README

Overview

The ***MITLL TweetE Twitter Analysis Tools*** perform multiple types of analysis on Twitter data:

- Unstructured tweets to structured data and text normalization
- Twitter graph creation

Tweets are ingested from a flat TSV (tab-separated value) file. Results are stored in a serialized Python object (text analysis and normalization) and multiple graph formats. Examples of research applications that used these tools are contained in the papers [WCampbell13] and [WCampbell14].

Provided a collection of tweets, the *MITLL TweetE Twitter Analysis Tools*:

1. Normalize the input text and remove links and non-language characters
2. Extract information: hashtags, links, at-mentions
3. Filter out all documents not matching the user-specified language
4. Filter by geo location
5. Store the results in serialized files for graph creation, analysis with the MIT topic tools, or other counts-based classifiers
6. Create rich Twitter graphs

These tools are command-line applications mainly suited for researchers who would like to convert Twitter data into structured form for further high-level analysis—e.g., natural language processing and graph analysis.

Download

The code for the MITLL Twitter Analysis Tools is publicly accessible through a git repository at https://github.com/mitll/TweetE . For help on *github* and *git*, please refer to https://github.com and http://git-scm.com/ respectively.

Once you have downloaded the code, you can run the program by installing the dependencies listed in section ***Dependencies***, and following the steps described in section ***Running the <>***.

Dependencies

The system is a command-line application for 64-bit Linux written in Python and C++. It requires the following dependencies:

- Python 2.7.x
- NumPy (Python module)
- NetworkX (https://networkx.github.io/). Install NetworkX for your local python version using the instructions at the NetworkX website.

Target Platform

- 64-bit Linux

Running the Twitter Ingester

The shell script

`run_ingest_all.sh`

shows a simple example of running the ingest tool.

In the scripts, the ingest tool takes "tsv.gz" files from the directory `twitter/user_tweets` and converts them into serialized files in the `twitter/serialized` directory.

Data Format and Ingestion

Twitter data is contained in tab-delimited UTF-8 text files that are gzipped. The ingester assumes the input file has one line per document in the following format:

<tweetid>TAB<date>TAB<language>TAB<geo-coordinates>TAB<username>TAB<tweet text>NEW_LINE

A more detailed description of these fields can be found at the Twitter API website, https://dev.twitter.com/overview/api/tweets .

An example from the file `aaastudio.tweets.tsv.gz` is shown as follows:

```
286163982614679552      Tue Jan 01 17:36:24 +0000 2013  en      (0.0, 0.0)      aaastudio
Happy New Year!! Wishing all a healthy, happy and prosperous 2013!!
286173401771565056      Tue Jan 01 18:13:50 +0000 2013  en      (0.0, 0.0)      aaastudio
@RW2Photo fantastic
286338113477939200      Wed Jan 02 05:08:20 +0000 2013  und     (0.0, 0.0)      aaastudio
Day 1 http://t.co/8W9h6tJZ
286898889729142784      Thu Jan 03 18:16:40 +0000 2013  es      (0.0, 0.0)      aaastudio
Carlos  (c)2012 Angel Alvarado / AAA Studio http://t.co/iOM5gH9Z
287696368833818624      Sat Jan 05 23:05:34 +0000 2013  en      (0.0, 0.0)      aaastudio
In the middle of a shoot
```

Output

After running the script ./`run_ingest_all.sh` in the `twitter_analysis` directory, serialized files are saved to the `twitter/serialized` directory. The output can be displayed using the tool `scripts/display_tweets.py`. An example is shown as follows:

```
$ scripts/display_tweets.py --in twitter/serialized/tw_user_tweets_aaastudio.pckl --output
tmp/a.txt
Reading in file: twitter/serialized/tw_user_tweets_aaastudio.pckl
Done
```

```
% head -18 tmp/a.txt

date Tue Aug 27 07:56:29 +0000 2013
geo (0.0, 0.0)
http_links [(117,http://t.co/ao8Uj2DteD)]
id 372266352943198208
lid_gnip en
msg Really great news I've just upgraded to Spotify Premium I've got unlimited, ad-free music on
my mobile and computer. http://t.co/ao8Uj2DteD
msg_norm Really great news I've just upgraded to Spotify Premium I've got unlimited ad-free music
on my mobile and computer.
userid aaastudio

date Sun May 26 20:23:26 +0000 2013
geo (40.76511645, -73.97349006)
http_links [(54,http://t.co/6X9OUPx3C7)]
id 338752255736045569
lid_gnip en
msg I'm at Central Park South (New York, NY) w/ 13 others http://t.co/6X9OUPx3C7
msg_norm I'm at Central Park South New York NY w 13 others
userid aaastudio
```

From the example, we see that the system has extracted http_links and has normalized the text. Further inspection of the structured output will show extracted at-mentions and hashtags.

Running the Graph Creation Tool

The shell script

`run_graph_all.sh`

shows an advanced example of running the graph creation tool. The `run_ingest_all.sh` tool must be run before graph creation.

The script `run_graph_all.sh` is designed for processing a large amount of Twitter data and creating graphs in multiple steps. The steps are:

1. Take multiple serialized files and create a graph per list (*run_p1* in the code)
2. Merge multiple graphs to create another set of graphs (*run_p2* in the code)
3. Merge multiple graphs to create a node set and edge set with duplicates (*run_p3* in the code)
4. Merge edge set duplicates (*run_p4*)
5. Create various versions of the final graph by pruning nodes by weighted degree

A typical run might create 100 graphs in step 1, 10 graphs in step 2, 1 graph in step 3, perform an edge merger in step 4, and then prune with different node weights. If only analysis of a small amount of Twitter data is needed, then using the code in steps 1-2 may be sufficient (the original code will need to be modified).

Output

After running the graph creation tool, the results will be stored in the directory `graph/`. For the example files the output files should be:

```
node_degree.txt.gz              twitter_prune_w10.edges.txt.gz
```

```
twitter_prune_w5.edges.txt.gz  twitter_all.edges.txt.gz          twitter_prune_w10.nodes.txt.gz
twitter_prune_w5.nodes.txt.gz twitter_all.nodes.txt.gz         twitter_prune_w1.edges.txt.gz
twitter_prune_w100.edges.txt.gz  twitter_prune_w1.nodes.txt.gz     twitter_prune_w50.edges.txt.gz
twitter_prune_w100.nodes.txt.gz  twitter_prune_w50.nodes.txt.gz
```

The files `twitter_all_*.txt.gz` contain the unpruned graphs and the files with `twitter_prune_w<weight>.*.txt.gz` contain the versions pruned by node degree. An example few nodes in the node file are:

```
$ zcat graph/twitter_all.nodes.txt.gz | head

1118 #truth
12102 @jayray809
7946 @hellocupkake
6617 @tomcrabtree
7947 @oglethewriter
1119 #eminem
1120 #dailyshow
9733 @doingitwrong
3827 @tomorrowsprjct
11040 @spacekisser
```

The format of this file is <nodeid> <twitter hashtag/username> in space delimited format. Note that the node id is *not* the Twitter user id and is arbitrary. An example from the edge file is as shown:

```
zcat graph/twitter_all.edges.txt.gz | head

0 6 1
0 9 0 0 1
0 10 6 0 3
0 15 2 0 0
0 20 2 0 1
0 21 0 0 2
0 24 4
0 37 7
0 39 2 0 0
0 40 1
```

The format is <src node> <dest node> <w1> <w2> <w3>. The source and destination nodes refer to the node ids in the `twitter_all.nodes.txt.gz` file. Note that the graph is directed and by design the adjacency matrix is not symmetric. The weights depend on the edge type. There are three cases:

- Hashtag -> Hashtag: w1=co-occurrence count of these hashtags
- User->Hashtag: w1=the number of times the hashtag occurred in tweets from this user
- User->User: w1=co-occurrence count of user names, w2=communication count of user src with user dest, w3=retweet of user dest by user src

More details on the node and edge structure and sample graph analysis can be found in [WCampbell13] and [WCampbell14].

References

[WCampbell13] W. M. Campbell, E. Baseman, K. Greenfield, "Content+Context Networks for User Classification in Twitter," NIPS 2014 Workshop, Frontiers of Network Analysis: Methods, Models, and Applications, 2013.

[WCampbell14] W. M. Campbell, E. Baseman, K. Greenfield, "Content+Context=Classification: Examining the Roles of Social Interactions and Linguistic Content in Twitter User Classification," Coling 2014 Workshop on Natural Language Processing for Social Media (SocialNLP), 2014.