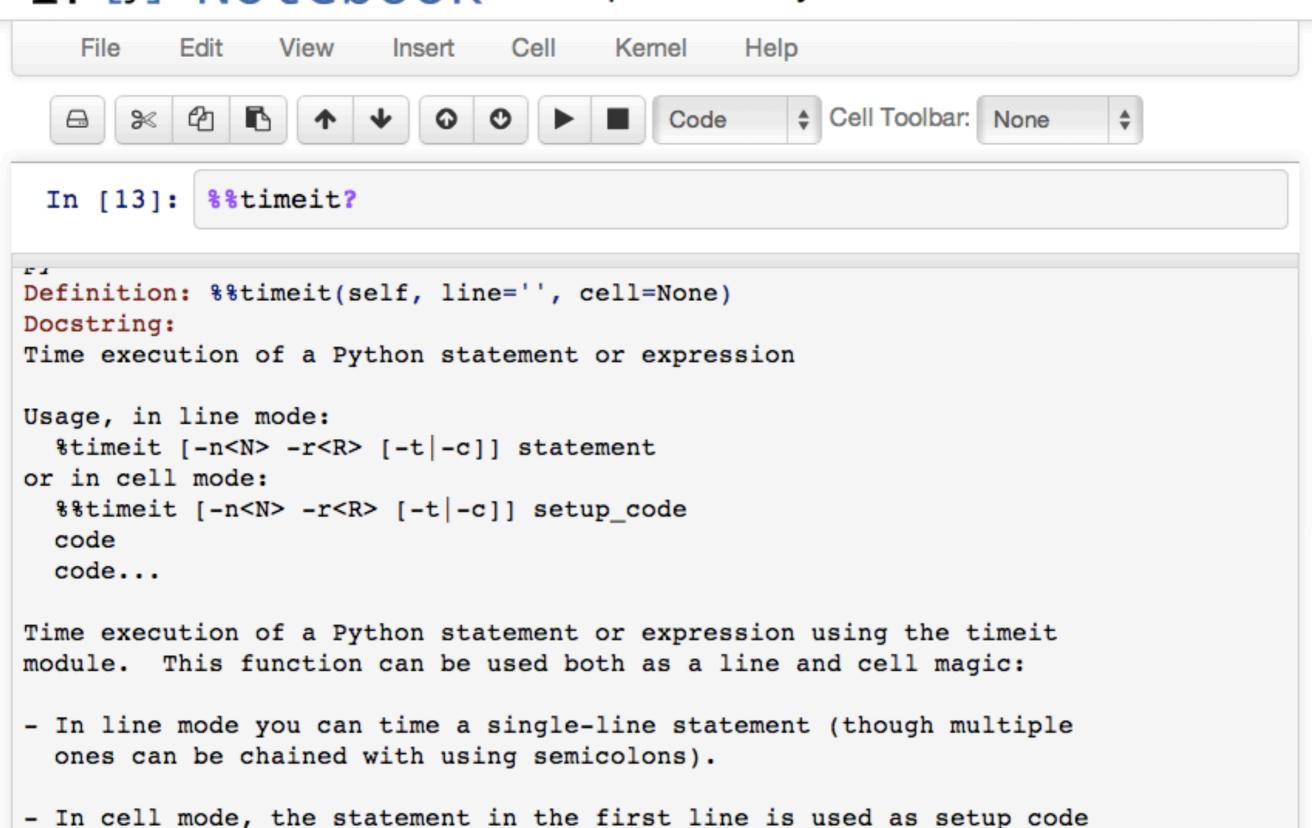
Cell magics are prefixed with a double %%, and they are functions that get as an argument not only the rest of the line, but also the lines below it in a separate argument. These magics are called with two arguments: the rest of the





← → C ↑ 127.0.0.1:8888/34e0fc74-de32-4b10-a61b-0714edc332bb#Example_of_an_IPython_Noteb... ⊙ ☆ ♪ » ≡

IP[y]: Notebook Example of an IPython Notebook



(executed but not timed) and the body of the cell is timed. The cell

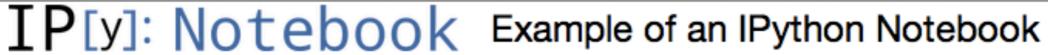


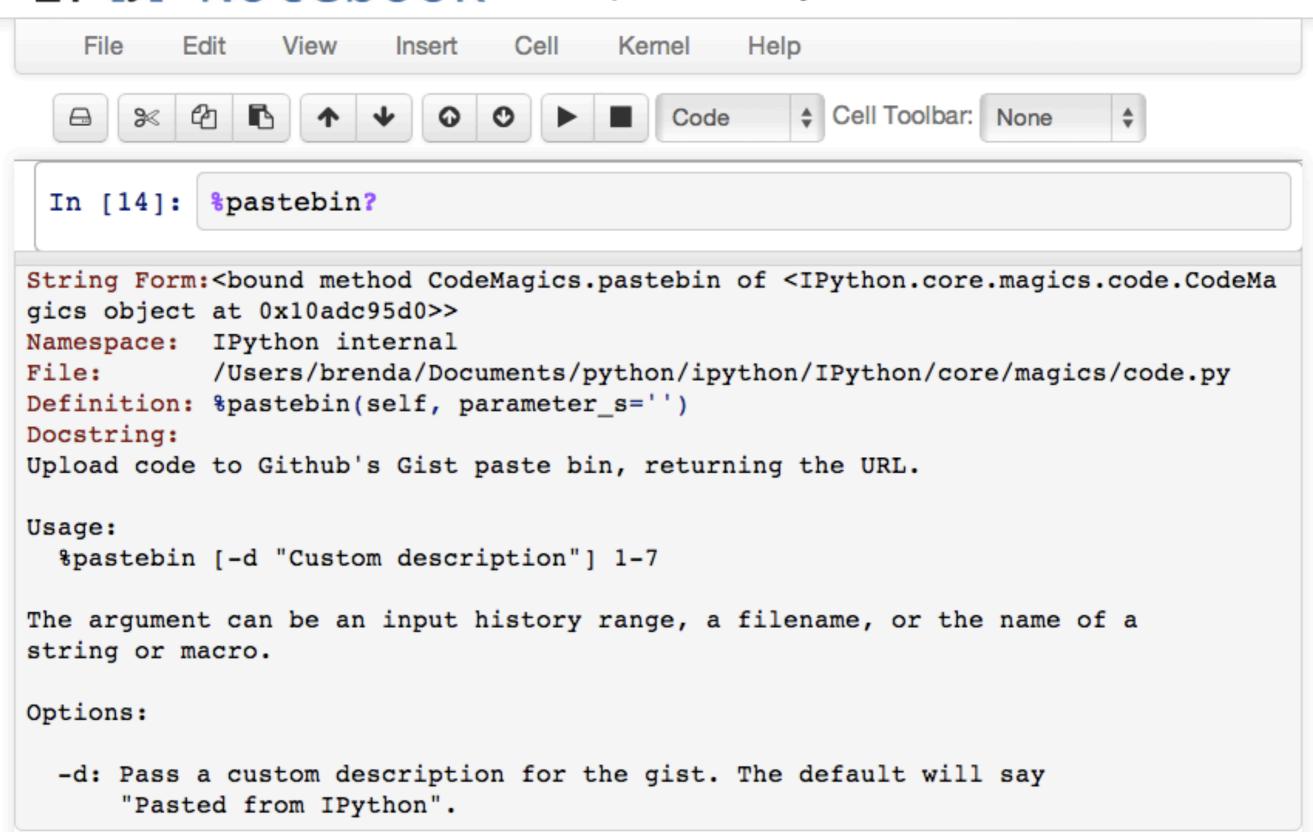


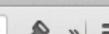
← → C ↑ 127.0.0.1:8888/34e0fc74-de32-4b10-a61b-0714edc332bb#Example_of_an_IPython_Noteb... ⊙ ☆ ♪











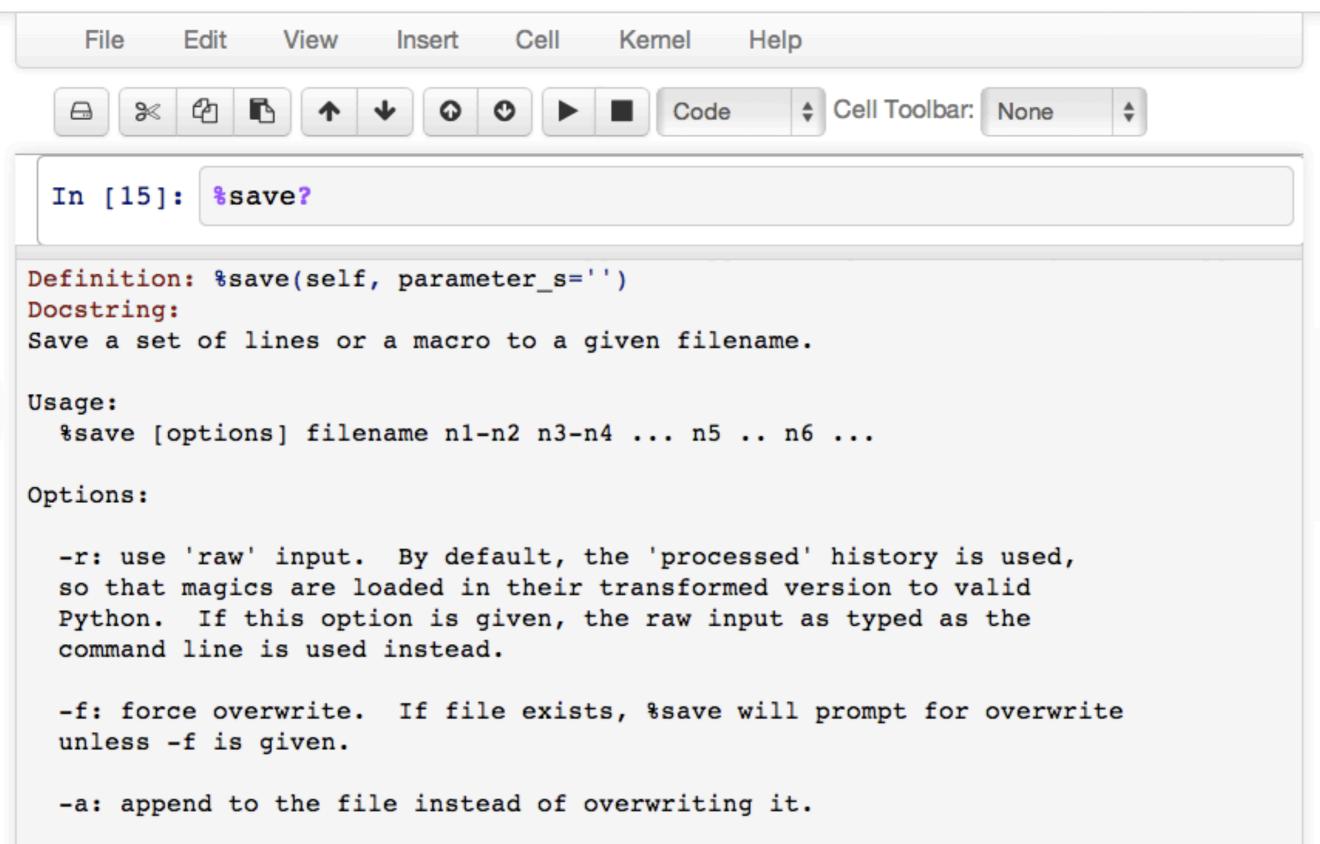






IP[y]: Notebook Example of an IPython Notebook

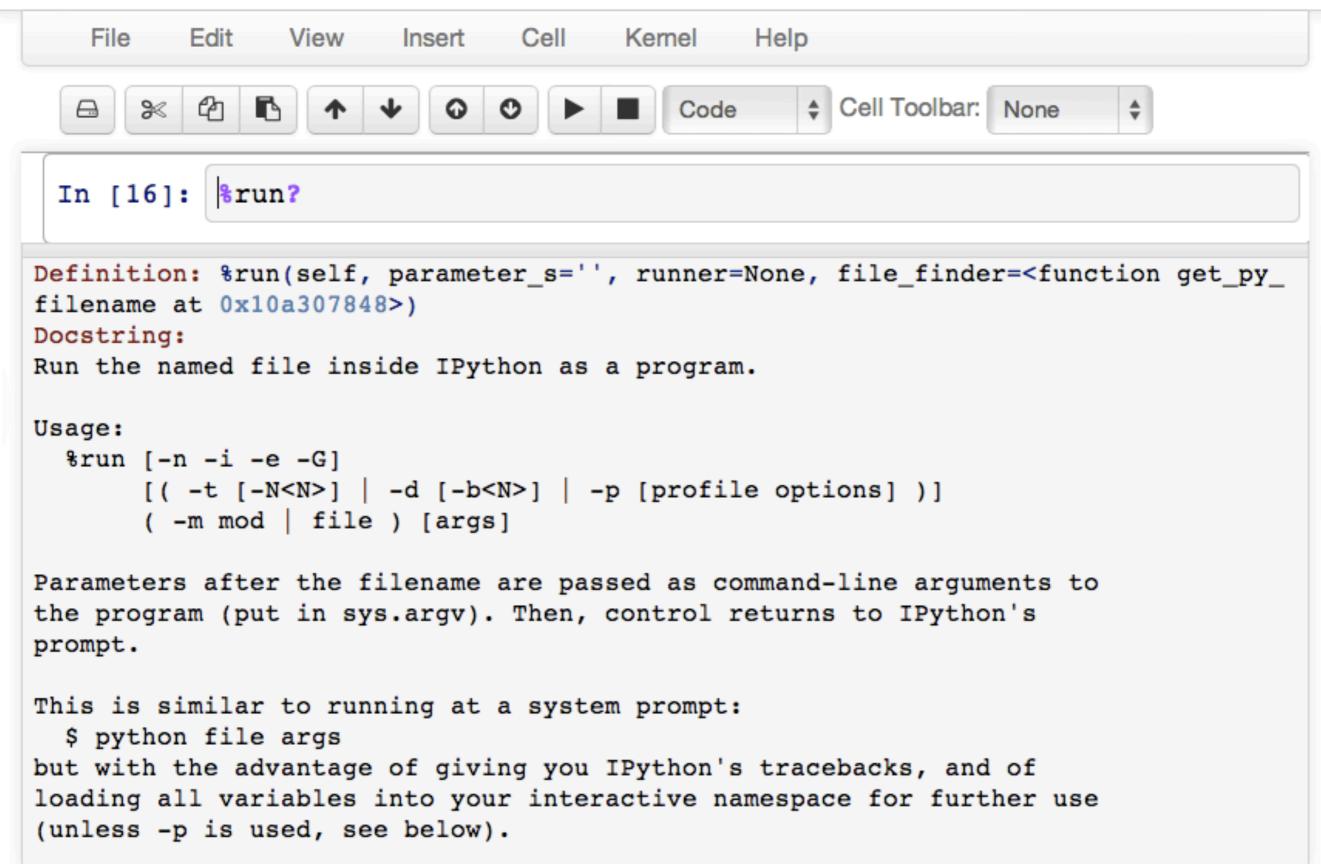
This function uses the same suntag as thistory for input ranges

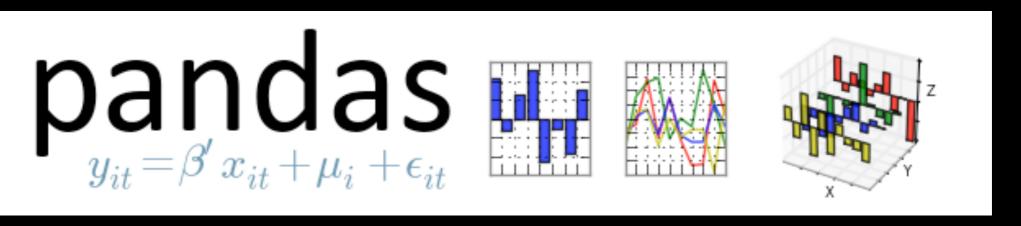






IP[y]: Notebook Example of an IPython Notebook





http://pandas.pydata.org/

"high-performance, easy-to-use data structures and data analysis tools"

- series (array like)
- dataFrame (table)

strong support for time based indexing

Tweets per Day

- Tweets containing the word 'science'
- 12 million 'science' tweets in 2011
- lets look at tweets per day

```
import pandas
import couchdbkit
server = couchdbkit.Server('http://brenda:XXXXXX@127.0.0.1:5984/')
tweetdb = server.get_db('tweets')
tweets = list(
    tweetdb.view(
        "tweetsPerDay/tweetsPerDay",
        reduce=True,
        group level=3,
        startkey=[2011, 1, 1],
        endkey=[2012, 1, 1]))
date list = [
    pandas.datetime(tweet["key"][0], (tweet["key"][1]), tweet["key"][2])
    for tweet in tweets]
date index = pandas.DatetimeIndex(date list)
data_list = [tweet["value"] for tweet in tweets]
science_tweets = pandas.Series(data_list, date_index)
# change to float so can have NaN values
science_tweets = science_tweets.astype(float)
# mask out the missing data period so it doesn't plot
science tweets['2011-03-31':'2011-04-12'] = numpy.NaN
# final check that start and end dates are correct
print 'first element: ', science_tweets.first('1D')
print 'last element: ', science_tweets.last('ID')
# check than the missing values don't plot.
science_tweets.plot()
science_tweets.to_pickle('dataFiles/2011TweetsPerDay-final.pkl')
```

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import pandas
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date list = [
    pandas.datetime(tweet["key"][0], (tweet["key"][1]), tweet["key"][2])
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date_index = pandas.DatetimeIndex(date_list)
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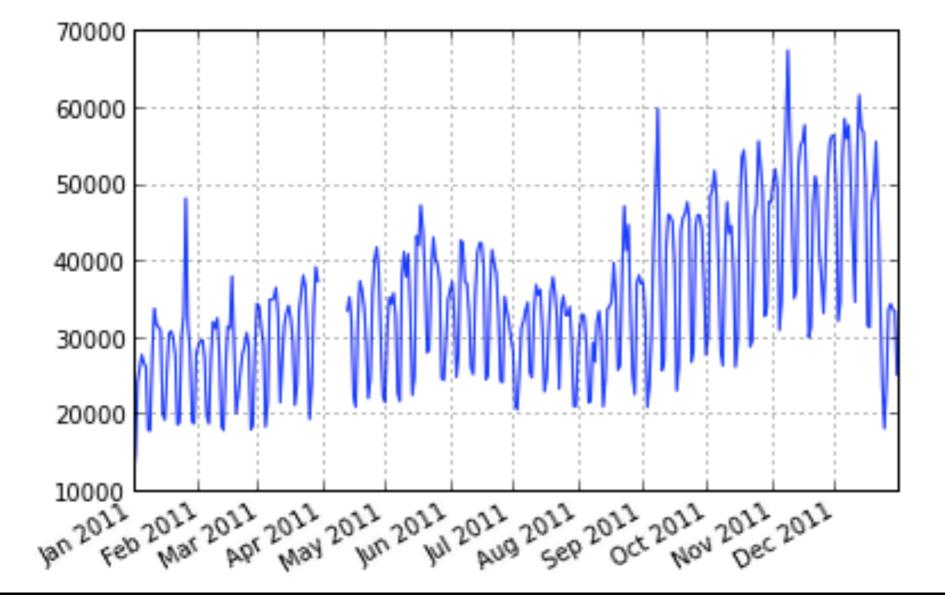
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science_tweets.plot()
science_tweets.to_pickle('dataFiles/2011TweetsPerDay-final.pkl')
```

first element: 2011-01-01 11070

dtype: float64

last element: 2011-12-31 25067

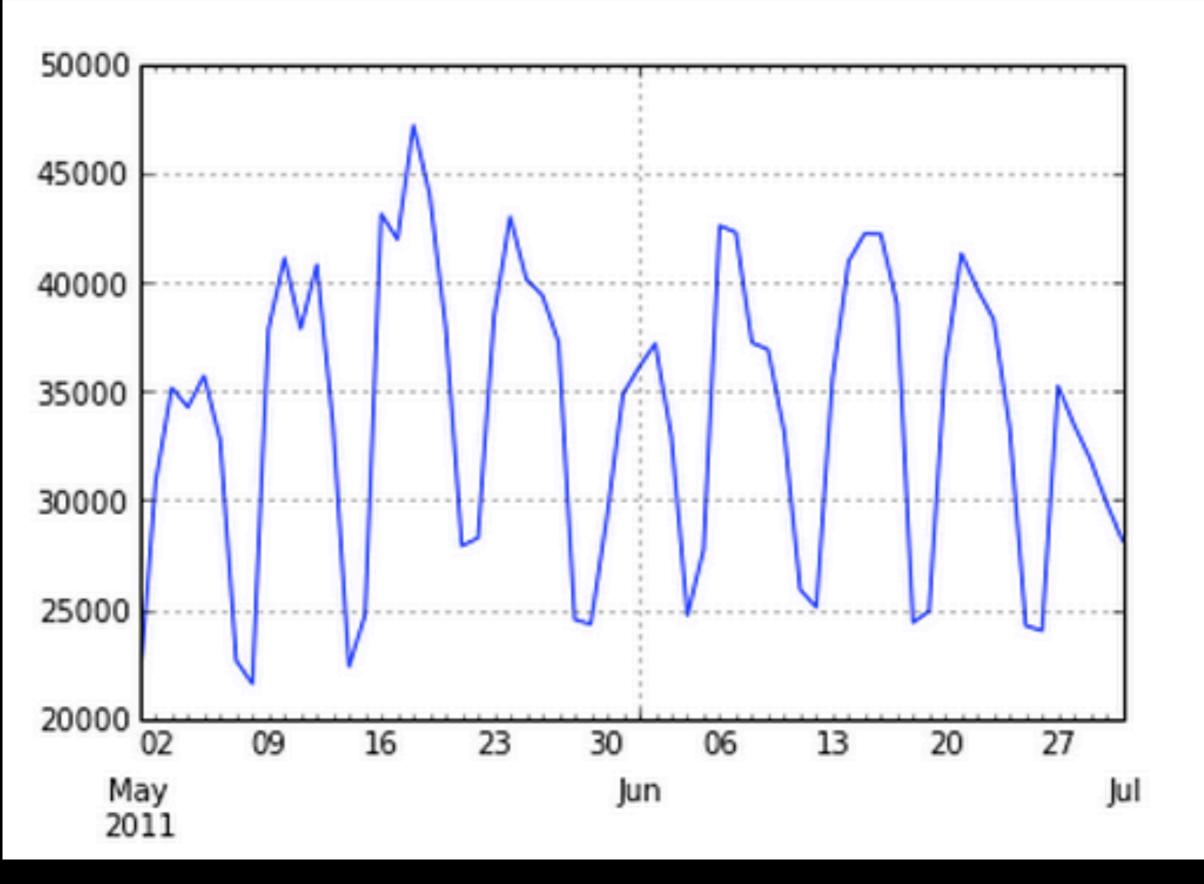
dtype: float64

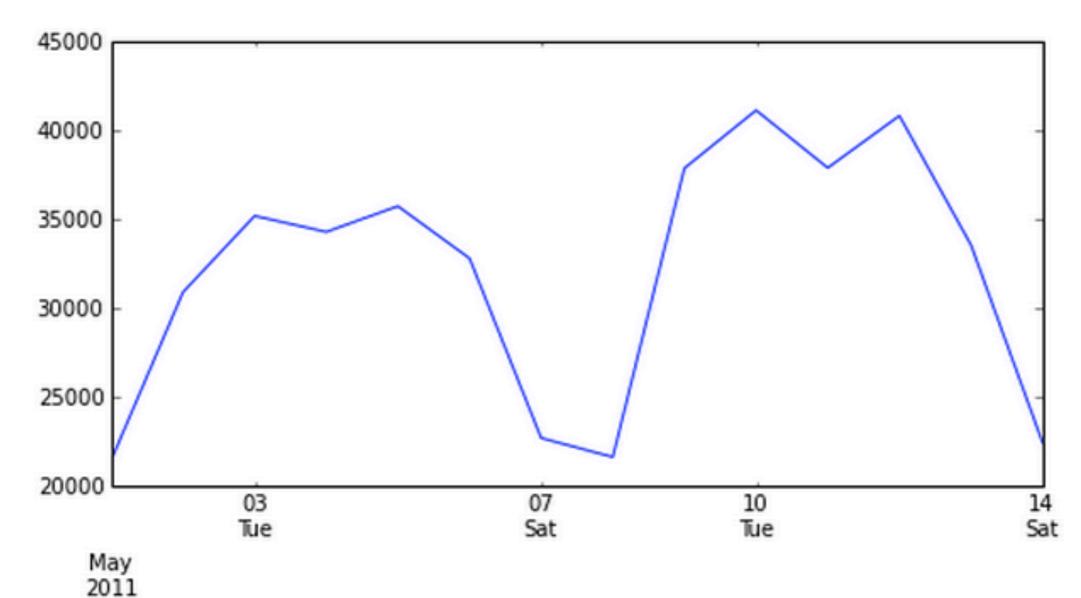


```
# different ways to access the start of the timeseries
print "science_tweets.first('ID')", science_tweets.first('ID')
print '\nscience tweets.head(1)', science tweets.head(1)
print '\nscience tweets[0]', science tweets[0]
print "\nscience tweets['2011-01-01']", science tweets['2011-01-01']
print "\nscience_tweets.first('1W')", science_tweets.first('1W')
science tweets.first('1D') 2011-01-01
                                        11070
dtype: float64
science_tweets.head(1) 2011-01-01 11070
dtype: float64
science_tweets[0] 11070.0
science_tweets['2011-01-01'] 11070.0
science tweets.first('1W') 2011-01-01 11070
2011-01-02 14542
dtype: float64
```

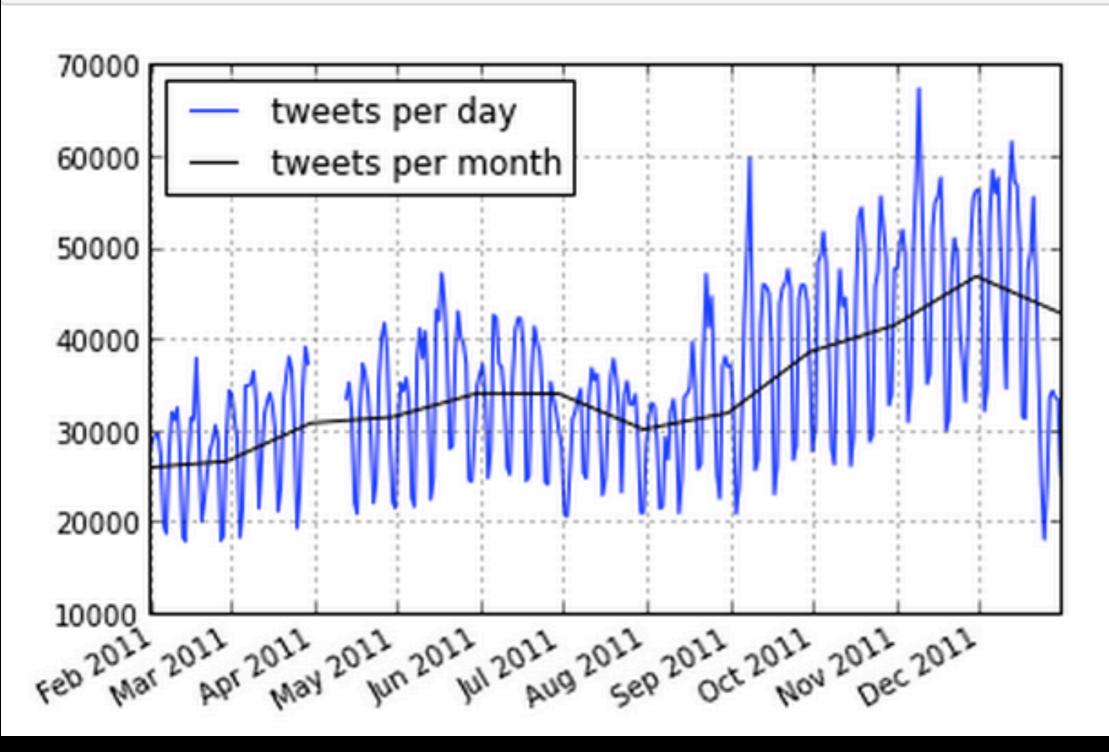
```
# ambiguity of date strings
print "\nscience tweets['2011-07-01']", science tweets['2011-07-01']
print "\nscience_tweets['2011-01-07']", science_tweets['2011-01-07']
print "\nscience tweets[pandas.datetime(2011,1,7)]"
print science tweets[pandas.datetime(2011,1,7)]
print "\npandas.to_datetime(['07-01-2011'], dayfirst=True)"
science_tweets[pandas.to_datetime(['07-01-2011'], dayfirst=True)]
science_tweets['2011-07-01'] 28146.0
science_tweets['2011-01-07'] 26095.0
science tweets[pandas.datetime(2011,1,7)]
26095.0
pandas.to_datetime(['07-01-2011'], dayfirst=True)
2011-01-07 26095
dtype: float64
```

zoom in using a date range
science_tweets['2011-05-01':'2011-07-01'].plot();





```
monthly_tweets = science_tweets.resample('M', how='mean')
science_tweets.plot(label='tweets per day')
monthly_tweets.plot('tweets per month', color='k')
legend(loc=0);
```



science_tweets.describe()

```
352.000000
count
         34646.264205
mean
         10178.287440
std
min
         11070.000000
         27626.750000
25%
50%
         33414.500000
75%
         40963.500000
          67375.000000
max
dtype: float64
```

	tweets per day	tweets per month
2011-01-29	18913	NaN
2011-01-30	18651	NaN
2011-01-31	27853	25897.903226
2011-02-01	28808	NaN
2011-02-02	29517	NaN

	tweets per day	tweets per month	cumulative	percent	dayofweek
2011-01- 01	11070	NaN	11070	0.090771	5
2011-01- 02	14542	NaN	25612	0.119241	6
2011-01- 03	24121	NaN	49733	0.197786	0
2011-01- 04	25984	NaN	75717	0.213062	1
2011-01- 05	27639	NaN	103356	0.226633	2

	tweets per day	tweets per month	cumulative	percent
dayofweek				
0	1833426	93883.930876	274073252	15.033646
1	2010642	33999.741935	276083894	16.486774
2	2084425	78640.716129	280636847	17.091776
3	2012464	64687.800000	280180783	16.501714
4	1758167	38536.800000	281938950	14.416540
5	1233825	74176.073477	283172775	10.117064
6	1262536	30059.516129	272239826	10.352487

	tweets per day	percent
dayofweek		
0	1833426	15.033646
1	2010642	16.486774
2	2084425	17.091776
3	2012464	16.501714
4	1758167	14.416540
5	1233825	10.117064
6	1262536	10.352487

```
df.groupby('dayofweek').sum()['percent'].sum()

100.0

print df.groupby('dayofweek').sum()['tweets per day'].sum()
print df['tweets per day'].sum()

12195485.0
12195485.0
```

```
df2 = df.groupby('dayofweek').sum()
print df2.to latex(columns=['tweets per day', 'percent'])
\begin{tabular}{lrr}
\toprule
{} & tweets per day & percent \\
\midrule
dayofweek &
                                          ١١
                            &
                    1833426 & 15.033646 \\
0
          &
                    2010642 & 16.486774 \\
          &
2
                    2084425 & 17.091776 \\
          &
                    2012464 & 16.501714 \\
          &
                    1758167 & 14.416540 \\
          &
5
          &
                    1233825 & 10.117064 \\
                    1262536 & 10.352487 \\
          &
\bottomrule
\end{tabular}
```

Questions?

Slides and notebook available on GitHub

https://github.com/brendam/pycon2013talk