



Shri Vile Parle Kelavani Mandal's

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Sentiment Analysis of Tweets during FIFA 2022

Web Intelligence Laboratory

By

Jazib Dawre	60004190054
Junaid Girkar	60004190057
Kanaad Deshpande	60004190058

Guide(s):

Prof. Sindhu Nair

Lecturer, DJSCE



Experiment 8

Date of Performance : 20/04/23

SAP ID: 60004190057

Div: A

Date of Submission: 27/04/23

Name : Junaid Girkar

Batch : A2

Aim of Experiment

Apply analytics to social media activity (Using FB, Twitter, Instagram or any social media dataset)

Theory / Algorithm / Conceptual Description

Social Media Analytics

Social media analytics is the ability to gather and find meaning in data gathered from social channels to support business decisions — and measure the performance of actions based on those decisions through social media.

Social media analytics is broader than metrics such as likes, follows, retweets, previews, clicks, and impressions gathered from individual channels. It also differs from reporting offered by services that support marketing campaigns such as LinkedIn or Google Analytics.

Social media analytics uses specifically designed software platforms that work similarly to web search tools. Data about keywords or topics is retrieved through search queries or web 'crawlers' that span channels. Fragments of text are returned, loaded into a database, categorized and analysed to derive meaningful insights.

Social media analytics includes the concept of social listening. Listening is monitoring social channels for problems and opportunities. Social media analytics tools typically incorporate listening into more comprehensive reporting that involves listening and performance analysis.



Social media analytics helps companies address these experiences and use them to:

1. Spot trends related to offerings and brands.
2. Understand conversations — what is being said and how it is being received.
3. Derive customer sentiment towards products and services.
4. Gauge response to social media and other communications.
5. Identify high-value features for a product or service.
6. Uncover what competitors are saying and its effectiveness.
7. Map how third-party partners and channels may affect performance.

Key capabilities of effective social media analytics

Natural language processing and machine learning technologies identify entities and relationships in unstructured data — information not pre-formatted to work with data analytics. Virtually all social media content is unstructured. These technologies are critical to deriving meaningful insights.

Segmentation is a fundamental need in social media analytics. It categorizes social media participants by geography, age, gender, marital status, parental status and other demographics. It can help identify influencers in those categories. Messages, initiatives and responses can be better tuned and targeted by understanding who is interacting on key topics.

Behavior analysis is used to understand the concerns of social media participants by assigning behavioral types such as user, recommender, prospective user and detractor. Understanding these roles helps develop targeted messages and responses to meet, change or deflect their perceptions.

Sentiment analysis measures the tone and intent of social media comments. It typically involves natural language processing technologies to help understand entities and relationships to reveal positive, negative, neutral or ambivalent attributes.

Share of voice analyzes prevalence and intensity in conversations regarding brand, products, services, reputation and more. It helps determine key issues and important topics. It also helps classify discussions as positive, negative, neutral or ambivalent.

Clustering analysis can uncover hidden conversations and unexpected insights. It makes associations between keywords or phrases that appear together frequently and derives new topics, issues and opportunities. The people that make baking soda, for example, discovered new uses and opportunities using clustering analysis.

Dashboards and visualization charts, graphs, tables and other presentation tools summarize and share social media analytics findings — a critical capability for communicating and acting on what has been learned. They also enable users to grasp meaning and insights more quickly and look deeper into specific findings without advanced technical skills.

FIFA Tweets Dataset

Football is one of the most loved sports worldwide. The FIFA World Cup, a global football sporting event that takes place every four years, is in Qatar this year. This dataset contains 30,000 tweets from the first day of the FIFA World Cup 2022.

Data Collection

The dataset was created using the Snsrape and the cardiffnlp/twitter-roberta-base-sentiment-latest model in Hugging Face Hub.

Data Preprocessing

The dataset includes tweets in English containing the hashtag #WorldCup2022. For data preprocessing, we used a tokenizer for the cardiffnlp/twitter-roberta-base-sentiment-latest model and the following function:

```
def preprocess(text):  
    new_text = []  
    for t in text.split(" "):  
        t = '@user' if t.startswith('@') and len(t) > 1 else t  
        t = 'http' if t.startswith('http') else t  
        new_text.append(t)  
    return " ".join(new_text)
```

Data Storage

The collected tweets have been consolidated into a single dataset & shared as a Comma Separated Values file, "fifa_world_cup_2022_tweets.csv".

Content

The dataset contains as following columns:

1. Date Created
2. Number of Likes
3. Source of Tweet
4. Tweet
5. Sentiment

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Experiment 8

April 18, 2023

```
[ ]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import datetime

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

/kaggle/input/fifa-world-cup-2022-tweets/fifa_world_cup_2022_tweets.csv

[ ]: df = pd.read_csv("/kaggle/input/fifa-world-cup-2022-tweets/
↳fifa_world_cup_2022_tweets.csv")
```

0.1 Exploratory Data Analysis

```
[ ]: df.head()
```

	Unnamed: 0	Date Created	Number of Likes	\
0	0	2022-11-20 23:59:21+00:00	4	
1	1	2022-11-20 23:59:01+00:00	3	
2	2	2022-11-20 23:58:41+00:00	1	
3	3	2022-11-20 23:58:33+00:00	1	
4	4	2022-11-20 23:58:28+00:00	0	

	Source of Tweet	Tweet	\
0	Twitter Web App	What are we drinking today @TucanTribe \n@MadB...	
1	Twitter for iPhone	Amazing @CanadaSoccerEN #WorldCup2022 launch ...	
2	Twitter for iPhone	Worth reading while watching #WorldCup2022 htt...	
3	Twitter Web App	Golden Maknae shinning bright\n\nhttps://t.co/...	
4	Twitter for Android	If the BBC cares so much about human rights, h...	

	Sentiment
0	neutral
1	positive
2	positive
3	positive

```

4 negative

[ ]: df['Date Created'] = pd.to_datetime(df['Date Created'], format='%Y-%m-%d %H:%M:
    %S')

df.info( )

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 22524 entries, 0 to 22523
Data columns (total 6 columns):
#   Column                Non-Null Count  Dtype
---  ---
0   Unnamed: 0             22524 non-null  int64
1   Date Created           22524 non-null  datetime64[ns, UTC]
2   Number of Likes        22524 non-null  int64
3   Source of Tweet        22524 non-null  object
4   Tweet                  22524 non-null  object
5   Sentiment              22524 non-null  object
dtypes: datetime64[ns, UTC](1), int64(2), object(3)
memory usage: 1.0+ MB

[ ]: df.shape

[ ]: (22524, 6)

[ ]: df.isnull().sum( )

[ ]: Unnamed: 0          0
     Date Created        0
     Number of Likes     0
     Source of Tweet     0
     Tweet               0
     Sentiment           0
     dtype: int64

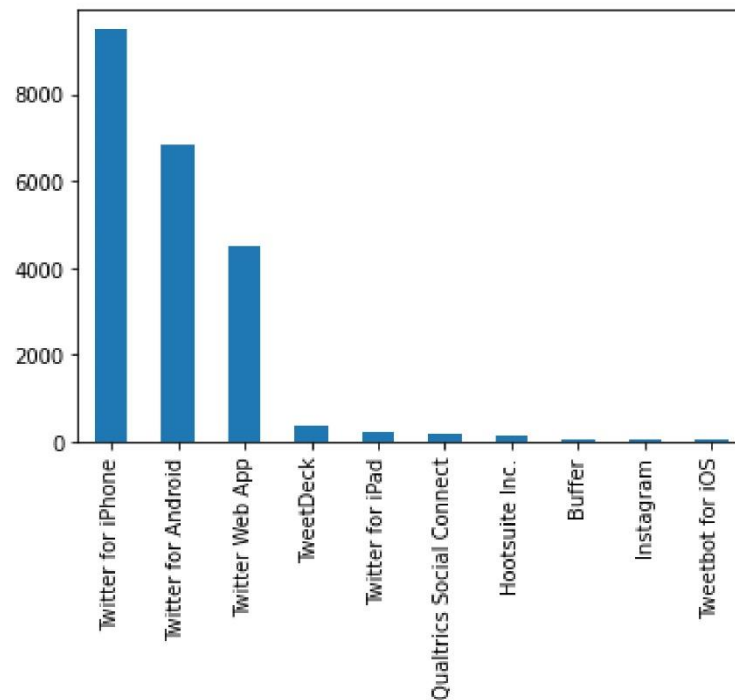
[ ]: df['Source of Tweet'].value_counts().head(10)

[ ]: Twitter for iPhone      9507
     Twitter for Android     6820
     Twitter Web App         4505
     TweetDeck               386
     Twitter for iPad        240
     Qualtrics Social Connect 165
     Hootsuite Inc.          146
     Buffer                   71
     Instagram               62
     Tweetbot for iOS        52
     Name: Source of Tweet, dtype: int64

```

```
[ ]: source_t = df['Source of Tweet'].value_counts()
      source_t.head(10).plot.bar()

      plt.show()
```



So turns out iphone users tweet more about the world cup, followed by android

```
[ ]: most_liked = df.sort_values(by=['Number of Likes'], ascending=False)
      most_liked.head(10)
```

```
[ ]:      Unnamed: 0      Date Created  Number of Likes  \
1287      1287  2022-11-20  19:39:11+00:00      316867
17359     17359  2022-11-20  12:41:31+00:00      31517
21267     21267  2022-11-20  06:17:00+00:00      20016
528        528  2022-11-20  22:20:55+00:00       5847
20073     20073  2022-11-20  09:05:12+00:00       5555
15332     15332  2022-11-20  14:57:02+00:00       5484
20101     20101  2022-11-20  09:02:09+00:00       5430
```

```

212      212 2022-11-20 23:06:08+00:00      3172
1635      1635 2022-11-20 19:00:00+00:00      2428
5761      5761 2022-11-20 16:23:30+00:00      2370

```

```

Source of Tweet \
1287      Twitter for iPhone
17359     Twitter Media Studio
21267     Twitter Web App
528       TweetDeck
20073     Twitter for iPhone
15332     Twitter Web App
20101     TweetDeck
212       Twitter Media Studio
1635     Sprinklr Publishing
5761     Twitter for Android

```

```

Tweet Sentiment
1287      I can't express my gratitude and happiness for... positive
17359     Football Legend Eric Cantona reminds football ... neutral
21267     #BitKeep FootBall Carnival Main Event is co... positive
528       WATCH: BTS's #Jungkook Performs At #WorldCup20... neutral
20073     The Leo with Louis Vuitton #WorldCup2022 h... positive
15332     Need them at the World Cup opening ceremony \... neutral
20101     WATCH: #BTS's #Jungkook Shares His Support For... neutral
212       "It's Palestine..." \n\nLebanese fans refuse to... negative
1635     .@reymysterio, @fightbobby, @BiancaBelairWWE a... neutral
5761     Qatar's Goalkeeper \n\n#WorldCup2022 https://... neutral

```

0.1.1 Most Liked Tweets

```
[ ]: df.iloc[1287].Tweet
```

```
[ ]: 'I can't express my gratitude and happiness for my participation in the biggest
event of all times the World Cup in my country #Qatar with the talented Jung
kook , thank you for everything\n#WorldCup \n#WorldCup2022 \n#JUNGKOOKx FIFA
\n#JungkookAtFIFAWorldCup \n@BTS_twt \n@bts_bighit https://t.co/PebpCj9uN2'
```

```
[ ]: df.iloc[20101].Tweet
```

```
[ ]: "WATCH: #BTS's #Jungkook Shares His Support For South Korea's National Soccer
Team At The #WorldCup2022 \nhttps://t.co/SHS3bmi60F https://t.co/SqSOKFKxXe"
```

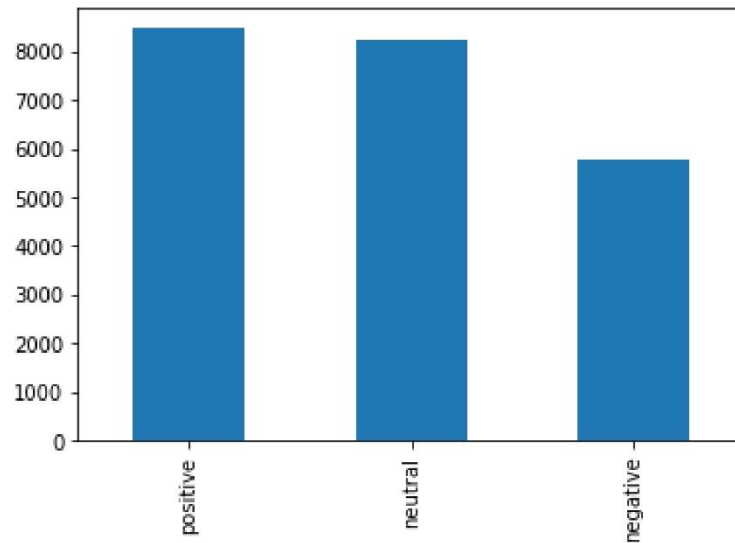
```
[ ]: df.iloc[212].Tweet
```

```
[ ]: '"It\'s Palestine..." \n\nLebanese fans refuse to be interviewed by an Israeli TV
reporter after he said that he is from \'Israel\' at #WorldCup2022 in Qatar.
https://t.co/shdWDn4ZvC'
```


0.2 Sentiment Analysis

```
[ ]: sentiment = df['Sentiment'].value_counts()
      sentiment.head().plot.bar()

      plt.show()
```

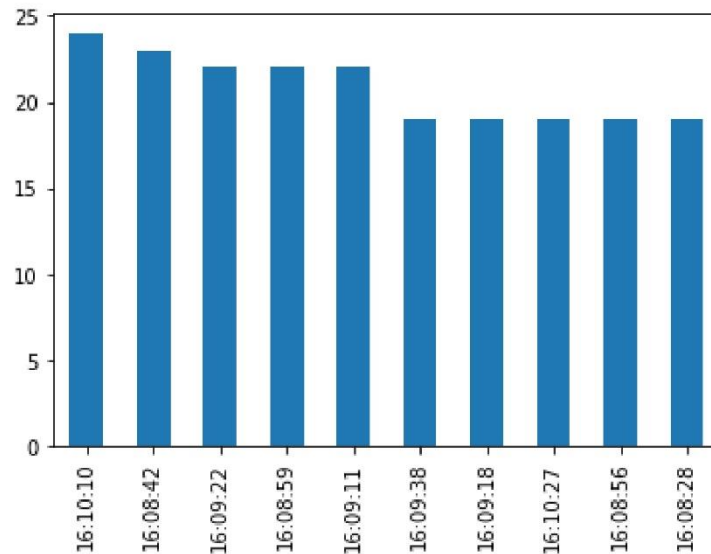


```
[ ]: df['Date Created'] = pd.to_datetime(df['Date Created']).dt.time
      df['Date Created']
```

```
[ ]: 0      23:59:21
      1      23:59:01
      2      23:58:41
      3      23:58:33
      4      23:58:28
      ...
      22519  00:00:21
      22520  00:00:03
      22521  00:00:01
      22522  00:00:00
      22523  00:00:00
      Name: Date Created, Length: 22524, dtype: object
```

```
[ ]: ppd = df['Date Created'].value_counts()
      ppd.head(10).plot.bar()

      plt.show()
```



```
[ ]: from wordcloud import WordCloud, STOPWORDS

      comment_words = ''
      stopwords = set(STOPWORDS)
      stopwords.add('https')
      stopwords.add('t')
      stopwords.add('co')

      # iterate through the csv file
      for val in df['Tweet']:

          # typecaste each val to string
          val = str(val)

          # split the value
          tokens = val.split()

          # Converts each token into lowercase
```

```
for i in range(len(tokens)):
    tokens[i] = tokens[i].lower()

comment_words += " ".join(tokens)+" "

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

# plot the WordCloud image
plt.figure(figsize = (8, 8), facecolor = None)
plt.imshow(wordcloud)
plt.axis("off")
plt.tight_layout(pad = 0)

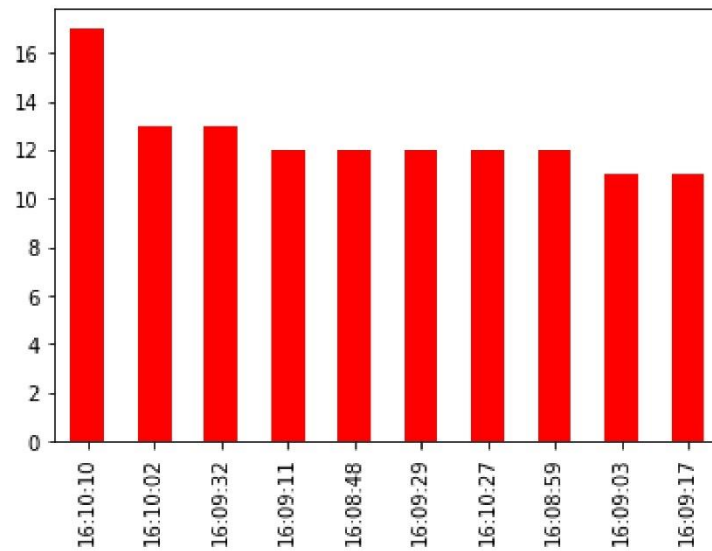
plt.show()
```



```
[ ]: neg_tweets = df[df['Sentiment'] == 'negative']
neg_tweets = neg_tweets['Date Created'].value_counts()
neg_tweets.head(10).plot.bar(color='red')

neg_tweets.head()
```

```
[ ]: 16:10:10    17
      16:10:02    13
      16:09:32    13
      16:09:11    12
      16:08:48    12
      Name: Date Created, dtype: int64
```



```
[ ]: negdf = pd.DataFrame()
negdf['negative'] = neg_tweets

postweets = df[df['Sentiment'] == 'positive']
posdf = pd.DataFrame()

posdf['positive'] = postweets['Date Created'].value_counts()
f_df = pd.DataFrame()

f_df = pd.merge(negdf, posdf, left_index=True, right_index=True)

[ ]: f_df = f_df.sort_index(ascending=True)
f_df
```

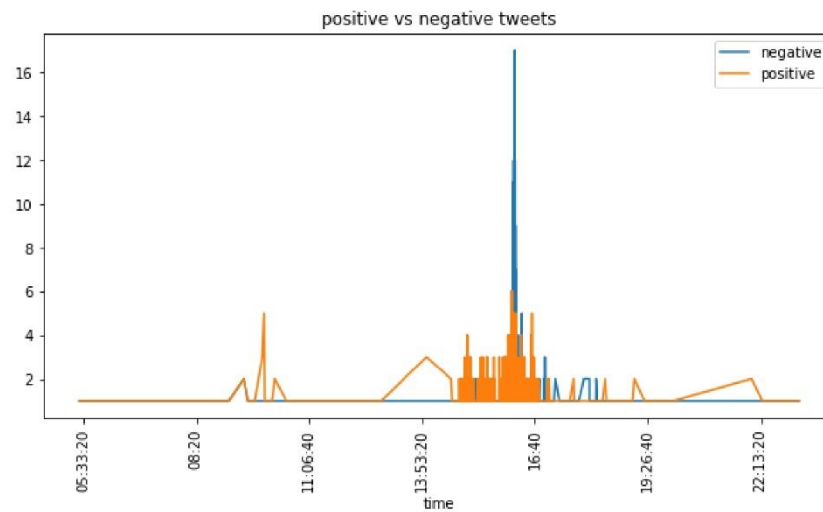
```
[ ]:      negative  positive
05:27:09         1         1
05:57:24         1         1
06:55:09         1         1
07:20:04         1         1
07:37:27         1         1
...           ...       ...
20:03:55         1         1
22:00:01         1         2
```

```
22:15:52      1      1
22:49:59      1      1
23:09:59      1      1
```

```
[1053 rows x 2 columns]
```

```
[ ]: f_df.plot(figsize=(10, 5), rot=90, title="positive vs negative tweets")
```

```
[ ]: <AxesSubplot:title={'center':'positive vs negative tweets'}, xlabel='time'>
```



Conclusion

Sentiment analysis is a powerful tool that allows generalization of user feedback for various demographics with multiple levels of granularity. The most common implementation over social media data allows companies and events to evaluate user feedback and consumer sentiment towards products and policies without conducting explicit surveys. It also removes bias that is inherent with organizational surveying by directly reading user interactions in a social web of communication.