Analyzing Spammers in Twitter User Network

Mananai Saengsuwan (@JustAProgrammer)

Motivation

- Spams irritate users
- If we could identify spammers, we could block those accounts
- It's a good ML and network analysis exercise

Steps to do the analysis

- 1) Collect Twitter data
- 2) Implement a Spam classifier
- 3) Import Twitter data to the graph database
- 4) Analyze the data

The 2nd-4th steps will be done in python notebooks

Collecting Twitter data

- There are a few resources cover this topic. The followings are samples:
 - Bonzanini, Marco. "Users, Followers, and Communities on Twitter". *Mastering Social Media Mining with Python*, Packt Publishing, 2016.
 - Saengsuwan, Mananai. "Building a network graph from Twitter data". Web blog post. *Towards Data Science*. October 6, 2020.
 - Saengsuwan, Mananai. "Analyzing Twitter User Network". Web blog post. *Towards Data Science*. June 23, 2022.

Collecting Twitter Data (Cont.)

Steps

- Identify 50 popular trends or hashtags
- For each trend, search for tweets
- For each tweet, identify user who writes or retweets it
- Finally, find friends of all those users

Collecting Twitter Data (Cont.)

- The data we collected are:
 - Statuses (Tweets)
 - Users
 - User-friends relationships
- Also, we have labeled statuses (Spam or Ham) collected on different days for classification
- All data are in CSV format which could be imported to the graph database

Spam classification

What we need

- Labeled tweets file (training data)
- Unlabeled tweets file. The classifier would predict this data
 Output
- Spam/Ham column in the tweets file

Spam Classification(Cont.)

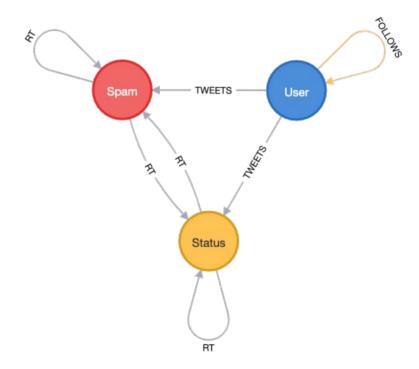
Steps:

- 1) Read training tweet data file
- 2) Convert to feature vectors using TFID
- 3) Split the data into training and testing
- 4) Use Bayes as a classifier.
- 5) Fit it with the training data
- 6) Run the classifier with the testing data and check the scores (recall, precision)
- 7) Apply the classifier with the real tweet data will be used in the next step

Importing data

- Copy the users, user_friends, and statuses files to the import directory
- Run import commands (Load CSV)
- Run a few Cypher commands to create relationships

Importing Data (Cont.)



Network analysis

- Visualization
- Triangle analysis
- Communities detection

Spammers Network Visualization using NX



Triangle analysis

- Three users follow each other. Two of them are spammer. The rest is a candidate
- To identify more spammers from existing ones.
- Limited success in our data.

Community detection

- Run Louvain community detection
- Identify communities with majorities are spammers. Assume all of them are spammers
- The community with a lot of spammer has very high relationship density. They follow each other a lot with in their group.
- In our data, we found one community. After a brief check, almost all are probably spammers

Conclusions

- From the visualization, spammers probably work together
- By using community detection, we could identify more spammers

Issue(s)

• Spam classifier may misses some spams. We would not detect some spammers.

Improvements

- More labeled tweet data may improve classifier performance
- Twitter data collected here is only a tiny amount. But to collect more data:
 - Need more time because of Twitter API rate limit
 - Bigger graph database consumes more resources (RAM and hard disk space)

Notable software library

In addition to several python libraries, I use the following Thai NLP software to tokenize text:

• Wannaphong Phatthiyaphaibun, Korakot Chaovavanich, Charin Polpanumas, Arthit Suriyawongkul, Lalita Lowphansirikul, & Pattarawat Chormai. (2016, Jun 27). PyThaiNLP: Thai Natural Language Processing in Python. Zenodo. http://doi.org/10.5281/zenodo.3519354

Github repository

For Twitter data collection:

• https://github.com/mananai/TwitterUserNetwork

For spam classification, data import, and spammer analysis:

• https://github.com/mananai/NODES2022-AnalyzingSpammers

