

Big Data: Data Wrangling Boot Camp

Python Sentiment Analysis

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28 January 2017

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What are we going to cover?

- Look to the future
- Talk briefly about sentiment analysis
- Address the polyglot of computer languages
- Talk about our sentiment analysis system
- **Data wrangle** tweets using Python

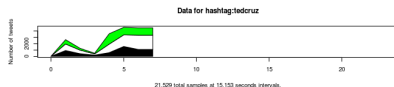
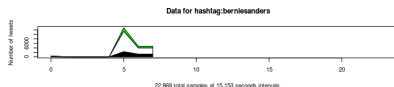
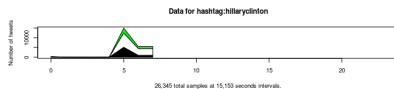
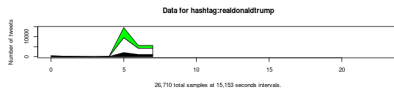




Things that will be happening today

Things that we will be doing.

- 1 **Data wrangle** tweets using python
- 2 Conduct sentiment analysis on tweets
- 3 Look at the sentiments in different ways

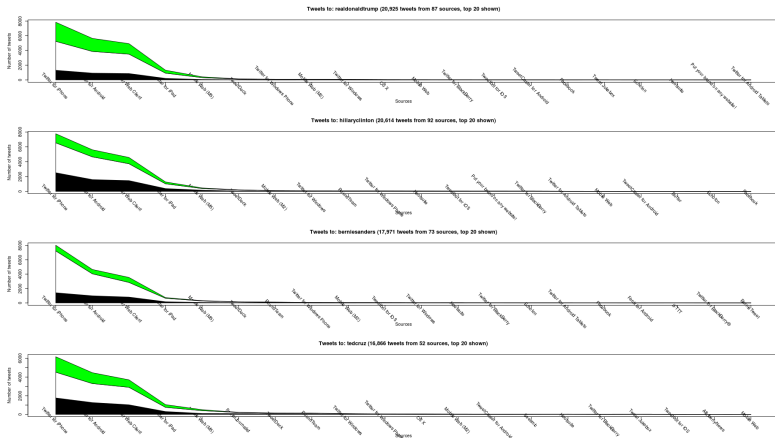


Sentiments over time.



Things that will be happening today

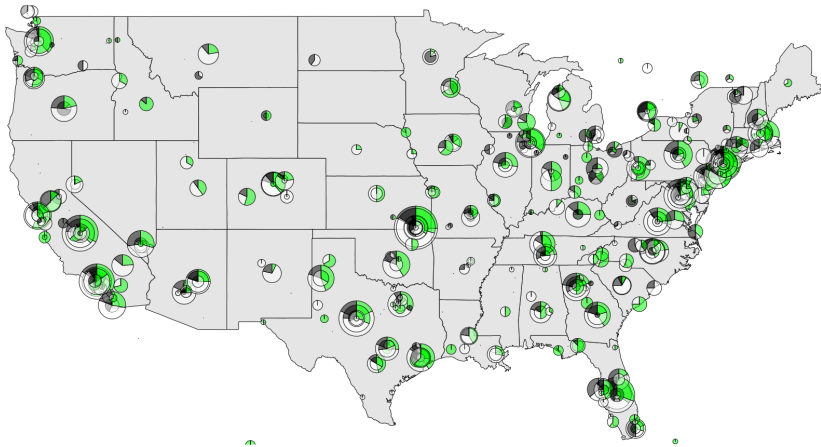
Sentiment by sending device

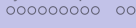




Things that will be happening today

Sentiment by geographic location



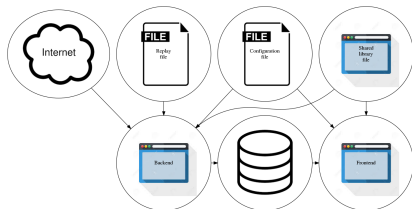


How we'll get to the images

We'll walk before we run.

- Start with a replay file
- **Data wrangle** using the library file
- Go live and download live tweets

Data flows through the backend, into the database, out the frontend.

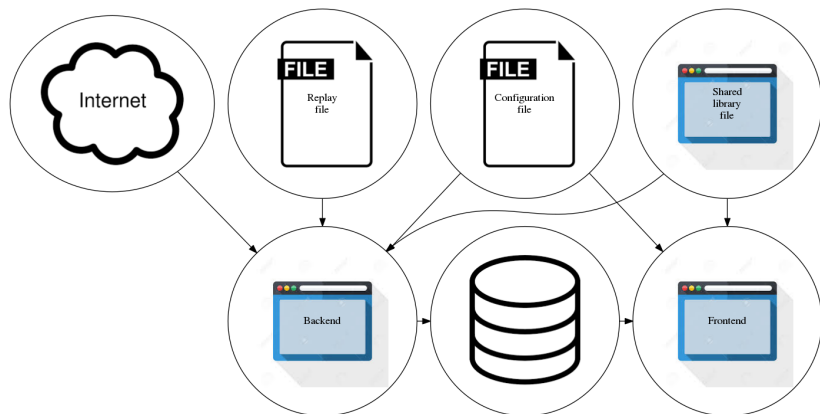


The software design document (attached) contains lots of details.



How we'll get there

Same image.



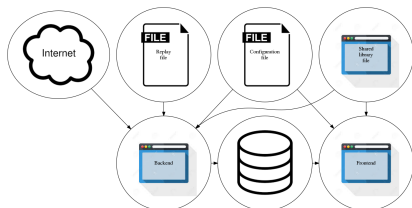
The software design document (attached) contains lots of details.



How we'll get there

What some of the files do:

- Replay file – previously recorded tweets
- Configuration file – directives used by both backend and frontend script files
- Shared library file – routines that are common to backend and frontend script files
- Backend file – script file that populates the database from the replay file, or the Internet
- Frontend file – script file that extracts data from the database and presents results





What is it, and why should I care?

A working definition

“Sentiment analysis (also known as opinion mining) refers to the use of natural language processing, text analysis and computational linguistics to identify and extract subjective information in source materials. Sentiment analysis is widely applied to reviews and social media for a variety of applications, ranging from marketing to customer service.”

W. Staff [3]



More formal definitions

“The field of opinion mining and sentiment analysis is well-suited to various types of intelligence applications. Indeed, business intelligence seems to be one of the main factors behind corporate interest in the field.”

Pang and Lee [2]

“Sentiment analysis, also called opinion mining, is the field of study that analyzes peoples opinions, sentiments, evaluations, appraisals, attitudes, and emotions towards entities such as products, services, organizations, individuals, issues, events, topics, and their attributes.”

Liu [1]

Our approach to sentiment analysis

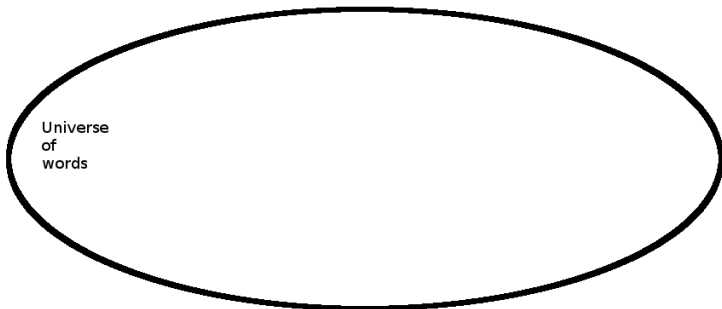
We will:

- 1 Search the “twitterverse” for tweets using specific hashtags
- 2 Tokenize each tweet
- 3 **Data wrangle** each token
- 4 Remove all stop words from the tokens
- 5 Count number of positive and negative tokens
- 6 Compute the positive, negative, or neutral sentiment for the tokens
- 7 Display the results

Our approach is language agnostic.

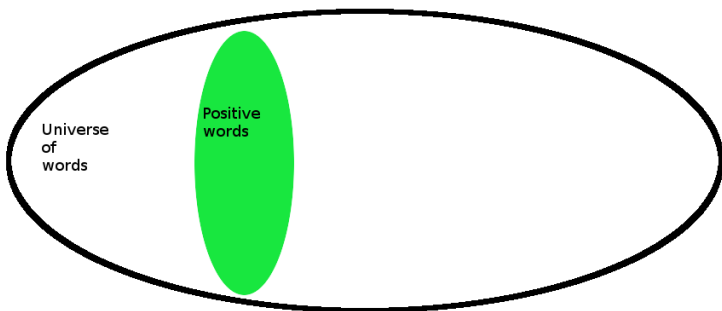


A “Universe” of words



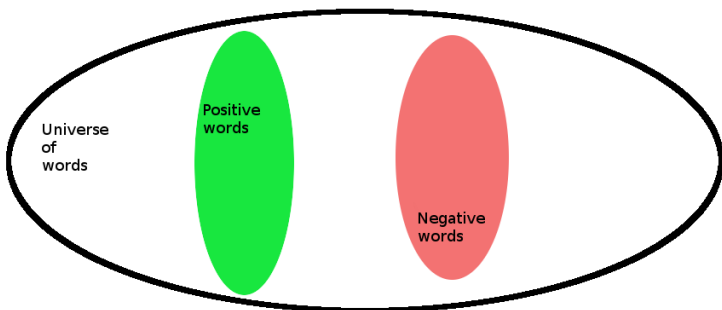


Some words are “Positive”





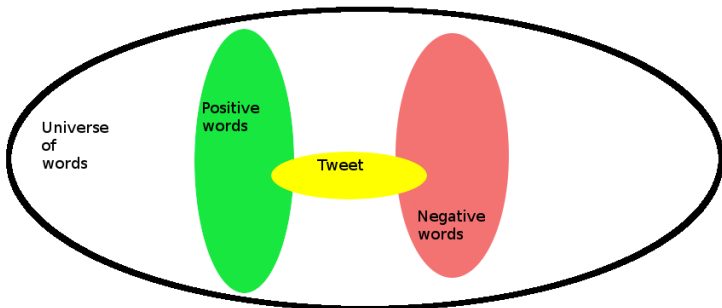
Some words are “Negative”





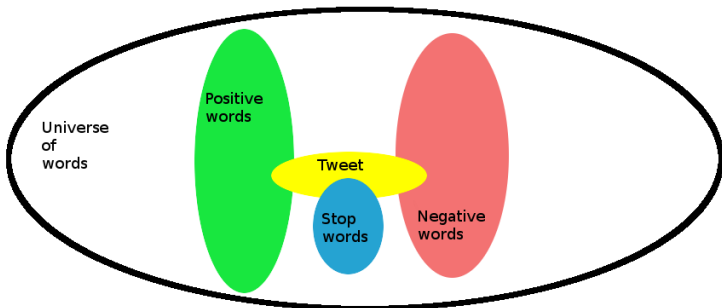
A visualization

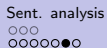
A tweet will/may have positive and negative words





Some words we don't care about

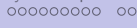




Mechanically this is what we are doing

The steps are:

- 1 Break the tweet into tokens
- 2 Remove stop words from the tokens
- 3 Compute the percentage of remaining tweet tokens that are positive
- 4 Compute the percentage of remaining tweet tokens that are negative
- 5 Classify the tokens as positive, negative, or neutral



Mathematically this is what we are doing

The steps are:

$$tokens = \{\text{words in tweet}\}$$

$$tokensLessStop = tokens - stopWords$$

$$positivePart = positiveWords \cap tokensLessStop$$

$$negativePart = negativeWords \cap tokensLessStop$$

$$classification = \begin{cases} \text{positive,} & \text{if } \frac{positivePart}{tokensLessStop} \leq \text{positiveThreshold} \\ & \text{AND} \\ & \frac{negativePart}{tokensLessStop} < \text{negativeThreshold} \\ \text{negative,} & \text{if } \frac{negativePart}{tokensLessStop} \leq \text{negativeThreshold} \\ & \text{AND} \\ & \frac{positivePart}{tokensLessStop} < \text{positiveThreshold} \\ \text{neutral,} & \text{otherwise} \end{cases}$$

Comparing a known language with Python and R

A specific language used was not a requirement for this boot camp. Python and R used in boot-camp. Too many languages to compare to Python and R.



<http://hyperpolyglot.org/>

Use Hyperpolyglot.org as a cross reference.

A software design document

The document contains:

- ① Overall system design
- ② Algorithms used through out the system
- ③ Details about the configuration file
- ④ Details about the database tables

softwareOverview.pdf - Adobe Reader
File Edit View Document Tools Window Help

softwareOverview... [X]

ODU Big Data, Data Wrangling Boot Camp
Software Overview and Design
Chuck Carlidge
August 11, 2016

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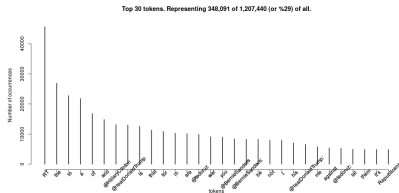
1

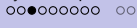
The file is attached.



What will happen to the tweets when we do nothing.

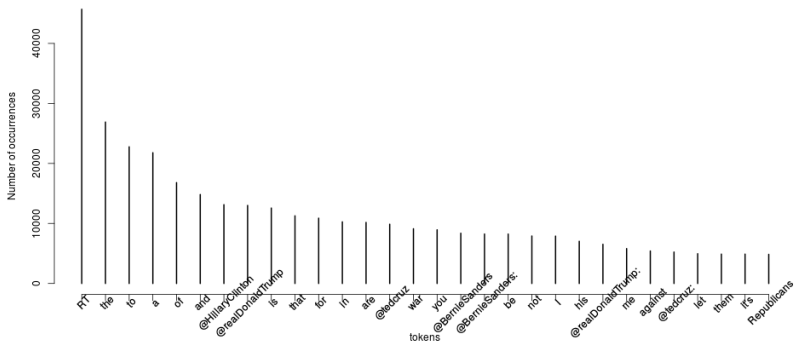
A histogram of how often a token occurs.

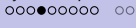




Same image.

Top 30 tokens. Representing 348,091 of 1,207,440 (or %29) of all.



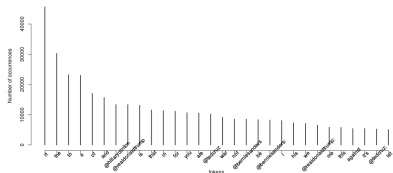


And implementation

What will happen to the tweets when we make everything the same case

Changing case is easy, unless they are emojis.

Top 30 tokens. Representing 365,040 of 1,207,440 (or %30) of all.

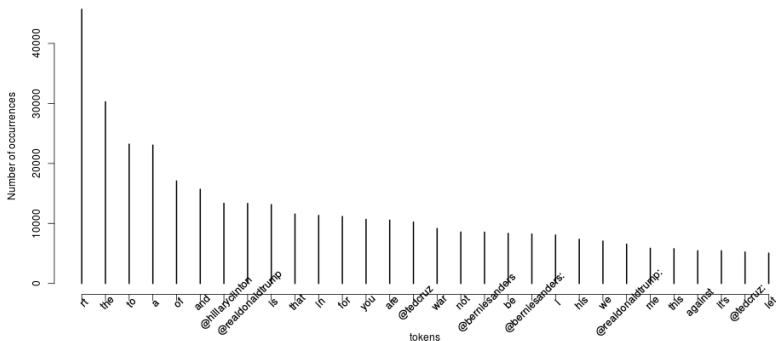




And implementation

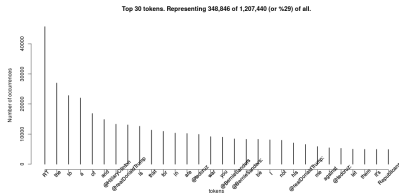
Same image.

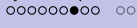
Top 30 tokens. Representing 365,040 of 1,207,440 (or %30) of all.



What will happen to the tweets we remove non-ASCII

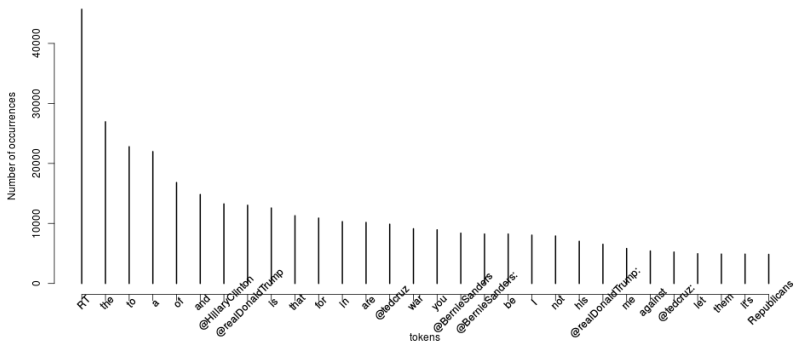
We need to talk about ASCII and the rest of the character systems.





Same image.

Top 30 tokens. Representing 348,846 of 1,207,440 (or %29) of all.

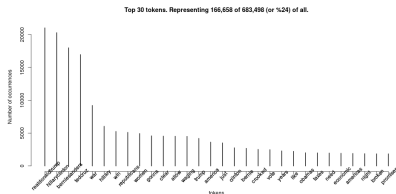


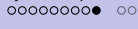


What will happen to the tweets when we remove “stop words”

Stop words are words/tokens that have no use in whatever we are doing. Stop words are domain specific.

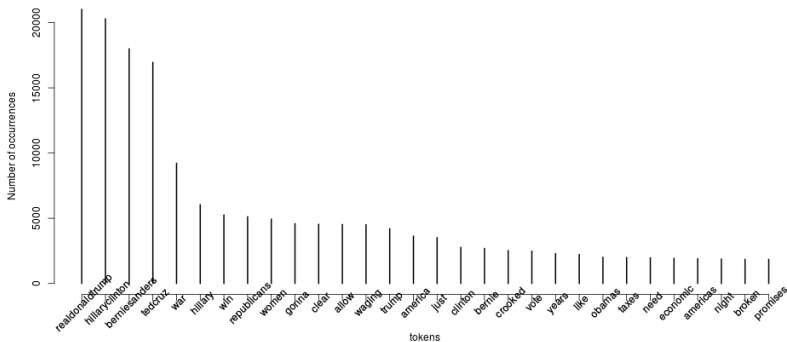
And we change case, remove non-ASCII, remove punctuation, etc.





Same image.

Top 30 tokens. Representing 166,658 of 683,498 (or %24) of all.

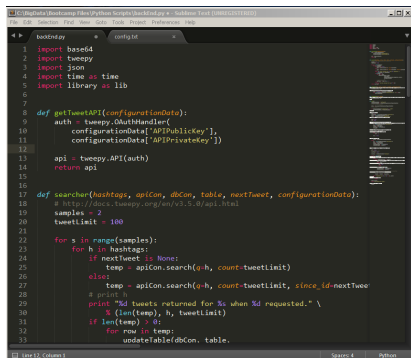


A complete IDE

A complete Python integrated development environment (IDE)

- 1 Text editor
- 2 Python console
- 3 Extensible
- 4 Tabbed display for multiple files

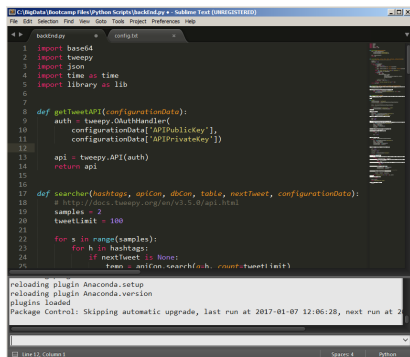
We incorporate the Anaconda plugin for additional functionality.



```
1 import base64
2 import tweepy
3 import json
4 import time as time
5 import library as lib
6
7
8 def getTweetAPI(configurationData):
9     auth = tweepy.OAuthHandler(
10         configurationData['APIPublicKey'],
11         configurationData['APIPrivateKey'])
12
13     api = tweepy.API(auth)
14     return api
15
16
17 def searcher(hashtags, apiCon, dbCon, table, nextTweet, configurationData):
18     # http://docs.tweepy.org/en/v3.5.0/api.html
19     samples = 2
20     tweetlimit = 100
21
22     for s in range(samples):
23         for h in hashtags:
24             if nextTweet is None:
25                 temp = apiCon.search(q=h, count=tweetlimit)
26             else:
27                 temp = apiCon.search(q=h, count=tweetlimit, since_id=nextTweet)
28             # print h
29             print "%d tweets returned for %s when %d requested." %
30                   % (len(temp), h, tweetlimit)
31             if len(temp) > 0:
32                 for row in temp:
33                     updateTable(dbCon, table,
```

Editor

- “Smart” editor
- CTRL + O to open a file
- CTRL + S to save a file
- CTRL + B to run a file
- Multiple files can be opened at once



```
File Edit Selection Find View Goto Tools Project Preferences Help
backEnd.py config.txt
1 import base64
2 import tweepy
3 import json
4 import time as time
5 import library as lib
6
7
8 def getTweepyAPI(configurationData):
9     auth = tweepy.OAuthHandler(
10         configurationData['APIPublicKey'],
11         configurationData['APIPrivateKey'])
12
13     api = tweepy.API(auth)
14     return api
15
16
17 def searchar(hashtags, apiCon, dbCon, table, nextTweet, configurationData):
18     # http://docs.tweepy.org/en/v3.5.0/api.html
19     samples = 2
20     tweettlimit = 100
21
22     for s in range(samples):
23         for h in hashtags:
24             if nextTweet is None:
25                 tweet = apiCon.search(auth, count=tweettlimit)
26
reloading plugin Anaconda.setup
reloading plugin Anaconda.version
plugins loaded
Package Control: Skipping automatic upgrade, last run at 2017-01-07 12:06:28, next run at 2
Line 12, Column 1 Spaces: 4 Python
```

The Anaconda plugin enhances the editor, and adds “lint” and style functions.

Q & A time.

Q: Do you know what the death rate around here is?

A: One per person.



What have we covered?

- A preview of today's activities
- An overview of the sublime IDE



Next: Hands on analysing tweets with Python.

References I

- [1] Bing Liu, [Sentiment analysis and opinion mining](#), 2012.
- [2] Bo Pang and Lillian Lee, [Opinion mining and sentiment analysis](#), 2008.
- [3] Wikipedia Staff, [Sentiment analysis](#),
https://en.wikipedia.org/wiki/Sentiment_analysis,
2016.

Files of interest



Software design

document

