Twitter Sentiment Analysis

Due Apr 5, 2017 by 11:59pm **Points** 30 **Submitting** a file upload **Available** Mar 28, 2017 at 12am - Jun 2, 2017 at 11:59pm 2 months

This assignment was locked Jun 2, 2017 at 11:59pm.

Twitter represents a fundamentally new instrument to make social measurements. Millions of people voluntarily express opinions across any topic imaginable --- this data source is incredibly valuable for both research and business.

For example, researchers have shown that the "mood" of communication on twitter reflects biological rhythms (http://www.nytimes.com/2011/09/30/science/30twitter.html) and can even used to predict the stock market (http://arxiv.org/pdf/1010.3003&embedded=true). A student here at UW used geocoded tweets to plot a map of locations where "thunder" was mentioned in the context of a storm system in Summer 2012. (http://cliffmass.blogspot.com/2012/07/thunderstorm-fest.html)

Researchers from Northeastern University and Harvard University studying the characteristics and dynamics of Twitter have an excellent resource (http://www.ccs.neu.edu/home/amislove/twittermood/) for learning more about how Twitter can be used to analyze moods at national scale.

In this assignment, you will

- · access the twitter Application Programming Interface (API) using python
- estimate the public's perception (the sentiment) of a particular term or phrase
- analyze the relationship between location and mood based on a sample of twitter data

Some points to keep in mind:

- This assignment is open-ended in several ways. You'll need to make some decisions about how best to solve the problem and implement them carefully.
- · It is perfectly acceptable to discuss your solution on the forum, but don't share code.
- Each student must submit their own solution to the problem.
- You will have an unlimited number of tries for each submission.
- Your code will be run in a protected environment, so you should only use the Python standard libraries unless you are specifically instructed otherwise. Your code should also not rely on any external libraries or web services.

The Twitter Application Programming Interface

Twitter provides a very rich REST API for querying the system, accessing data, and control your account. You can read more about the Twitter API (https://dev.twitter.com/docs)

Python 2.7 environment

If you are new to Python, you may find it valuable to work through the **codeacademy Python tutorials** (http://www.codecademy.com/tracks/python). Focus on tutorials 1-9, plus tutorial 12 on File IO. In addition, many students have recommended **Google's Python class** (https://developers.google.com/edu/python/).

You will need to establish a Python programming environment to complete this assignment. You can install Python yourself by downloading it from the Python website (http://www.python.org/download/), or can use the class virtual machine (https://spark-public.s3.amazonaws.com/datasci/Coursera-Data-Science-Ubuntu.ova).

Unicode strings

Strings in the twitter data prefixed with the letter "u" are unicode strings. For example:

u"This is a string"

Unicode is a standard for representing a much larger variety of characters beyond the roman alphabet (greek, russian, mathematical symbols, logograms from non-phonetic writing systems such as kanji, etc.)

In most circumstances, you will be able to use a unicode object just like a string.

If you encounter an error involving printing unicode, you can use the **encode** (http://docs.python.org/2/library/stdtypes.html#str.encode) method to properly print the international characters, like this:

unicode_string = u"aaaàçççñññ" encoded_string = unicode_string.encode('utf-8') print encoded_string

Once again: If you are new to Python, many students have recommended Google's Python class. (https://developers.google.com/edu/python/)

If you use Python 3, use this version of the starter file: twitterstream_py3.py

Problem 1: Get Twitter Data

The first step is to download the materials for the assignment.

Important: For this assignment, we hope to have you work with the live Twitter stream using the development API. However, as of 2014, Twitter requires you to attach a mobile phone number to your account in order to use the developer API. If you do not have mobile phone, or if your carrier is not supported, you may find this workaround

 $(http://meberhard.me/workaround-twitter-application-write-access-mobile-number-accepted-twitter-website/) \ \ \textbf{useful.}$

However, if you cannot get access to Twitter data due to problems with their login verification protocol, you can instead use the sample dataset

(https://raw.githubusercontent.com/uwescience/datasci_course_materials/master/assignment1/three_minutes_tweets.json) in the course github repository called three minute tweets.json.

To access the live stream, you will need to install the oauth2 library (http://pypi.python.org/pypi/oauth2/) so you can properly authenticate.

The command

\$ pip install oauth2

should work for most environments.

If you are on Windows, before you can run the command above, you may need to install pip. You can watch a video on installing pip on Windows (https://www.youtube.com/watch?v=zPMr0IEMgpo)



(https://www.youtube.com/watch?v=zPMr0IEMqpo)

The steps below will help you set up your twitter account to be able to access the live 1% stream.

- 1. Create a twitter account if you do not already have one.
- 2. Go to https://dev.twitter.com/apps (https://dev.twitter.com/apps) and log in with your twitter credentials.
- 3. Click "Create New App"
- 4. Fill out the form and agree to the terms. Put in a dummy website if you don't have one you want to use.
- 5. At this point you will be prompted to attach a mobile phone number to your account (https://support.twitter.com/articles/110250) if you have not previously done so. Follow the instructions at the link provided. If you cannot complete this protocol, you may use the sample dataset (https://raw.githubusercontent.com/uwescience/datasci_course_materials/master/assignment1/three_minutes_tweets.json) in the repository, but we encourage you to connect to Twitter if possible so you can build your own applications.
- 6. On the next page, click the "Keys and Access Tokens" tab along the top, then scroll all the way down until you see the section "Your Access Token"
- 7. Click the button "Create My Access Token". You can Read more about Oauth authorization. (https://dev.twitter.com/docs/auth)
- 8. You will now copy four values into the file twitterstream.py. These values are your "Consumer Key (API Key)", your "Consumer Secret (API Secret)", your "Access token" and your "Access token secret". All four should now be visible on the "Keys and Access Tokens" page. (You may see "Consumer Key (API Key)" referred to as either "Consumer key" or "API Key" in some places in the code or on the web; all three are synonyms.) Open twitterstream.py and set the variables corresponding to the api key, api secret, access token, and access secret. You will see code like the below:

api_key = "<Enter api key>" api_secret = "<Enter api secret>" access_token_key = "<Enter your access token key here>" access token secret = "<Enter your access token secret here>"

Now that your keys are ready, you are ready to test your access.

1. Run the following and make sure you see data flowing and that no errors occur.

```
$ python twitterstream.py > output.txt
```

- 2. This command pipes the output to a file. Stop the program with Ctrl-C, but wait at least 3 minutes for data to accumulate. Keep the file output.txt for the duration of the assignment; we will be reusing it in later problems. Don't use someone else's file!
- 3. If you wish, modify the file to use the twitter search API (https://dev.twitter.com/docs/api/1.1/get/search/tweets) to search for specific terms. For example, to search for the term "microsoft," you can pass the following url to the twitterreq function:

https://api.twitter.com/1.1/search/tweets.json?q=microsoft

What to turn in: The first 20 lines of the twitter data you downloaded from the web (or the first 20 lines from the sample file if you are unable to access the Twitter Developer API due to the requirement to use a mobile phone).

You can save the first 20 lines to a file problem_1_submission.txt by using the following command:

```
$ head -n 20 output.txt > problem_1_submission.txt
```

Problem 2: Derive the sentiment of each tweet

For this part, you will compute the sentiment of each tweet based on the sentiment scores of the terms in the tweet. The sentiment of a tweet is equivalent to the sum of the sentiment scores for each term in the tweet.

You are provided with a skeleton file tweet_sentiment.py which accepts two arguments on the command line: a *sentiment file* and a tweet file like the one you generated in Problem 1. You can run the skeleton program like this:

```
$ python tweet_sentiment.py AFINN-111.txt output.txt
```

The file AFINN-111.txt contains a list of pre-computed sentiment scores. Each line in the file contains a word or phrase followed by a sentiment score. Each word or phrase that is found in a tweet but not found in AFINN-111.txt should be given a sentiment score of 0. See the file AFINN-README.txt for more information.

To use the data in the AFINN-111.txt file, you may find it useful to build a dictionary. Note that the AFINN-111.txt file format is tab-delimited, meaning that the term and the score are separated by a tab character. A tab character can be identified a "\t".

The data in the tweet file you generated in Problem 1 is represented as *JSON*, which stands for JavaScript Object Notation. It is a simple format for representing nested structures of data --- lists of lists of dictionaries of lists of you get the idea.

Each line of output.txt represents a **streaming message** (https://dev.twitter.com/docs/streaming-apis/messages). Most, but not all, will be **tweets** (https://dev.twitter.com/docs/platform-objects/tweets). (The skeleton program will tell you how many lines are in the file.)

It is straightforward to convert a JSON string into a Python data structure; there is a library to do so called json.

To use this library, add the following to the top of tweet sentiment.py

import json

Then, to parse the data in output.txt, you want to apply the function json.loads to every line in the file.

This function will parse the json data and return a python data stucture; in this case, it returns a dictionary. If needed, take a moment to read the documentation for Python dictionaries (http://docs.python.org/2/library/stdtypes.html#typesmapping).

You can read the **Twitter documentation** (https://dev.twitter.com/docs/platform-objects/tweets) to understand what information each tweet contains and how to access it, but it's not too difficult to deduce the structure by direct inspection.

Your script should print to stdout the sentiment of each tweet in the file, one numeric sentiment score per line. The first score should correspond to the first tweet, the second score should correspond to the second tweet, and so on. If you sort the scores, they won't match up. If you sort the tweets, they won't match up. If you put the tweets into a dictionary, the order will not be preserved. Once again: The nth line of the file you submit should contain only a single number that represents the score of the nth tweet in the input file!

NOTE: You must provide a score for **every** tweet in the sample file, even if that score is zero. You can assume the sample file will only include English tweets and no other types of streaming messages.

Hint: This is real-world data, and it can be messy! Refer to the **twitter documentation**(https://dev.twitter.com/docs/platform-objects/tweets) to understand more about the data structure you are working with.
Don't get discouraged, and ask for help on the forums if you get stuck!

What to turn in: The file tweet sentiment.py after you've verified that it returns the correct answers.

Problem 3: Derive the sentiment of new terms

In this part you will be creating a script that computes the sentiment for the terms that **do not** appear in the file AFINN-111.txt.

Here's how you might think about the problem: We know we can use the sentiment-carrying words in AFINN-111.txt to deduce the overall sentiment of a tweet. Once you deduce the sentiment of a tweet, you can work backwards to deduce the sentiment of the non-sentiment carrying words that do *not* appear in AFINN-111.txt. For example, if the word *soccer* always appears in proximity with positive words like great and fun, then we can deduce that the term *soccer* itself carries a positive sentiment.

Don't feel obligated to use it, but the following paper may be helpful for developing a sentiment metric. Look at the Opinion Estimation subsection of the Text Analysis section in particular.

O'Connor, B., Balasubramanyan, R., Routedge, B., & Smith, N. From Tweets to Polls: Linking Text Sentiment to Public Opinion Time Series. (ICWSM), May 2010.

(http://www.cs.cmu.edu/~nasmith/papers/oconnor+balasubramanyan+routledge+smith.icwsm10.pdf)

You are provided with a skeleton file term_sentiment.py which accepts the same two arguments as tweet_sentiment.py and can be executed using the following command:

\$ python term_sentiment.py AFINN-111.txt output.txt

Your script should print output to stdout. Each line of output should contain a term, followed by a space, followed by the sentiment. That is, each line should be in the format <term:string> <sentiment:float>

For example, if you have the pair ("foo", 103,256) in Python, it should appear in the output as:

foo 103.256

The order of your output does not matter.

What to turn in: The file term_sentiment.py

How we will grade Part 3: We will run your script on a file that contains strongly positive and strongly negative tweets and verify that the non-sentiment-carrying terms in the strongly positive tweets are assigned a higher score than the non-sentiment-carrying terms in negative tweets. Your scores need not (and likely will not) exactly match any specific solution.

If the grader is returning "Formatting error: ", make note of the line of text returned in the message. This line corresponds to a line of your output. The grader will generate this error if line.split() does not return **exactly two items**. One common source of this error is to not remove the two calls to the "lines" function in the solution template; this function prints the

number of lines in each file. Make sure to check the first two lines of your output! (Some terms in AFINN are short phrases; we will not test your script using any multi-word combinations.)

Problem 4: Compute Term Frequency

Write a Python script frequency.py to compute the *term frequency histogram* of the livestream data you harvested from Problem 1.

The frequency of a term can be calculated as [# of occurrences of the term in all tweets]/[# of occurrences of all terms in all tweets]

Your script will be run from the command line like this:

\$ python frequency.py <tweet_file>

You should assume the tweet file contains data formatted the same way as the livestream data.

Your script should print output to stdout. Each line of output should contain a term, followed by a space, followed by the frequency of that term in the **entire file**. There should be one line per **unique** term in the entire file. Even if 25 tweets contain the word lol, the term lol should only appear once in your output (and the frequency will be at least 25!) Each line should be in the format <term:string> <frequency:float>

For example, if you have the pair (bar, 0.1245) in Python it should appear in the output as:

bar 0.1245

If you wish, you may consider a term to be a multi-word phrase, but this is not required. You may compute the frequencies of individual tokens only.

Depending on your method of parsing, you may end up computing frequencies for hashtags, links, stop words, phrases, etc. If you choose to filter out these non-words, that's ok too.

What to turn in: The file frequency.py

Problem 5: Which State is happiest?

Write a Python script happiest state by that returns the name of the happiest state as a string.

Your script happiest state by should take a file of tweets as input. It will be called from the command line like this:

\$ python happiest_state.py <sentiment_file> <tweet_file>

The file AFINN-111.txt contains a list of pre-computed sentiment score.

Assume the tweet file contains data formatted the same way as the livestream data.

It's a good idea to make use of your solution to Problem 2.

There are different ways you might assign a location to a tweet. Here are three:

- Use the coordinates field (a part of the place object, if it exists), to geocode the tweet. This method gives the most reliable location information, but unfortunately this field is not always available and you must figure out some way of translating the coordinates into a state.
- Use the other metadata in the place field. Much of this information is hand-entered by the twitter user and may not always be present or reliable, and may not typically contain a state name.
- Use the user field to determine the twitter user's home city and state. This location does not necessarily correspond to the location where the tweet was posted, but it's reasonable to use it as a proxy.

You are free to develop your own strategy for determining the state that each tweet originates from.

You may find it useful to use this python dictionary of state abbreviations (http://code.activestate.com/recipes/577305-python-dictionary-of-us-states-and-territories/).

You can ignore any tweets for which you cannot assign a location in the United States.

In this file, each line is a Tweet object, as described in the twitter documentation (https://dev.twitter.com/docs/platform-objects/tweets).

Note: Not every tweet will have a text field --- again, real data is dirty! Be prepared to debug, and feel free to throw out tweets that your code can't handle to get something working. For example, you might choose to ignore all non-English tweets.

Your script should print the two letter state abbreviation of the state with the highest average tweet sentiment to stdout.

Note that you may need a **lot** of tweets in order to get enough tweets with location data. Let the live stream run for a while if you wish.

Your script will not have access to the Internet, so you cannot rely on third party services to resolve geocoded locations!

What to turn in: The file happiest_state.py

Problem 6: Top ten hash tags

Write a Python script top_ten.py that computes the ten most frequently occurring hashtags from the data you gathered in Problem 1.

Your script will be run from the command line like this:

\$ python top ten.py <tweet file>

You should assume the tweet file contains data formatted the same way as the livestream data.

In the tweet file, each line is a Tweet object, as **described in the twitter documentation**(https://dev.twitter.com/docs/platform-objects/tweets). To find the hashtags, you should not parse the text field; the hashtags have already been extracted by twitter.

Your script should print to stdout each hashtag-count pair, one per line, in the following format:

Your script should print output to stdout. Each line of output should contain a hashtag, followed by a space, followed by the frequency of that hashtag in the **entire file.** There should be one line per **unique** hashtag in the entire file. Each line should be in the format hashtag:string frequency:float>

For example, if you have the pair (bar, 30) in Python it should appear in the output as:

bar 30

What to turn in: the file top_ten.py

Twitter Correctness

Criteria Part 1 turned in	Ratings		Pts
	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts
Part 1 is a valid tweet file	2.0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 2 turned in	2,0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 2 produces correct output	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts
Part 3 turned in	2.0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 3 produces correct output	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts
Part 4 turned in	2.0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 4 produces correct output	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts
Part 5 turned in	2.0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 5 produces correct output	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts
Part 6 turned in	2.0 pts Full Marks	0.0 pts No Marks	2.0 pts
Part 6 produces correct output	3.0 pts Full Marks	0.0 pts No Marks	3.0 pts

Total Points: 30.0