NLP Twitter Analysis

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1 Introduction

This is the final report for the NLP course project. In this project we have collected tweets from Twitter and labeled them using ChatGPT model. Then we have augmented the data using GPT-3.5-turbo model. After that we have trained a classifier using the augmented data and evaluated the results.

Word2vec, tokenizer and language model and other stuffs also have been trained on the collected data, which we will discuss in the following sections.

2 Repository

You can access the source code of this project at https://github.com/hamedhf/nlp_twitter_analysis

3 Installation

You should create a file named "users.csv" inside src folder which contains the Twitter username, University name an Actual name of the users you wish to analyze.

Furthermore installation instructions are provided in the README.md file of the repository.

4 Project Structure

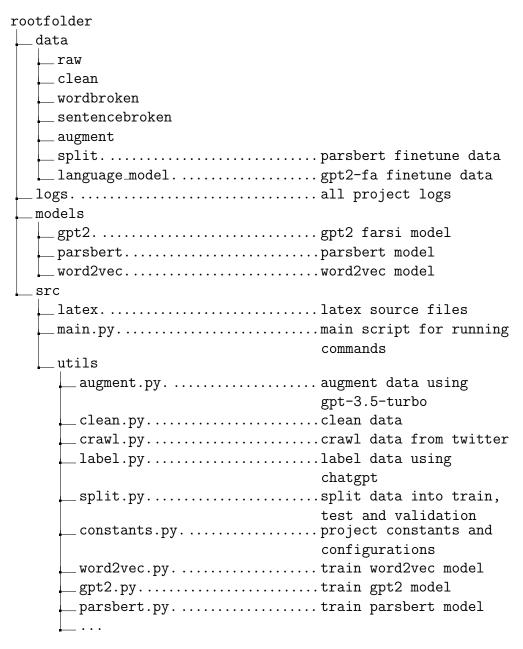


Figure 1: Project Tree

5 Data Collection

We used selenium >= 4.6.0 for collecting data from Twitter. This tool helps us to bring up an actual browser and navigate through the pages. Note that you should have Chrome installed on your system for this to work. Then you can simply install other dependencies form pyproject.toml file using poetry or other package managers. The crawler script reads the users.csv file and for each user, it navigates to the user's profile and collects the tweets. The tweets are stored in a file named unlabeled.db inside data/raw folder. Then labeling script uses this and with the help of ChatGPT model, it generates the labels for each tweet and stores them in data/raw/labeled-run-date.csv file.

6 Data Format

The data is stored in a csv file with the following format: tweet_time, tweet_owner, tweet_text, owner_university, owner_name, label. We use tweet_time, tweet_owner as unique identifiers for each tweet. The tweet_owner is Twitter username of the owner. The tweet_text is the actual text of the tweet. owner_university and owner_name are the university and

actual name of the tweet owner. The label is the generated label for the tweet.

7 Data Preprocessing

We have splitted data with three criteria: split by sentence with hazm sentence tokenizer, split by word with hazm word tokenizer, split by word with hazm lemmatizer.

For cleaning the data, we used the following steps: remove emojis, remove urls, remove hashtags, remove mentions, remove numbers, remove punctuations. We used the hazm, cleantext and nltk libraries for this purpose.

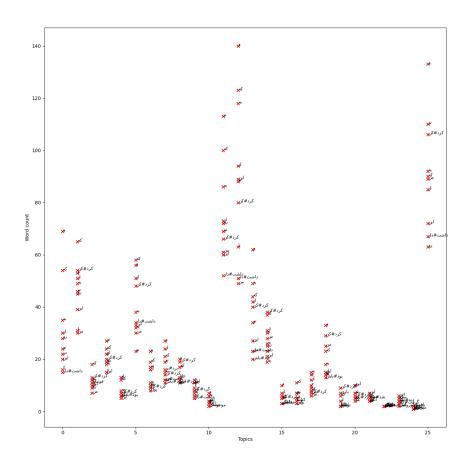
8 Labeling

We give label to the whole tweet using ChatGPT. For more info about labeling see the src/utils/label.py file. You can also see the labels in src/utils/

constants.py file.

9 Statistics

tweet-count	word-count	sentence-count	unique-word-count
2079	31987	2944	6725



10 Augmenting Data

For this part to work you need to sign up for an account at https://platform.openai.com and provide your openai api key in .env file. Then you can run the augment-data command.

The augmentation script takes a cleaned csv file as input and counts how many tweet we have for each label. Then it will fix the imbalance of the data by generating new tweets for the labels with less tweets. The generated tweets are stored in data/augment folder.

You can see the implementation detail of the augmentation script in src/utils/augment.py file. We have used gpt-3.5-turbo model for this purpose and the given prompt is like this:

```
import openai
      label = "home_and_garden"
      temperature = 0.6
      system_message = "Generate an informal Persian tweet
     about the given topic without any hashtags, mentions,
     links, or emojis." # noqa
      messages = [
          {
              "role": "system",
              "content": system_message
          {"role": "user", "content": f"topic: {label}"}
      response = openai.ChatCompletion.create(
12
          model="gpt-3.5-turbo",
13
          messages=messages,
14
          temperature=temperature,
          timeout=120
      )
17
```

Temperature is a parameter that controls the randomness of the generated text. The higher the temperature, the more random the text. The lower the temperature, the more predictable the text. We have used 0.6 for this parameter and it is random enough for our purpose and if we increase it will take much more time to generate the text and this is not practical for our purpose.

Using this approach we have doubled our total data size and each label has at least 200 tweets. It is worth mentioning that because of openai api rate limit, it took us about **2** and a half days to generate the data.

10.1 Generated Tweets

Some of the generated tweets:



10.2 Augmented Data Statistics

	label	tweet count
1	politics_and_current_affairs	200
2	$entertainment_and_pop_culture$	200
3	$sports_and_athletics$	200
4	technology_and_innovation	200
5	science_and_discovery	200
6	$health_and_wellness$	200
7	business_and_finance	200
8	$travel_and_adventure$	200
9	food_and_cooking	200
10	$fashion_and_style$	200
11	environment_and_sustainability	200
12	education_and_learning	216
13	social_issues_and_activism	217
14	$in spirational_and_motivational$	200
15	$funny_and_humorous$	200
16	$\operatorname{art_and_design}$	200
17	books_and_literature	200
18	religion_and_spirituality	200
19	family_and_parenting	200
20	gaming	200
21	beauty_and_cosmetics	200
22	home_and_garden	200
23	automotive	200
24	pets_and_animals	200
25	$weather_and_seasons$	200
26	other	413

11 Word2Vec

We used gensim library for training skipgram word2vec model because it is easy to use and fast. The implementation is in src/utils/word2vec.py file.

All of the available commands are listed in **ReadME.md** file. Here we explain some of them.

11.1 Training

This command trains word2vec for a specific label.

```
python src/main.py train-word2vec-label path-to-augmented
-csv home_and_garden
```

This command trains word2vec for some preselected labels.

```
python src/main.py train-word2vec-preselected path-to-
augmented-csv
```

This command trains word2vec for all labels.

```
python src/main.py train-word2vec-all path-to-augmented-
csv
```

11.2 Evaluation

Let's find that in topic of home_and_ garden, which words are similar to the Persian word for home (khaneh).

```
python src/main.py get-most-similar-words home_and_garden عنائات --topn 10

[2023-07-11 18:34:11,783: INFO/main-get_most_similar_words] Most similar words to عنائات in home_and_garden are:

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9983492493629456

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982892870903015

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982603788375854

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982492859483261108

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982428550720215

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982321858406067

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982321858406067

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.998212857610015

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9982027411460876

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.998189857521057

2026-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9981461763381958

[2023-07-11 18:34:11,790: INFO/main-get_most_similar_words] 0.9981461763381958
```

Some of the similarity results are shown in the following image. We use cosine similarity for measuring the similarity between two words. The higher the similarity, the more similar the words are.

```
کلمات مشابه برای کلمه سیاست:
                                             کلمات مشابه برای کلمه ایران:
[('جاري', 0.984733521938324),
                                             [('تعداد', 0.9875343441963196),
('امور', 0.9736297726631165),
                                             ('خوابگاه', 0.9871758818626404),
('پيچيده', 0.9675320982933044),
                                             ('حكم', 9862680435180664),
                                             ('ددلاين', 0.9860983490943909),
('روزها', 0.9606905579566956),
('سياسي', 0.9490455389022827),
                                             ('اونور', 0.9860574007034302),
('יָוֹלֶוֹלֶ, 0.9372583627700806),
                                             ('اولش', 0.9858595728874207),
('اخبار', 0.9359972476959229),
                                             ('انقلاب', 0.9857739210128784),
('شده', 0.9271780252456665),
                                             ('تنها\u200cتری', u200c/1419334412),
                                             ('بي\u200cگناه', u200c611785889),
('تجارت', 0.925977885723114),
('گرون', 0.9195088744163513)]
                                             ('تومن', 0.9853056073188782)]
```

```
كلمات مشابه براى كلمه گل:
(الباغچه\0.9820027351379395, |u200c\0.9828597564697266, |u200c\0.9808597564697266, |u200c\0.9766149520874023, |u200c\0.970939040184021, |u200c\0.9690070748329163, |u200c\0.9681393504142761, |u200c\0.9678846001625061, |u200c\0.9676481485366821, |u200c\0.9649395942687988, |u200c\0.96494884, |u200c\0.96494884, |u200c\0.9649484, |u200c\0.9649484, |u200c\0.9649484, |u200c\0.9649484, |u200c\0.9649484, |u200c\0.964944, |u200c\0.96494, |u200c\0.96494, |u200
```

كلمات مشابه براى كلمه ماشين: (ابخرم', 0.9738640189170837, (اروزى', 0.9613807797431946,), (ايكيشونو', 0.9528212547302246,), (اقدرتمند', 0.948646605014801, (ابتونم', 0.9442844390869141, (اخفن', 0.9425691366195679, (اخيابونا', 0.9352948069572449, (اميخوام', 0.9346776604652405, (ابشم', 0.9335148334503174, (اخريد', 0.9271440505981445,

12 Resources

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