

```

from google.colab import drive
drive.mount('/content/drive')
%cd /content/drive/MyDrive/Precog

Mounted at /content/drive
/content/drive/MyDrive/Precog

import nltk
nltk.download('stopwords')

[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data]   Package stopwords is already up-to-date!
True

```

## ▼ References

<https://www.dezyre.com/student-project/toly-novik-text-mining-and-clustering-of-tweets-based-on-context/2#:~:text=Identify%20tweets%20that%20are%20talking,used%20words%20in%20each%20topic.>

## ▼ Extract tweets

```
# !python3 tw.py
```

## ▼ Imports

```

import pandas as pd
import numpy as np
import re

# plot
# import seaborn as sns
import matplotlib.pyplot as plt

# Gensim for sentiment analysis
import gensim
from gensim.utils import simple_preprocess

import nltk
from nltk.corpus import stopwords, twitter_samples

from collections import Counter
from wordcloud import WordCloud

```

```
from wordcloud import WordCloud
```

```
import warnings
warnings.filterwarnings("ignore", category=DeprecationWarning)
```





```
import tweepy
from tweepy import OAuthHandler
from textblob import TextBlob
```

```
from tqdm import tqdm
import matplotlib.pyplot as plt
```

## ▼ Preprocessing

```
df = pd.read_csv('./df.csv')
```

```
df = df.dropna()
df.head()
```

|   | CreatedAt              | tweets   | User_ID      | User_Name   |
|---|------------------------|--|--------------|---|
| 0 | 2020-12-28<br>14:09:51 | b'RT @TicketNew:<br>\xf0\x9f\x9a\xa8\xf0\x9f\x94\... | 1.189225e+18 | Raj    |
| 2 | 2020-12-28<br>14:09:50 | b'@Aji_spartan \n\n@sharan14110326<br>\n\n@Vasant... | 1.251350e+18 |  Jeni  Ice<br>Doll  LK |
| 4 | 2020-12-28<br>14:09:46 | b"RT @ManokarVj: Teaser & Trailer<br>announce...     | 1.008243e+18 | Bigil Ashok   |

```
df.rename(columns = {'Tweet': 'tweets'}, inplace=True)
```

```
count = df['tweets'].str.split().str.len()
count.index = count.index.astype(str)+' words:'
print("Total number of words in the tweets =",count.sum())
print("Mean number of words per tweet :", round(count.mean(), 2))
```

```
Total number of words in the tweets = 186033
Mean number of words per tweet : 13.68
```

```
print('Users: ', len(df['User_ID'].unique()))
df['tweet_length'] = df['tweets'].str.len()
print("total characters in tweets =", df['tweet_length'].sum())
print("mean number of characters per tweet =", df['tweet_length'].mean())
```

```

df = df.drop(['tweet_length'], axis=1)

Users: 3678
total characters in tweets = 2242975
mean number of characters per tweet = 164.9246323529412

def remove_users(tweet, pattern1, pattern2):
    r = re.findall(pattern1, tweet)
    for i in r:
        tweet = re.sub(i, '', tweet)
    r = re.findall(pattern2, tweet)
    for i in r:
        tweet = re.sub(i, '', tweet)
    return tweet

df['tidy_tweets'] = np.vectorize(remove_users)(df['tweets'], "@ [\w]*", "@[\w]*")

df['tidy_tweets'] = df['tidy_tweets'].str.lower()
df['tidy_tweets'] = df['tidy_tweets'].apply(lambda x: x.strip('b').strip('"').strip('\'').replac

# has hashtags so removing them
df['tidy_tweets'] = np.vectorize(remove_users)(df['tidy_tweets'], "# [\w]*", "#[\w]*")

filtr = df['tidy_tweets'].str.len() != 0
df = df[filtr]

def remove_links(tweet):
    tweet_no_link = re.sub(r"http\S+", "", tweet)
    return tweet_no_link

# removing links

df['tidy_tweets'] = np.vectorize(remove_links)(df['tidy_tweets'])

# removing punctuations
df['tidy_tweets'] = df['tidy_tweets'].str.replace("[^a-zA-Z#]", " ")

# removing shortwords
df['tidy_tweets'] = df['tidy_tweets'].apply(lambda x: ' '.join([i for i in x.split() if len(i

def tokenize(tweet):
    for word in tweet:
        yield(gensim.utils.simple_preprocess(str(word), deacc=True))

df['tidy_tweet_tokens'] = list(tokenize(df['tidy_tweets']))

```

```
def remove_stopwords(tweets):
    return [[word for word in simple_preprocess(str(tweet)) if word not in stop_words] for

# now remove stopwords
stop_words = stopwords.words('english')
stop_words.extend(['from', 'https', 'twitter', 'religions', 'pic', 'twitt',])
df['tokens_no_stop'] = remove_stopwords(df['tidy_tweet_tokens'])
print("\nSTOPWORDS REMOVED\n")
print(df['tokens_no_stop'].head())

# REMOVE TWEETS LESS THAN 3 TOKENS
df['length'] = df['tokens_no_stop'].apply(len)
df = df.drop(df[df['length']<3].index)
df = df.drop(['length'], axis=1)
```

STOPWORDS REMOVED

```
0          [tomorrow, ready]
2          [follow, keep, suppor]
4  [teaser, trailer, announcement, pics, thalapat...
6          [chances, falling, sunday, papom]
8          [update, today]
Name: tokens_no_stop, dtype: object
```

```
!pip install tqdm
```

Requirement already satisfied: tqdm in /usr/local/lib/python3.6/dist-packages (4.41.1)

```
from tqdm import tqdm
# generating a wordcloud for visual representation
string=''
for i,r in tqdm(df.iterrows()):
    string+= ' '.join(r['tokens_no_stop'])+' '

# wordcloud = WordCloud(width = 800, height = 800,
#                         background_color = 'white',
#                         stopwords = stop_words,
#                         min_font_size = 10).generate(string)

wordcloud = WordCloud(width = 3000, height = 2000, random_state=1, background_color='black',
# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)
plt.show()
```



```
def get_tweet_sentiment(tweet):
    '''
    Utility function to classify sentiment of passed tweet
    using textblob's sentiment method
    '''
    # create TextBlob object of passed tweet text
    analysis = TextBlob(tweet)

    # print(analysis.sentiment)

    # set sentiment
    if analysis.sentiment.polarity > 0:
        return 'positive', analysis.sentiment.polarity, analysis.sentiment.subjectivity

    elif analysis.sentiment.polarity == 0:
        return 'neutral', analysis.sentiment.polarity, analysis.sentiment.subjectivity
    else:
        return 'negative', analysis.sentiment.polarity, analysis.sentiment.subjectivity
```

```
'''
```

calculating polarities for tweets on the given hashtag

link to output of the cell :

<https://colab.research.google.com/drive/1B1KqIjwpTMgC8L1C0eooeUNscjkQXdua?authuser=1#scrollTo>

```
'''
```

```
from tqdm import tqdm
sentiments = []
polarities = []
subjectivities = []
for i,r in tqdm(df.iterrows()):
    temp = get_tweet_sentiment(r['tidy_tweets'])
    sentiments.append(temp[0])
    polarities.append(temp[1])
    subjectivities.append(temp[2])
df['sentiment'] = sentiments
df['polarity'] = polarities
df['subjectivity'] = subjectivities
df.head()

"""
The below line was used to save the modified dataframe
Uncomment to do save the dataframe with polarities.
"""

# df['User_ID'] = df['User_ID'].astype('int64')
# df.to_csv("./sentiments_and_polarities_with_pre_processed.csv", index=False)
```

9691it [00:06, 1433.25it/s]

```
df.to_csv('./sentiments_and_polarities_with_pre_processed.csv')
```

```
df = pd.read_csv( './sentiments_and_polarities_with_pre_processed.csv' )
```

```
# getting unique users
user = df['User_ID'].unique()
print("Number of unique users in the sampled dataset",len(user))
```

Number of unique users in the sampled dataset 3026

```
# Creating a dictionary which stores number of tweets of a user on "#MasterTrailer"
user_freq = dict(zip(user, [0]*len(user)))
for i,r in df.iterrows():
    user_freq[r['User_ID']] += 1
#this shows how active a user was
```

```
...
```

The below code was just used to save the users' active status.  
Uncomment to do save the findings.

```
...
```

```
# file = open("unique_users.txt", "w+")
# for i in user:
#     file.write(str(i))
#     file.write("\n")
# file.close()
```

```
# file = open("users_freq.txt", "w+")
# file.write(str(user_freq))
# file.close()
```

```
# [negative, neutral, positive]
df = pd.read_csv('sentiments_and_polarities_with_pre_processed.csv')
sentiment_of_user=dict(zip(user, [0]*len(user)))
for i,r in tqdm(df.iterrows()):
    polarity = r['polarity']
    if type(sentiment_of_user[r['User_ID']])!=int:
        sentiment_of_user[r['User_ID']]=[0,0,0]
    if polarity>0:
        sentiment_of_user[r['User_ID']][2] += 1
    elif polarity==0.0:
        sentiment_of_user[r['User_ID']][1] += 1
    else:
        sentiment_of_user[r['User_ID']][0] += 1
```

```
...
```

The below code is just to save our findings  
Uncomment to do save the dictionary.

```
...
```

```
# file = open("sentiments_of_users.txt", "w+")
# file.write(str(sentiment_of_user))
# file.close()
```

9691it [00:01, 9485.62it/s]

'\n\nThe below code is just to save our findings\n\nUncomment to do save the dictionary.\n'

```
req_list = list(sentiment_of_user.items())
req_list.sort(key=lambda x: x[0])

y1 = [i[1][0] for i in req_list] #negative
y2 = [i[1][1] for i in req_list] #neutral
y3 = [i[1][2] for i in req_list] #positive

x1 = [i[0] for i in req_list] # users
x2 = [i[0] for i in req_list] # users
x3 = [i[0] for i in req_list] # users

plt.figure(figsize=(10,5))
plt.bar(list(range(len(req_list))),y1, label="Neg", width=5)

# plt.bar(left, height, tick_label = tick_label,
#         width = 0.8, color = ['red', 'green'])

plt.xlabel("User's Index in req_list")
plt.ylabel("# negative tweets by the user")

plt.legend()
# plt.savefig('plot_neg.png')
plt.show()

plt.figure(figsize=(10,5))
plt.bar(list(range(len(req_list))),y2, label="Neu", width = 5)
plt.xlabel("User's Index in req_list")
plt.ylabel("# neutral tweets by the user")

plt.legend()
# plt.savefig('plot_neu.png')
plt.show()

plt.figure(figsize=(10,5))
plt.bar(list(range(len(req_list))),y3, label="Pos")
plt.xlabel("User's index in the req_list")
plt.ylabel("# positive tweets by the user")

plt.legend()
# plt.savefig('plot_pos.png')
plt.show()
```

```
req_list = list(sentiment_of_user.items())
req_list.sort(key=lambda x: x[0])
```

```
neg=pos=neu=0
```



```
    ,
    for i in req_list:
        neg+=i[1][0]
    for i in req_list:
        neu+=i[1][1]
    for i in req_list:
        pos+=i[1][2]

x1 = ['negative', 'neutral', 'positive'] # users

plt.figure(figsize=(10,5))
plt.bar(x1,[neg, neu, pos], label="# of tweets")
plt.xlabel("Sentiment of tweets")
plt.ylabel("# of tweets")

plt.legend()
plt.savefig('plot_tweets_sentiments.png')
plt.show()

# Pie chart, where the slices will be ordered and plotted counter-clockwise:
# labels = ['Same Polarities', 'Different Polarities']
# sizes = [same_polarities, different_polarities]
explode = (0.1, 0, 0)

fig1, ax1 = plt.subplots()
ax1.pie([neg, neu, pos], explode=explode, labels=x1, autopct='%1.1f%%',
        shadow=True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title("Distribution of polarities among tweets ")
plt.savefig('plot_tweets_pie.png')
plt.show()
```



## ▼ Friends

negative                      neutral                      positive  
Sentiment of tweets

'''  
To get the tweets of friends of the users on the "#MasterTrailer"  
Uncomment below line  
'''

```
# !python3 get_friends.py
# !python3 get_friends_tweets.py
```

45.3%

```
df_friends_tweets = pd.read_csv('friend_tweets.csv')
df_friends_tweets.head()
```

|   | user_id             | friend_id           | created_at          | text  |
|---|---------------------|---------------------|---------------------|---|
| 0 | 1039062325645328384 | 1344621554550890498 | 2020-12-31 12:28:55 | #MasterTrailer release date Poster ?              |
| 1 | 1039062325645328384 | 1344588889629970433 | 2020-12-31 10:19:07 | Keep Calm and be Don't Confused 👍 #MasterTrail... |
| 2 | 1039062325645328384 | 1344584551918567426 | 2020-12-31 10:01:53 | #MasterTrailer Holds Whole Movie Cast & Crew 🎬    |

```
from tqdm import tqdm
# df['sentiment'] = ""
# df['polarity'] = None
# df['sentiment'].apply(get_tweet_sentiment)
sentiments = []
polarities = []
subjectivities = []
for i,r in tqdm(df_friends_tweets.iterrows()):
    temp = get_tweet_sentiment(r['text'])
    sentiments.append(temp[0])
    polarities.append(temp[1])
    subjectivities.append(temp[2])
df_friends_tweets['sentiment'] = sentiments
```

```
df_friends_tweets['polarity'] = polarities
df_friends_tweets['subjectivity'] = subjectivities
```

```
df_friends_tweets.head()
```

```
141it [00:00, 897.66it/s]
```

|   | user_id             | friend_id           | created_at             | text   | sentiment |
|---|---------------------|---------------------|------------------------|--|-----------|
| 0 | 1039062325645328384 | 1344621554550890498 | 2020-12-31<br>12:28:55 | #MasterTrailer<br>release date<br>Poster ?                 | neutral   |
| 1 | 1039062325645328384 | 1344588889629970433 | 2020-12-31<br>10:19:07 | Keep Calm and<br>be Don't<br>Confused 👍<br>#MasterTrail... | negative  |

```
...
```

```
user_friend_dict has (user_id, [polarity_sum, subjectivity_sum]) as (Key,Value) pair
polarity_sum is nothing but sum of the polarities of tweets retrieved of the friends of key (
subjectivity_sum is nothing but sum of the subjectivities of tweets retrieved of the friends
...
```

```
temp_friends = df_friends_tweets.groupby('user_id')
user_friend_dict = {}
for i,r in temp_friends:
    polarity = r['polarity'].sum()/r['polarity'].count()
    subjectivity = r['subjectivity'].sum()/r['polarity'].count()
    user_friend_dict[i]=[polarity, subjectivity]
```

```
user_friend_dict
```

```
user_polarity = {}
for i, r in df.groupby('User_ID'):
    polarity = r['polarity'].sum()/r['polarity'].count()
    subjectivity = r['subjectivity'].sum()/r['polarity'].count()
    user_polarity[i]=[polarity, subjectivity]
user_polarity
```

```
users_list = []
friends_list = []
for k in user_friend_dict.keys():
    users_list.append(user_polarity[k])
    friends_list.append(user_friend_dict[k])
same_polarities = different_polarities = 0
```

```
for i in range(len(users_list)):
    if users_list[i][0]>0.0 and friends_list[i][0]>0.0:
        same_polarities+=1
    elif users_list[i][0]==0.0 and friends_list[i][0]==0.0:
```

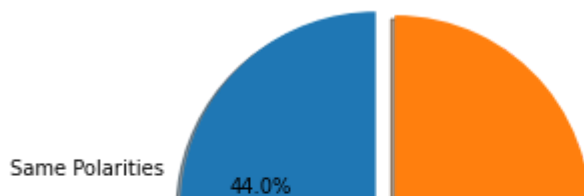
```
    same_polarities+=1
elif users_list[i][0]<0.0 and friends_list[i][0]<0.0:
    same_polarities+=1
else:
    different_polarities+=1

# Pie chart, where the slices will be ordered and plotted counter-clockwise:
labels = ['Same Polarities', 'Different Polarities']
sizes = [same_polarities, different_polarities]
explode = (0, 0.1)

fig1, ax1 = plt.subplots()
ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',
        shadow=True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title("Difference in opinion between a user and its friends")
plt.savefig('plot_users_friends_polarities.png')
plt.show()

plt.figure(figsize=(10,5))
plt.bar(labels,sizes, label="# of users")
plt.xlabel("Sentiment")
plt.ylabel("# of users")
plt.legend()
plt.show()
```

Difference in opinion between a user and its friends



```

friend_count_dict = {}
file = open('./list_friends.txt', "r")
lines = file.readlines()
for i in tqdm(range(0, len(lines), 2)):
    user, followers = lines[i], eval(lines[i+1])
    followers = followers[:20]
    friend_count_dict[int(user)] = [0, len(followers)]

for i, r in df_friends_tweets.groupby('user_id'):
    friend_count_dict[i][0] = r.shape[0]

no_tweet_by_friends = 0

for i in friend_count_dict.keys():
    if friend_count_dict[i][0] == 0:
        no_tweet_by_friends += 1

# friend_count_dict

# Pie chart, where the slices will be ordered and plotted counter-clockwise:
labels = ['Tweeted', 'Did not Tweet']
sizes = [len(list(friend_count_dict.keys())) - no_tweet_by_friends, no_tweet_by_friends]
explode = (0, 0.1)

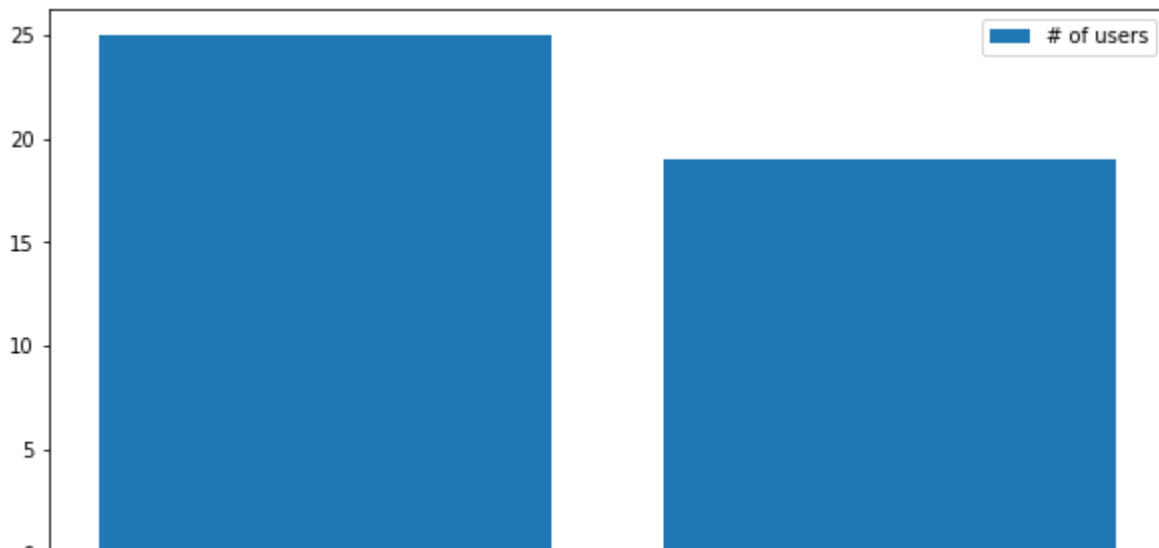
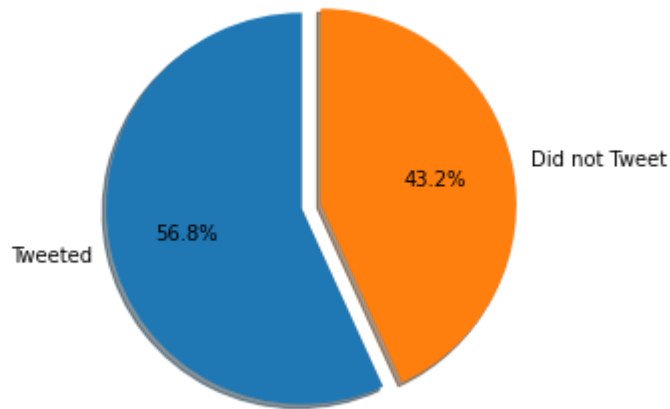
fig1, ax1 = plt.subplots()
ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',
        shadow=True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title("Difference in the number of users with friends who tweeted regarding the same hasht")
plt.savefig('plot_users_friends_tweetted.png')
plt.show()

plt.figure(figsize=(10, 5))
plt.bar(labels, sizes, label="# of users")
# plt.xlabel("Sentiment")
# plt.ylabel("# of users")
plt.legend()
plt.show()

```

100% |██████████| 44/44 [00:00<00:00, 558.41it/s]

Difference in the number of users with friends who tweeted regarding the same hashtag



## ▼ Followers

```
...
```

To extract followers of users and their respective tweets with the hashtag "#MasterTrailer"  
Uncomment the below lines

```
...
```

```
# !python3 get_followers.py
```

```
# !python3 get_followers_tweets.py
```

```
df_followers_tweets = pd.read_csv('follower_tweets.csv')
```

```
from tqdm import tqdm
```

```
sentiments = []
```

```
polarities = []
```

```
subjectivities = []
```

```
for i,r in tqdm(df_followers_tweets.iterrows()):
```

```
    temp = get_tweet_sentiment(r['text'])
```

```
    sentiments.append(temp[0])
```

```
    polarities.append(temp[1])
```

```

    subjectivities.append(temp[2])
df_followers_tweets['sentiment'] = sentiments
df_followers_tweets['polarity'] = polarities
df_followers_tweets['subjectivity'] = subjectivities

```

```

344it [00:00, 1290.99it/s]

```

```

user_follower_dict = {}
for i,r in df_followers_tweets.groupby('user_id'):
    polarity = r['polarity'].sum()/r['polarity'].count()
    subjectivity = r['subjectivity'].sum()/r['polarity'].count()
    user_follower_dict[i]=[polarity, subjectivity]

```

```

user_follower_dict

```

```

users_list = []
followers_list = []
for k in user_follower_dict.keys():
    users_list.append(user_polarity[k])
    followers_list.append(user_follower_dict[k])
same_polarities = different_polarities = 0

```

```

for i in range(len(users_list)):
    if users_list[i][0]>0.0 and followers_list[i][0]>0.0:
        same_polarities+=1
    elif users_list[i][0]==0.0 and followers_list[i][0]==0.0:
        same_polarities+=1
    elif users_list[i][0]<0.0 and followers_list[i][0]<0.0:
        same_polarities+=1
    else:
        different_polarities+=1

```

```

# Pie chart, where the slices will be ordered and plotted counter-clockwise:
labels = ['Same Polarities', 'Different Polarities']
sizes = [same_polarities, different_polarities]
explode = (0, 0.1)

```

```

fig1, ax1 = plt.subplots()
ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',
        shadow=True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title("Difference in opinion between a user and its followers")
plt.savefig('plot_users_followers_polarities.png')
plt.show()

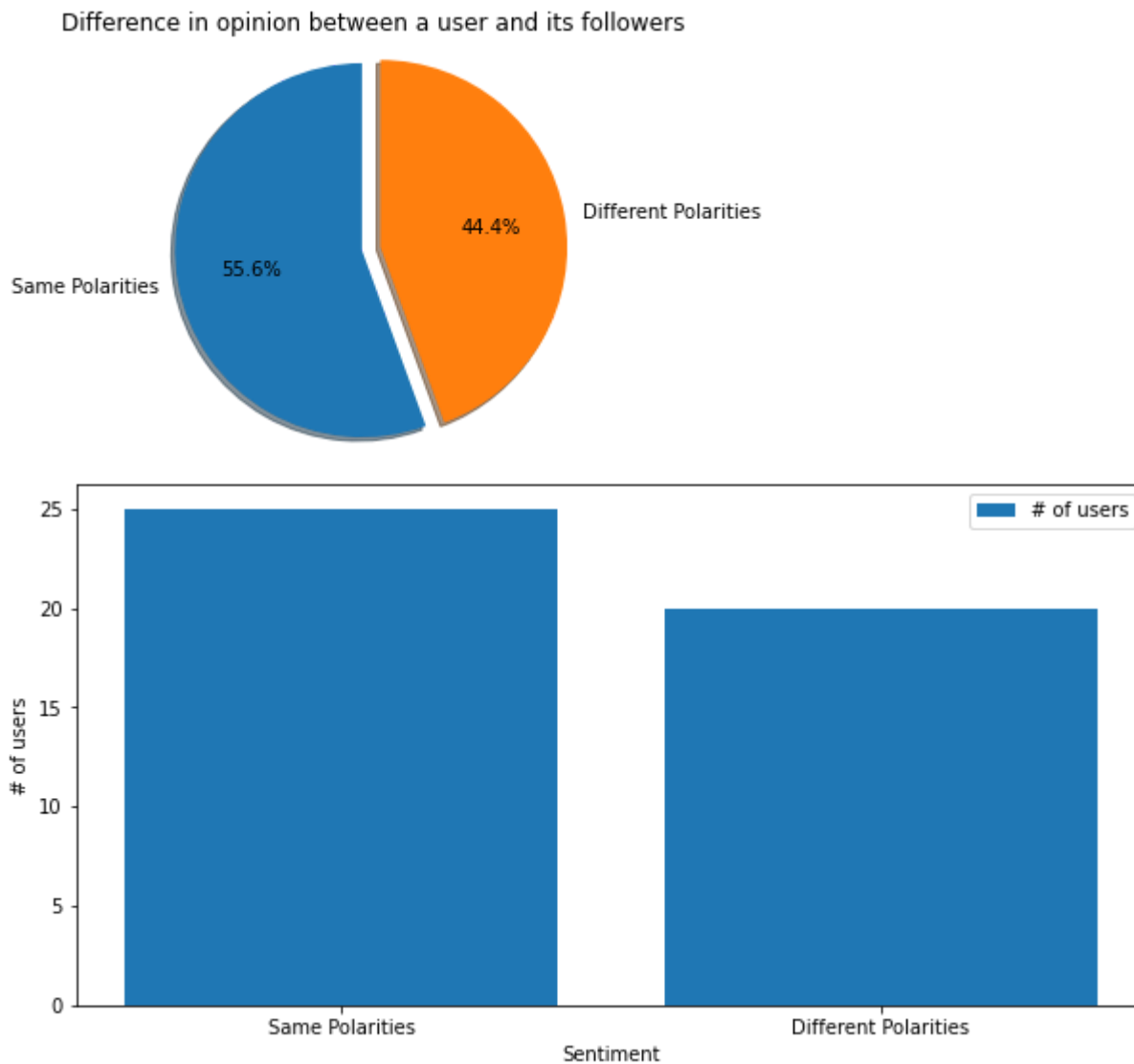
```

```

plt.figure(figsize=(10,5))
plt.bar(labels,sizes, label="# of users")
plt.xlabel("Sentiment")
plt.ylabel("# of users")

```

```
plt.legend()
plt.show()
```



```
follower_count_dict = {}
file = open('./list_followers.txt', "r")
lines = file.readlines()
for i in tqdm(range(0, len(lines), 2)):
    user, followers = lines[i], eval(lines[i+1])
    followers = followers[:20]
    follower_count_dict[int(user)] = [0, len(followers)]

for i, r in df_followers_tweets.groupby('user_id'):
    follower_count_dict[i][0] = r.shape[0]

no_tweet_by_followers = 0

for i in follower_count_dict.keys():
    if follower_count_dict[i][0] == 0:
        no_tweet_by_followers += 1
```



```
# friend_count_dict

# Pie chart, where the slices will be ordered and plotted counter-clockwise:
labels = ['Tweeted', 'Did not Tweet']
sizes = [len(list(follower_count_dict.keys()))-no_tweet_by_followers, no_tweet_by_followers]
explode = (0, 0.1)

fig1, ax1 = plt.subplots()
ax1.pie(sizes, explode=explode, labels=labels, autopct='%1.1f%%',
        shadow=True, startangle=90)
ax1.axis('equal') # Equal aspect ratio ensures that pie is drawn as a circle.
plt.title("Difference in the number of users with followers who tweeted regarding the same ha
plt.savefig('plot_users_followers_tweeted.png')
plt.show()

plt.figure(figsize=(10,5))
plt.bar(labels,sizes, label="# of users")
# plt.xlabel("Sentiment")
# plt.ylabel("# of users")
plt.legend()
plt.show()
```



100% | 53/53 [00:00&lt;00:00, 700.70it/s]

Difference in the number of users with followers who tweeted regarding the same hashtag



## ▼ Report



The above analysis shows a wide range of factor to compare and contrast the users, their followers and their friends.

But this is not just it. A lot more could be done with the provided data. I fell short on time, else Latent Dirichlet Allocation (LDA) could be used for topic modelling the users' tweets and we can get an in-depth insight into the tweets. Moreover, I have used TextBlob library of Python for sentiment analysis. Rather a better way to do it was writing a model for sentiment analysis from scratch where Long Short Term Memory(LSTM) could be used and fine-tuned to our necessity.

