

EECS 510: SOCIAL MEDIA MINING  
SPRING 2016

# **Data Mining Essentials 2: Data Mining in Practice, with Python**

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# Outline

- Why Python?
- Intro to Python
- Intro to Scikit-Learn
- Unsupervised Learning
  - Demo on PCA, K-Means
- Supervised Learning
  - Demo on Linear Regression, Logistic Regression



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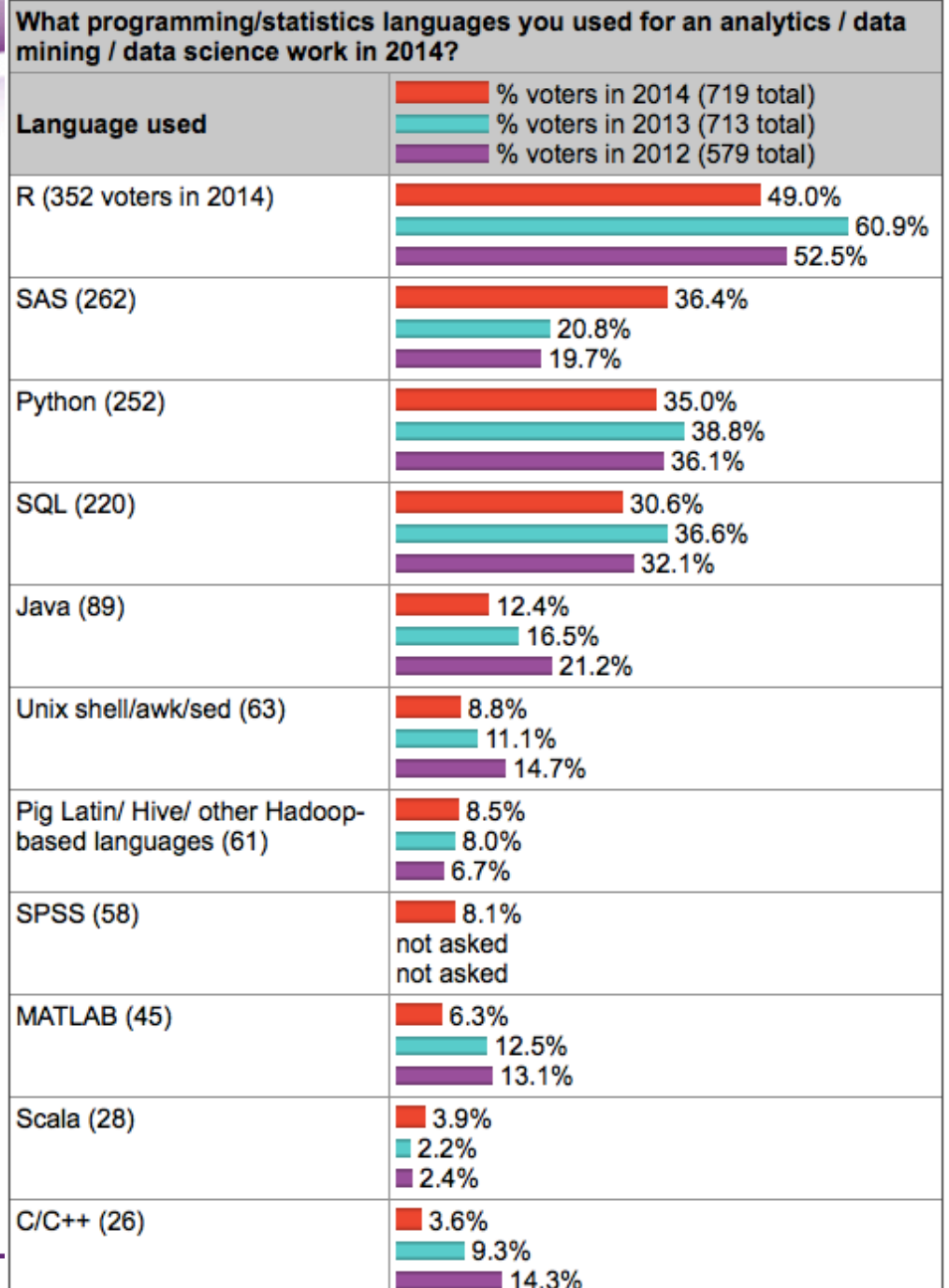
- **Why Python?**

**What programming language  
do you use for data mining?**



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Source from: <http://www.kdnuggets.com/polls/index.html>



How much is your salary as analytics, data mining, data science professionals?



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Analytic Role	Salary or Income
Manage teams which analyze data (18%)	\$141K
Data Scientist/Data Miner (47%)	\$118K
Data Analyst/Business Analyst (support data analysis) (22%)	\$70K
other role (6.5%)	\$73K
Academic Researcher (4.3%)	\$80K
Student (1.7%)	\$26K

Region	Employer Type	Salary or Income
US/Canada (154)	Company/Self	\$128K
	Academic/Gov/Non-profit	\$86K
Europe (43)	Company/Self	\$82K
	Academic/Gov/Non-profit	\$35K
Asia (14)	Company/Self	\$59K
	Academic/Gov/Non-profit	\$40K
Australia/NZ (9)	Company/Self	\$90K
	Academic/Gov/Non-profit	\$105K
Other (6)	Company/Self	\$75K
	Academic/Gov/Non-profit	\$88K



**Should data scientist / data miners be responsible for their predictions?**

Should data scientists / data miners be responsible for their predictions? [242 voters]	
No, they should not be responsible (108)	45%
Not sure (32)	13%
They can be held financially responsible, but if they also benefit from correct predictions (89)	37%
They can be held criminally responsible for wrong predictions (13)	5%



# Why Python?

- **Why Python?**

Not

## **Think about the scientist's needs:**

- Get data (simulation, experiment control)
- Manipulate and process data.
- Visualize results... to understand what we are doing!
- Communicate results: produce figures for reports or publications, write presentations.



# Why Python?

- **Why Python?**

Not

- Easy

- Easy to learn, easily readable
- Scientists first, programmers second

- Efficient

- Managing memory is easy – if you just don't care

- A single Language for everything

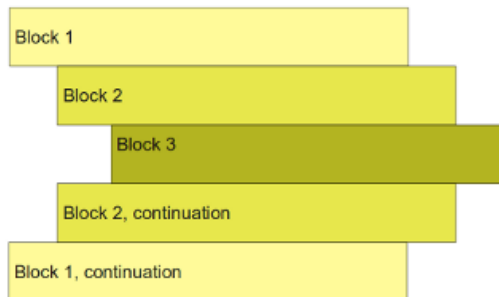
- Avoid learning a new software for each new problem





## More to Take Away

- Free distribution from <http://www.python.org>
- Known for it's "batteries included" philosophy  
Similar to R, Python has a fantastic community around it and, luckily for you, this community can write
- Two popular versions, 2.7 or 3.x
- A single-click installer: Enthought Canopy
- Prepare yourself for code indentation heaven



```
from math import sqrt
n = input("Maximal Number? ")
n = int(n)+1
for a in range(1,n):
    for b in range(a,n):
        c_square = a**2 + b**2
        c = int(sqrt(c_square))
        if ((c_square - c**2) == 0):
            print(a, b, c)
```



# All the Good Modules

- **numpy, scipy**: basics for almost everything
- **Matplotlib**, a Python 2D plotting library <http://matplotlib.org>
- **NLTK**, Natural Language Toolkit <http://www.nltk.org>
- **Pandas**, Python Data Analysis Library <http://pandas.pydata.org>
- **mrjob**, route to writing MapReduce jobs <https://pythonhosted.org/mrjob/>
- **IPython**, Interactive console with IDE-like features <http://ipython.org>
- **Scikit-Learn**, ML resource and library <http://scikit-learn.org/dev/index.html>
- **Theano/Pylearn2**, deep learning  
<http://deeplearning.net/software/theano/>  
<http://deeplearning.net/software/pylearn2/>
- **More**: mlpy, PyBrain, Orange, Scrapy, ...



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# The Use of Python: Simple demos

0 – Python Intro.ipynb



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# What is Scikit-learn

- A Python Machine Learning Library
- Focused on modeling data
- Developed by David Cournapeau as a Google summer of code project in 2007.
- First public release (v0.1 beta) published in late January 2010.
- Now has more than 30 active contributors and has had paid sponsorship from INRIA, Google, Tinyclues and the Python Software Foundation.
- The library is built upon the SciPy that must be installed before you can use scikit-learn.



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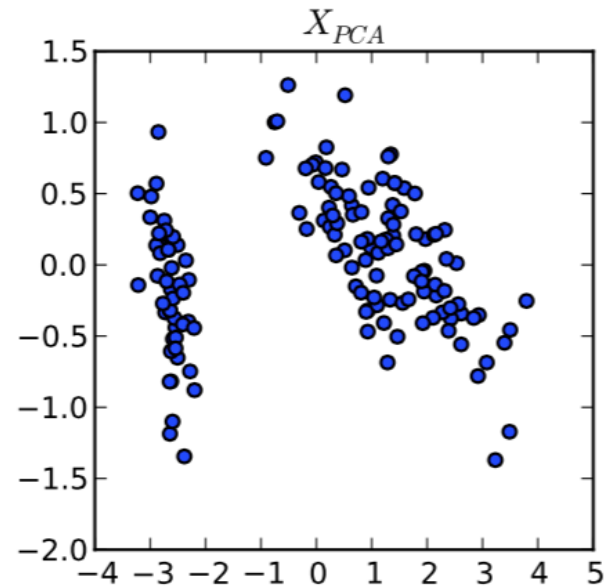
# The use of Scikit-Learn: unsupervised learning demos



# PCA Summary

- PCA projects to axis with greatest variance
- Often provides good first insight into dataset

$$\begin{aligned}\bar{X} &\leftarrow X - \text{mean}(X) & \bar{X} &\in \mathbb{R}^{n \times N} \\ W &\leftarrow \text{PCA}(\bar{X}, 2) & W &\in \mathbb{R}^{N \times M} \\ X_{\text{PCA}} &\leftarrow \bar{X} \cdot W & X_{\text{PCA}} &\in \mathbb{R}^{n \times M}\end{aligned}$$



- Identify important variables in projection matrix  $W$ :

$$W = \begin{bmatrix} 0.36 & -0.08 & 0.85 & 0.35 \\ -0.65 & -0.72 & 0.17 & 0.07 \end{bmatrix}$$



1 – PCA.ipynb





# K-Means Algorithm

$k$ -Means finds assignments  $j$  and cluster centers  $\mu$  by solving

$$\min_{\mu} \sum_{i=0}^N \min_j \|\mu_j - x_i\|^2 \quad (1)$$

The algorithm is simple:

1. Set  $\mu, j$  to a random value
2. Solve (1) for  $j$
3. Solve (1) for  $\mu$
4. If  $j$  or  $\mu$  changed significantly, go to step 2.



2 – k means.ipynb



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  - Demo on Linear Regression, Logistic Regression, kNN

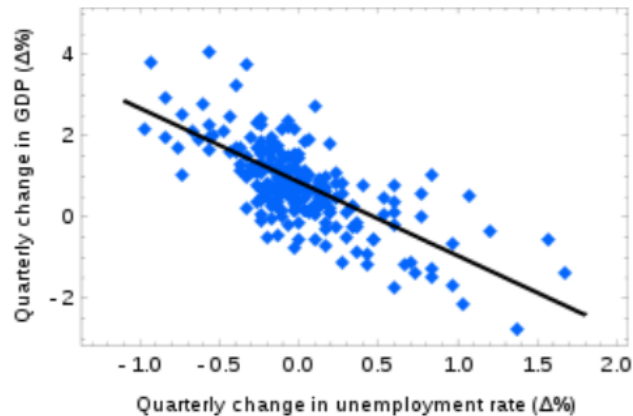


# The use of Scikit-Learn: supervised learning demos



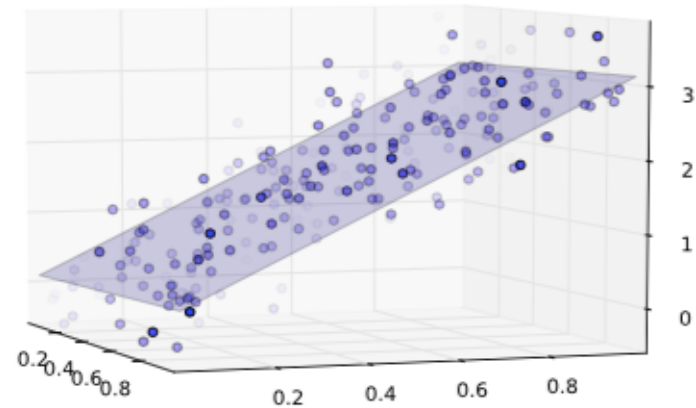
# Linear Regression

$$y = w_1x_1 + b$$



1D

$$y = w_2x_2 + w_1x_1 + b$$



2D

To find  $w$  and  $b$ , minimize the error:

$$E = \sum_{i=0}^N (y_i - (w_i x_i + b))^2$$





3 – LinearRegression1.ipynb

3 – LinearRegression2.ipynb



# Logistic Regression

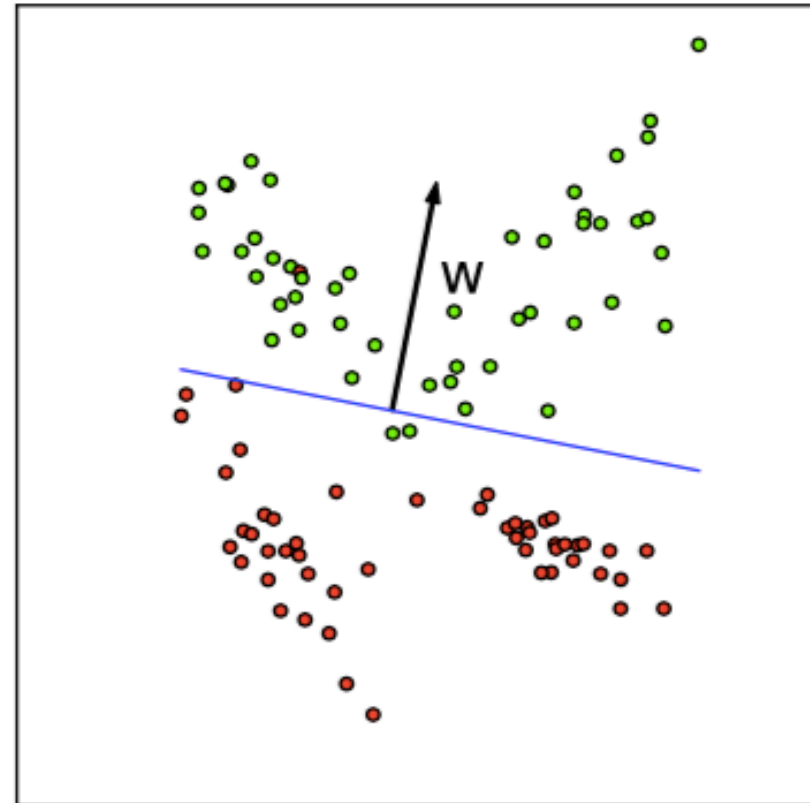
For two classes  $-1, +1$ .

Decision boundary given by hyperplane.

Hyperplane defined by normal vector and offset:

$$y = \text{sign}(\langle w, x \rangle + b)$$

$$w \in \mathbb{R}^n, b \in \mathbb{R}$$



# Logistic Regression

Relation to regression:

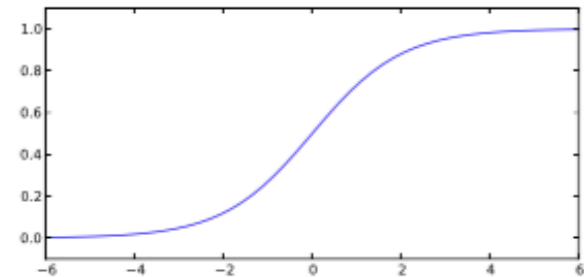
$$p(y = +1 | x) = \text{logistic}(\langle w, x \rangle + b)$$

As probabilities are between 0 and 1, the logistic function squashes the regression result:

$$p(y = +1 | x) > 0.5 \Leftrightarrow \langle w, x \rangle + b > 0$$

Need to solve:

$$\max_w \sum_{i=0}^n \log(p(Y = y_i | x_i))$$

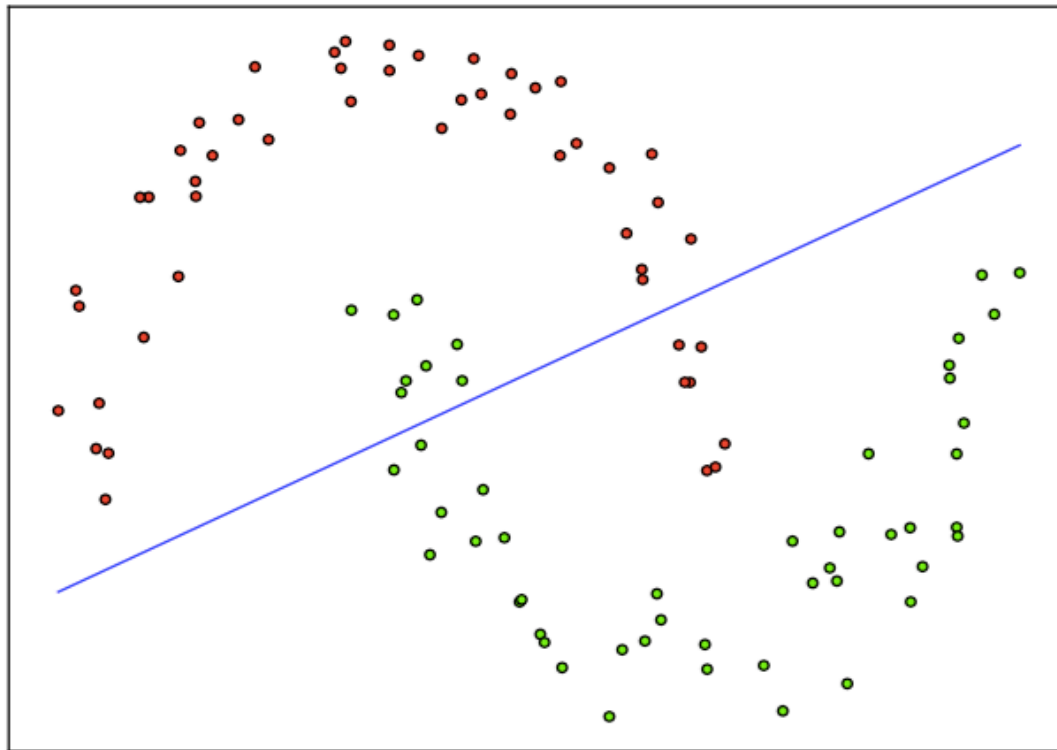


## 4 – LogisticRegression.ipynb



# Nonlinear Problems

- Logistic regression works well if the data is linearly separable, but...





# K Nearest Neighbors

- Classification: same setup as logistic regression.
- Very simple but powerful idea: Do as your neighbors do.
- For a new point  $x$  look at the nearest (or the two nearest or three nearest, ...) point(s) in the training data for a label.
- Usual distance measure: Euclidean distance



# Simple Algorithm

- Pick a  $k$ , for example  $k = 3$ .
- Want to classify new example  $x$ .
- Compute  $d_i = d(x_i, x)$ , i.e.  $d(x_i, x) = ||x_i - x||$ .
- Sort  $d_i$ , take  $k$  smallest:  $d_{i_0}, d_{i_1}, d_{i_2}$ .
- Assign  $y$  that appears most often among  $y_{i_0}, y_{i_1}, y_{i_2}$ .



5 – kNN.ipynb

