SENTIMENTAL ANALYSIS

```
import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

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- LogisticRegression
- Accuracy Score: 85%

Importing Libraries

```
import matplotlib.pyplot as plt
import seaborn as sns
from string import punctuation
from nltk.tokenize import word_tokenize
from nltk.stem import LancasterStemmer
from string import punctuation
from nltk.corpus import stopwords
from nltk.tokenize import word_tokenize
from nltk.stem import LancasterStemmer
from nltk.stem.wordnet import WordNetLemmatizer
import re
import warnings
warnings.filterwarnings('ignore')
```

→ Reading csv file

```
df = pd.read_csv('/content/training.1600000.processed.noemoticon.csv',delimiter=',', encoding='ISO-8859-1')
```

df.head()

	polarity of tweet	id of the tweet	date of the tweet	query	user	text of the tweet
0	0	1467810672	Mon Apr 06 22:19:49 PDT 2009	NO_QUERY	scotthamilton	is upset that he can't update his Facebook by
1	0	1467810917	Mon Apr 06 22:19:53 PDT 2009	NO_QUERY	mattycus	@Kenichan I dived many times for the ball. Man
2	0	1467811184	Mon Apr 06 22:19:57 PDT 2009	NO_QUERY	ElleCTF	my whole body feels itchy and like its on fire

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 53164 entries, 0 to 53163
Data columns (total 6 columns):
# Column Non-Null Count Dtype
```

```
0 polarity of tweet 1 53164 non-null int64
1 id of the tweet 53164 non-null int64
2 date of the tweet 53164 non-null object
3 query 53164 non-null object
4 user 53164 non-null object
5 text of the tweet 53164 non-null object
dtypes: int64(2), object(4)
memory usage: 2.4+ MB

df.isnull().sum()

polarity of tweet 0
id of the tweet 0
date of the tweet 0
query 0
user 0
text of the tweet 0
dtype: int64
```

→ Simplify the data

```
df.columns=['sentiment','id','date','query','username','text']
df.head()
```

	sentime	ent	id	date	query	username	text
	0	0	1467810672	Mon Apr 06 22:19:49 PDT 2009	NO_QUERY	scotthamilton	is upset that he can't update his Facebook by
	1	0	1467810917	Mon Apr 06 22:19:53 PDT 2009	NO_QUERY	mattycus	@Kenichan I dived many times for the ball. Man
	2	n	1467811184	Mon Apr 06 22-19-57	NO OHERY	FILECTE	my whole body feels
df.sh	ape						
	(53164, 6)						

→ Analysis

```
df['sentiment'].value_counts()
     0     53164
     Name: sentiment, dtype: int64

import seaborn as sns
import matplotlib.pyplot as plt

sns.histplot(df['sentiment'],kde=True)
```

dropping unnecessary columns

```
df=df.drop(columns=['query'])
```

df.head()

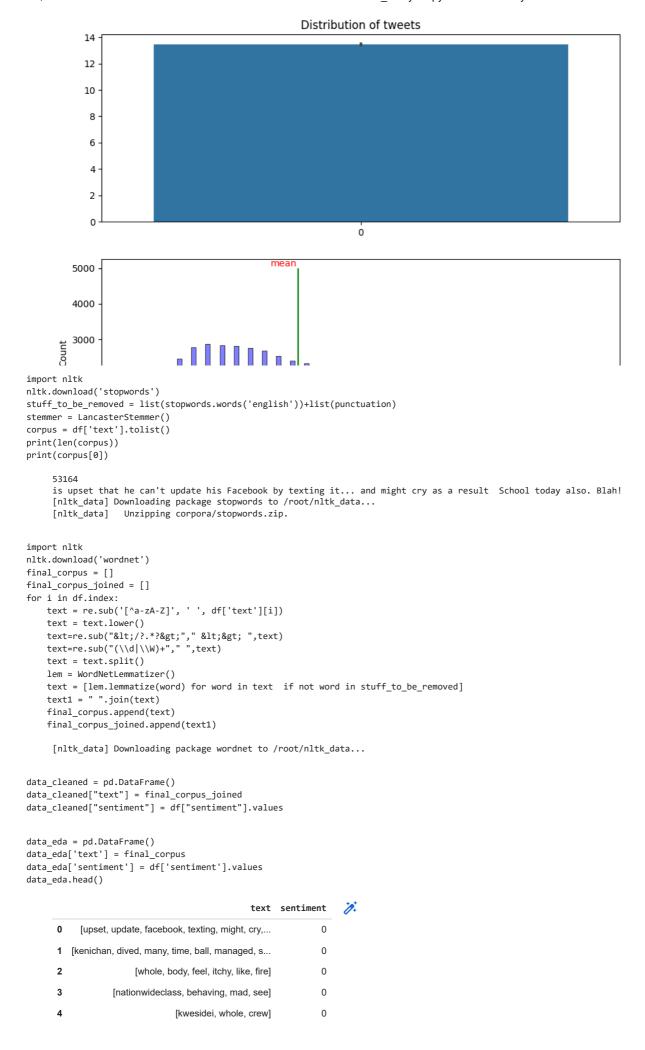
	sentiment	id	date	username	text
0	0	1467810672	Mon Apr 06 22:19:49 PDT 2009	scotthamilton	is upset that he can't update his Facebook by $\hfill \dots$
1	0	1467810917	Mon Apr 06 22:19:53 PDT 2009	mattycus	@Kenichan I dived many times for the ball. Man
2	0	1467811184	Mon Apr 06 22:19:57 PDT 2009	ElleCTF	my whole body feels itchy and like its on fire
^	^	1107011100	Mon Apr 06 22:19:57 PDT	17 11	@nationwideclass no, it's not behaving at

texts = df['text']

```
text_lens = [len(t.split()) for t in texts.values]
len_mean = np.mean(text_lens)
```

- EDA

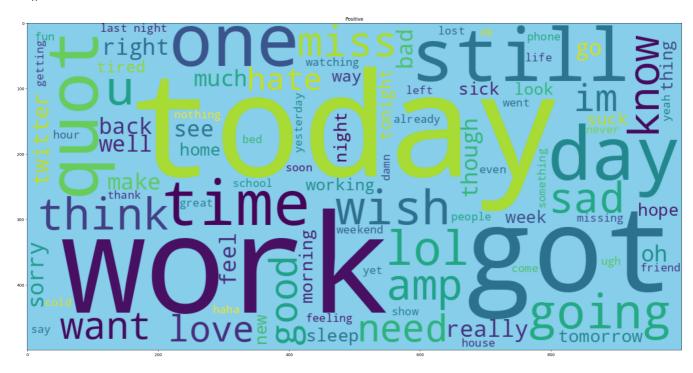
```
fig, axes = plt.subplots(2,1, figsize=(10, 8))
axes[0].set_title('Distribution of tweets')
sns.barplot(text_lens, ax=axes[0])
sns.histplot(text_lens,bins=100, kde=True, ax=axes[1],color='blue')
axes[1].vlines(len_mean, 0, 5000, color = 'g')
plt.annotate("mean", xy=(len_mean, 5000), xytext=(len_mean-2, 5050),color='r')
plt.show()
```



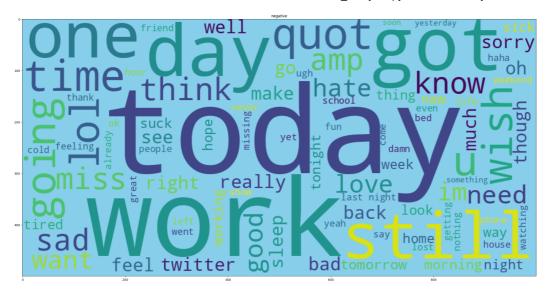
```
positive = data_eda[data_eda['sentiment'] == 4]
positive_list = positive['text'].tolist()
negative = data_eda[data_eda['sentiment'] == 0]
negative_list = negative['text'].tolist()

positive_all = " ".join([word for sent in positive_list for word in sent ])
negative_all = " ".join([word for sent in negative_list for word in sent ])
```

Word cloud positive data



Word cloud negative data



▼ TFIDF for sentiment analysis

```
from sklearn.feature_extraction.text import TfidfVectorizer
tfidf = TfidfVectorizer()
xt = tfidf.fit_transform(data_cleaned['text'])
y = data_cleaned['sentiment']
```

→ Train Test Split

→ LogisticRegression

```
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
from sklearn.linear_model import LogisticRegression

lr = LogisticRegression()
lr.fit(X_train,y_train)

* LogisticRegression
```

→ Accuracy Score: 85%

LogisticRegression()

```
y_train_pred = lr.predict(X_train)
y_test_pred = lr.predict(X_test)
accuracy_score(y_train,y_train_pred)*100
```

85.72699464659102

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