

TITLE : DECODING EMOTION THROUGH SENTIMENT ANALYSIS OF SOCIAL MEDIA COVERSATION

OUTPUT

Upload the dataset

```
# Step 1: Upload
Dataset (Only for Google Colab)
from google.colab import files
uploaded = files.upload()
```



Choose Files elonmusk.csv.csv

- **elonmusk.csv.csv**(text/csv) - 4171382 bytes, last modified: 7/14/2020 - 100% done
Saving elonmusk.csv.csv to elonmusk.csv.csv

Load the dataset

```
# Step 2: Load the CSV
import pandas as pd
df = pd.read_csv('/content/drive/MyDrive/Elon
musk_dataset.csv/elonmusk_dataset.csv') # Replace with actual filename if
different
```

Data Exploration

```
# Step 3: Basic Info
df.head()
print("Shape:", df.shape)
print("Columns:", df.columns.tolist())
df.info()
```

	id	conversation_id	created_at	date	time	timezone	user_id	username	name	place	...	user_rt	retweet_id	reply_to	retweet_date	translate	trans_src	trans_dest	tweet_length	year	sentiment
0	12828399231796993	128283078431151618	159471683000	2020-07-14	07:28:03	UTC	44196397	elonmusk	Elon Musk	NaN	...	NaN	NaN	[[User_ID: '44196397', Username: 'elonmusk...	NaN	NaN	NaN	NaN	4	2020	positive
1	1282844872571904000	128283078431151618	1594689026000	2020-07-14	01:10:26	UTC	44196397	elonmusk	Elon Musk	NaN	...	NaN	NaN	[[User_ID: '44196397', Username: 'elonmusk...	NaN	NaN	NaN	NaN	3	2020	positive
2	128285559834492329	1282758932312584193	1594679663000	2020-07-13	22:34:13	UTC	44196397	elonmusk	Elon Musk	NaN	...	NaN	NaN	[[User_ID: '44196397', Username: 'elonmusk...	NaN	NaN	NaN	NaN	141	2020	negative
3	1282800187308572872	1282671714657157120	1594678372000	2020-07-13	22:12:52	UTC	44196397	elonmusk	Elon Musk	NaN	...	NaN	NaN	[[User_ID: '44196397', Username: 'elonmusk...	NaN	NaN	NaN	NaN	11	2020	positive
4	12828000780008023840	1282738486816964615	1594678346000	2020-07-13	22:12:26	UTC	44196397	elonmusk	Elon Musk	NaN	...	NaN	NaN	[[User_ID: '44196397', Username: 'elonmusk...	NaN	NaN	NaN	NaN	85	2020	positive

5 rows x 37 columns

```

Shape: (9286, 34)
Columns: ['id', 'conversation_id', 'created_at', 'date', 'time', 'timezone', 'user_id', 'username', 'name', 'place', 'tweet', 'mentions', 'urls', 'photos', 'replies_count', 'retweets_count', 'likes_count', 'hashtags', 'cashtags', 'link', 'retweet', 'quote_url', 'video', 'near', 'geo', 'source', 'user_rt_id', 'user_rt', 'retweet_id', 'reply_to', 'retweet_date', 'translate', 'trans_src', 'trans_dest', 'tweet_length', 'year', 'sentiment']
dtypes: int64(1), float64(33)
RangeIndex: 9286 entries, 0 to 9285
Data columns (total 34 columns):
 #   Column                Non-Null Count  Dtype  
---  --
 0   id                    9286 non-null   int64  
 1   conversation_id       9286 non-null   int64  
 2   created_at           9286 non-null   int64  
 3   date                  9286 non-null   object  
 4   time                  9286 non-null   object  
 5   timezone              9286 non-null   object  
 6   user_id               9286 non-null   int64  
 7   username              9286 non-null   object  
 8   name                  9286 non-null   object  
 9   place                 0 non-null      float64 
10  tweet                 9286 non-null   object  
11  mentions              9286 non-null   object  
12  urls                  9286 non-null   object  
13  photos                9286 non-null   object  
14  replies_count         9286 non-null   int64  
15  retweets_count        9286 non-null   int64  
16  likes_count           9286 non-null   int64  
17  hashtags              9286 non-null   object  
18  cashtags              9286 non-null   object  
19  link                  9286 non-null   object  
20  retweet               9286 non-null   bool    
21  quote_url             361 non-null    object  
22  video                 9286 non-null   int64  
23  near                  0 non-null      float64 
24  geo                   0 non-null      float64 
25  source                0 non-null      float64 
26  user_rt_id            0 non-null      float64 
27  user_rt               0 non-null      float64 
28  retweet_id            0 non-null      float64 
29  reply_to              9286 non-null   object  
30  retweet_date          0 non-null      float64 
31  translate              0 non-null      float64 
32  trans_src              0 non-null      float64 
33  trans_dest            0 non-null      float64 

```

Checking for Missing Values

Step 4: Missing values and duplicates

```
print("Missing values:\n", df.isnull().sum())
print("Duplicate rows:", df.duplicated())
```

```

Missing values:
id                    0
conversation_id       0
created_at           0
date                 0
time                 0
timezone             0
user_id              0
username             0
name                 0
place                9286
tweet                0
mentions             0
urls                 0
photos               0
replies_count        0
retweets_count       0
likes_count          0
hashtags             0
cashtags             0
link                 0
retweet              0
quote_url            8931
video                0
near                 9286
geo                  9286
source               9286
user_rt_id           9286
user_rt              9286
retweet_id           9286
reply_to             0
retweet_date         9286
translate            9286
trans_src            9286
trans_dest           9286
dtype: int64
Duplicate rows: 0

```

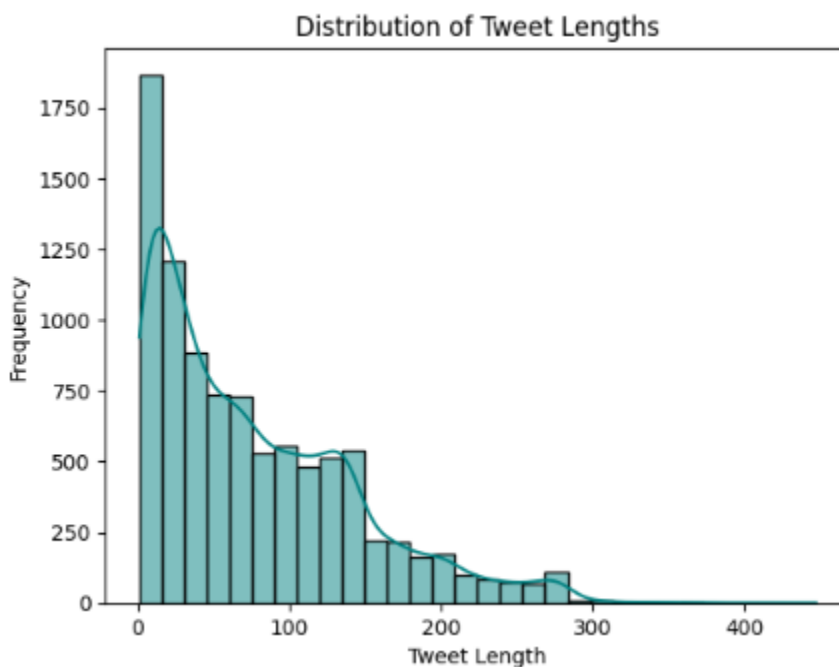
Visualize a New Features

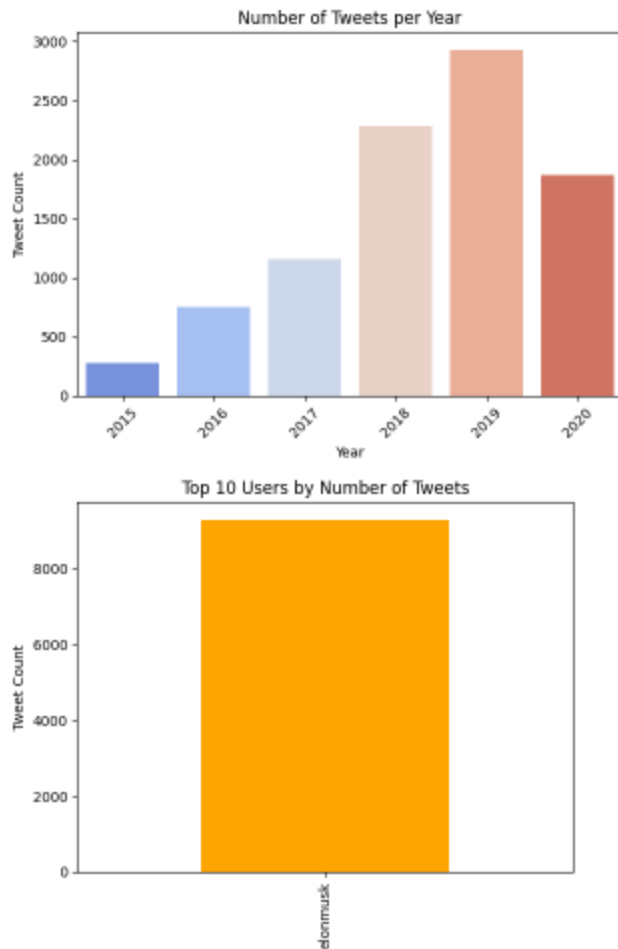
```
# Step 5: Visualizations
import seaborn as sns
import matplotlib.pyplot as plt
df['tweet_length'] = df['tweet'].astype(str).apply(len)
# Tweet length distribution

sns.histplot(df['tweet_length'], kde=True, bins=30, color='teal')

plt.title('Distribution of Tweet Lengths')
plt.xlabel('Tweet Length')
plt.ylabel('Frequency')
plt.show()

# Tweets per year
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```





Top usernames

```
top_users = df['username'].value_counts().head(10)
top_users.plot(kind='bar', color='orange')
plt.title('Top 10 Users by Number of Tweets')
plt.xlabel('Username')
plt.ylabel('Tweet Count')
plt.show()
```

Sentiment Analysis using TextBlob

```
!pip install -q textblob
from textblob import TextBlob

def get_sentiment(text):
    polarity = TextBlob(str(text)).sentiment.polarity
```

```

    if polarity > 0:
        return 'positive'
    elif polarity == 0:
        return 'neutral'
    else:
        return 'negative'

df['sentiment'] = df['tweet'].apply(get_sentiment)
print(df[['tweet', 'sentiment']].head())

```

Train-Test Split

```

# Step 7: Train-Test Split
from sklearn.model_selection import train_test_split
X = df['tweet'].astype(str)
y = df['sentiment']
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
stratify=y, random_state=42)

```

TF-IDF + Logistic Regression Pipeline

```

from sklearn.pipeline import Pipeline
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score, classification_report,
confusion_matrix

model = Pipeline([
    ('tfidf', TfidfVectorizer(max_features=5000, stop_words='english')),
    ('clf', LogisticRegression(max_iter=1000))
])

model.fit(X_train, y_train)
y_pred = model.predict(X_test)

```

Evaluation

```

# Step 9: Evaluation
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test, y_pred))

```

 Accuracy: 0.7981700753498385

Test on new tweets

```
new_tweets = [
    "Tesla's new update is amazing!",
    "I'm not happy with Twitter's new algorithm.",
    "SpaceX launch was a huge success!",
    "This is just disappointing."
]
predictions = model.predict(new_tweets)
for tweet, sentiment in zip(new_tweets, predictions):
    print(f"Tweet: {tweet}\nPredicted Sentiment: {sentiment}\n")
```

```
Tweet: Tesla's new update is amazing!
Predicted Sentiment: positive

Tweet: I'm not happy with Twitter's new algorithm.
Predicted Sentiment: positive

Tweet: SpaceX launch was a huge success!
Predicted Sentiment: positive

Tweet: This is just disappointing.
Predicted Sentiment: positive
```

Gradio Web App

```
# Step 11: Gradio Web App
!pip install -q gradio
import gradio as gr

def predict_sentiment(tweet):
    return model.predict([tweet])[0]

iface = gr.Interface(
    fn=predict_sentiment,
    inputs=gr.Textbox(lines=3, placeholder="Enter a tweet..."),
    outputs="text",
    title="Elon Musk Tweet Sentiment Analyzer",
    description="Enter a tweet related to Elon Musk to classify sentiment as Positive, Neutral, or Negative."
)
iface.launch()
```

It looks like you are running Gradio on a hosted a Jupyter notebook. For the Gradio app to work, sharing must be enabled. Automatically setting `share=True` (you can turn this off by setting `share=False` in `launch()`) explicitly).

Colab notebook detected. To show errors in colab notebook, set debug=True in launch()

* Running on public URL: <https://8962889e53d6fe7d9c.gradio.live>

This share link expires in 1 week. For free permanent hosting and GPU upgrades, run `gradio deploy` from the terminal in the working directory to deploy to Hugging Face Spaces (<https://huggingface.co/spaces>)

Elon Musk Tweet Sentiment Analyzer

Enter a tweet related to Elon Musk to classify sentiment as Positive, Neutral, or Negative.

tweet

Nilaa is good girl

output

positive

Clear

Submit

Flag