PyData 2013 Highlights

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http://wiki.wgenhq.net/wiki/index.php/PyData_2013 Milen Pavlov 2013-11-21

0.1 Conference Themes

- · data analysis
- scale and parallelism
- visualization

0.2 Conference dates

```
In [8]: print "\n".join(['2013-11-%d' % day for day in range(8,11)])

2013-11-8
2013-11-9
2013-11-10
```

0.3 Venue

1 Chase Manhattan Plaza



In [10]: Image(filename='img/venue2.jpg')

Out [10]:



0.4 General Observations

In [12]: display(Image(filename='img/Borat_big_data.png'))
 print "When dealing with big data avoid ETLs, bring code to data"





For handle big data, solution is very simple: buy bigger monitor and use smaller font in the terminal.

When dealing with big data avoid ETLs, bring code to data

In [13]: display(Image('img/Borat_80_20_rule.png'))
print "These days learning algorithms are easy to apply, most time is spent exploring





In Data Science, 80% of time spent prepare data, 20% of time spent complain about need for prepare data.

These days learning algorithms are easy to apply, most time is spent exploring the data and its features

```
In [5]: display(Image("img/heart.png", width=100))
    display(Image("http://ipython.org/ipython-doc/rel-1.1.0/_images/ipynb_icon_128x128.png
    print "Presentation media: "
    print " iPython 20"
    print " Keynote 1"
    print " Google 1"
    print " PowerPoint 0"
```





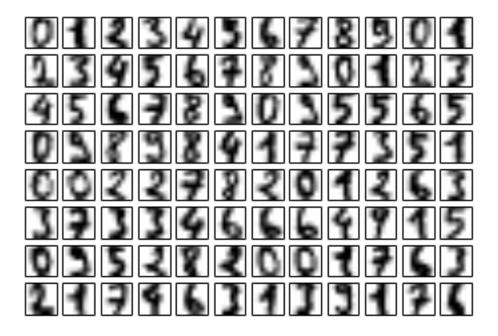
Presentation media:

iPython 20
Keynote 1
Google 1
PowerPoint 0

0.5 Notable talks

scikit-learn: Machine learning at U of Washington (Jake Vanderplas)

```
In [15]: from sklearn import datasets
         digits = datasets.load_digits()
         print digits.DESCR[:52]
         print digits.data.shape
         Optical Recognition of Handwritten Digits Data Set
         (1797, 64)
In [16]: print digits.target[0]
         print digits.data[0]
         0
         [
           0.
                 0.
                      5.
                          13.
                                 9.
                                           0.
                                                 0.
                                                      0.
                                                           0.
                                                               13.
                                                                     15.
                                                                          10.
                                                                               15.
                                      1.
         5.
                 0.
                          15.
                                                                     12.
            0.
                      3.
                                 2.
                                      0.
                                          11.
                                                 8.
                                                      0.
                                                           0.
                                                                 4.
                                                                           0.
                                                                                 0.
         8.
                           5.
                                 8.
                                                 9.
                                                           0.
                                                                 0.
                                      0.
            8.
                 0.
                      0.
                                           0.
                                                      8.
                                                                      4.
                                                                          11.
                                                                                 0.
         1.
           12.
                 7.
                      0.
                           0.
                                 2. 14.
                                           5. 10. 12.
                                                           0.
                                                                 0.
                                                                      0.
                                                                           0.
                                                                                 6.
         13.
           10.
                 0.
                      0.
                           0.1
In [17]: for i in range(8):
             print ["%2.0f" % d for d in digits.data[0][8*i:8*i+8]]
         [' 0', ' 0', ' 5', '13', ' 9', ' 1', ' 0', ' 0']
         [' 0', ' 0', '13', '15', '10', '15', ' 5', ' 0']
         [' 0', ' 3', '15', ' 2', ' 0', '11', ' 8', ' 0']
         [' 0', ' 4', '12', ' 0', ' 0', ' 8', ' 8', '
         [' 0', ' 5', ' 8', ' 0', ' 0', ' 9', ' 8', ' 0']
         ['0', '4', '11', '0', '1', '12', '7', '0']
         [' 0', ' 2', '14', ' 5', '10', '12', ' 0', ' 0']
         ['0', '0', '6', '13', '10', '0', '0', '0']
In [18]: %matplotlib inline
         import numpy as np
         import matplotlib.pyplot as plt
         fig, ax = plt.subplots(8, 12, subplot_kw={'xticks':[], 'yticks':[]})
         for i in range(ax.size):
             ax.flat[i].imshow(digits.data[i].reshape(8, 8), cmap=plt.cm.binary)
```



```
In [19]: from sklearn.naive_bayes import GaussianNB
    X = digits.data
    y = digits.target

# Instantiate the estimator
    clf = GaussianNB()

# Fit the estimator to the data, leaving out the last five samples
    clf.fit(X[:-15], y[:-15])

# Use the model to predict the last five labels
    y_pred = clf.predict(X[-15:])

print y_pred
print y[-15:]
[2 8 5 7 9 5 4 8 1 4 9 0 8 9 8]
[2 2 5 7 9 5 4 8 8 4 9 0 8 9 8]
```

pymc3: Bayesian data analysis (Thomas Wiecki, Quantopian & Brown University)

- scikit learn is cool, but models are black boxes
- not good at conveying what they have learned
- probabilistic programming: black box inference engine, but open box models
- pymc3
 - good at linear regression, bayesian inference, state of the art MCMC methods
 - good at speed: compiles to C just-in-time using Theano
 - not good at usability

```
In [163]: print "sorry, no easy example to show :("
```

Packaging and distributing with Anaconda (Travis Oliphant, Continuum Analytics)

- conda = a cross-platform system-level language-independent virtualenv
 - good scientific packages support currently, not all packages
 - user installable (don't need root)
- · conda package
- · conda build
- wakari = conda + ipython-notebook

```
In [164]: !~/anaconda/bin/conda info -e
       # conda environments:
       #
                         /Users/mpavlov/anaconda/envs/py3
      руЗ
       root
                       * /Users/mpavlov/anaconda
In [165]: !~/anaconda/bin/conda remove --name py3 --all --yes
       !~/anaconda/bin/conda info -e
      Package plan for package removal in environment
       /Users/mpavlov/anaconda/envs/py3:
       The following packages will be UN-linked:
          package
                                        build
          _____I___I____I
          distribute-0.6.45
                                       py33_1
                            ·
          pip-1.4.1
                                       py33_0
          python-3.3.2
                             1
          readline-6.2
                                           1
          sqlite-3.7.13
                              tk-8.5.13
                                           1
          zlib-1.2.7
       Unlinking packages ...
      ] || 28%
```

```
In [166]: !~/anaconda/bin/conda create --name py3 python=3.3 pip --yes
```

* /Users/mpavlov/anaconda

100% [COMPLETE] || 100%

conda environments:

root Package plan for creating environment at /Users/mpavlov/anaconda/envs/py3:

The following packages will be linked:

package		build	
distribute-0.6.45	 	py33_1	hard-link
pip-1.4.1		py33_0	hard-link
python-3.3.2		1	hard-link
readline-6.2		1	hard-link
sqlite-3.7.13		1	hard-link
tk-8.5.13		1	hard-link
zlib-1.2.7	I	1	hard-link
] 14% [p: 42% [readline] 85% [:] 100%	_] 28%] 57% [sqlite]
<pre># To activate this environ # \$ source activate py3 #</pre>	onment, use:		
<pre># To deactivate this env: # \$ source deactivate #</pre>	ronment, use:		

In [167]: source ~/anaconda/bin/activate py3 && which python

prepending /Users/mpavlov/anaconda/envs/py3/bin to PATH
/Users/mpavlov/anaconda/envs/py3/bin/python

```
In [168]: !~/anaconda/bin/conda list --name py3
```

```
# packages in environment at /Users/mpavlov/anaconda/envs/py3:
distribute
                          0.6.45
                                                    py33_1
pip
                          1.4.1
                                                    py33_0
python
                          3.3.2
                                                         1
                          6.2
readline
                                                         1
                          0.6c11
setuptools
                                                     <pip>
                          3.7.13
sqlite
                                                         1
                          8.5.13
                                                         1
tk
zlib
                          1.2.7
                                                         1
```

- tried running disco_profiling_task_status_service on python 3.3 in a conda env
 - while pyramid is python 3.2 compatible, pyramid_whoauth has invalid syntax (except Exception, e)
 - inconsistencies between tools: conda install, pip, easy_install
 - needs more work

ddpy: Data-driven music for big data analysis (Thomas Levine, CSV Soundsystem)

- visualizations alone can only show you 2D, if you're skilled you can do 5-6 dimensions at most
- music can show you dozens
- example: http://fms.csvsoundsystem.com/

```
In [169]: print "The Future: data gastronomification can use all 5 senses"
Image("https://github.com/csv/ddpy/raw/master/img/artichoke.jpg")
```

The Future: data gastronomification can use all 5 senses

Out [169]:

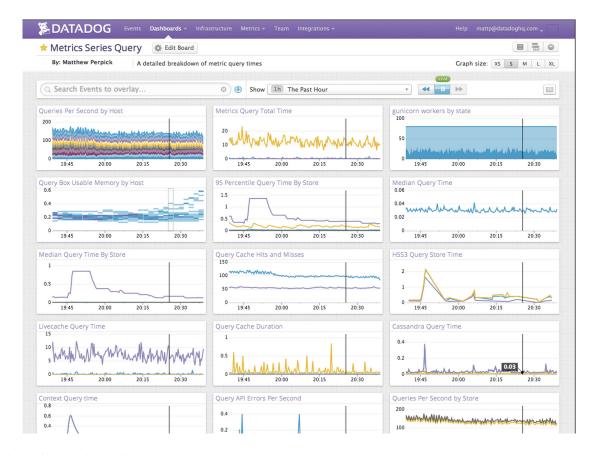


Python @ Datadog: Building High-Volume Data Systems in the Python Ecosystem (Matt Perpick, Datadog)

• dashboards, metrics, alerting

```
In [170]: Image("img/datadog.png")
```

Out [170]:



- billions of data points daily, low latency
- how to eliminate Python overhead (for the code sections that matter)
 - use c bindings, skip dynamic types, interpreter
 - cython is your friend: compiles to c, reads like python
- concurrency soup datadog has tried everything and it all sucks; no miracle solutions

Keynote: iPython, attributes of software, and our work (Brian Granger, Cal Poly State University, IPython)

- the tools we use affect our behavior
- · our behavior affects our work
- tools' attributes get inherited, through our behavior, by our work output
- · ipython's attributes
 - useful in multiple contexts
 - * Individual, interactive exploration, Debugging, testing, Production runs, Parallel computing, Collaboration, Publication, Presentation, Teaching/Learning
 - close to data
 - * visualize, interact, compute, repeat
 - open
 - * everything is public: code, data formats, issues, roadmap, chatroom, google hangout meetings broadcasted live, recorded and available online
 - multilingual
 - * supported: ruby, bash, julia, many others
 - * not yet: R (wanted!)

- also supports
 - latex equations
 - audio
 - video
 - inline html
 - iframes
 - interactive plots
 - partial example: http://nbviewer.ipython.org/url/github.com/ipython/ipython/raw/master/examples/notebooks/Part%204%20
 %20Markdown%20Cells.ipynb

Lightning talk: prettyplotlib

- matplotlib wrapper
- sensible defaults
- see https://github.com/olgabot/prettyplotlib

0.6 Summary of cool tools

- ipython and nbviewer.ipython.org
- anaconda
- python 3
- pandas, numpy, cython
- scikit-learn and pymc3
- prettyplotlib

the end