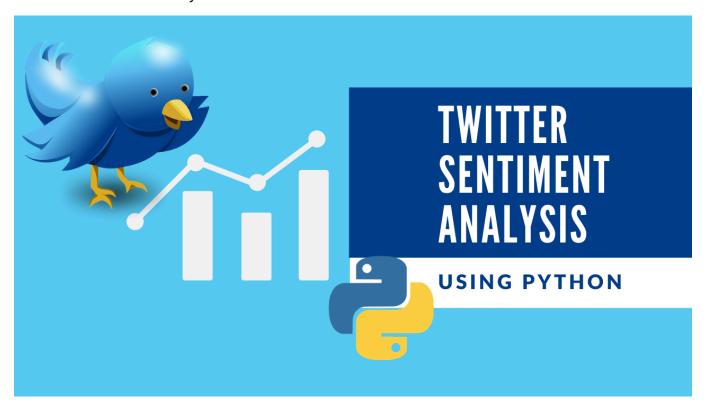
Twitter Sentiment Analysis



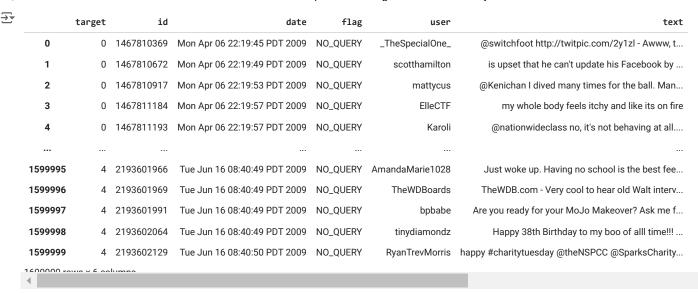
Importing libraries

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from nltk.corpus import stopwords
from wordcloud import Wordcloud
from nltk.stem.porter import PorterStemmer
from sklearn.feature_extraction.text import TfidfVectorizer
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LogisticRegression
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
from sklearn.metrics import confusion_matrix, classification_report
import re
```

Loading Dataset

```
columns_names = ['target', 'id', 'date', 'flag', 'user', 'text']
```

twitter_data=pd.read_csv('/kaggle/input/sentiment140/training.1600000.processed.noemoticon.csv',names=columns_names, encoding = 'ISO-8859-1'
twitter_data



Exploring and Pre-Processing

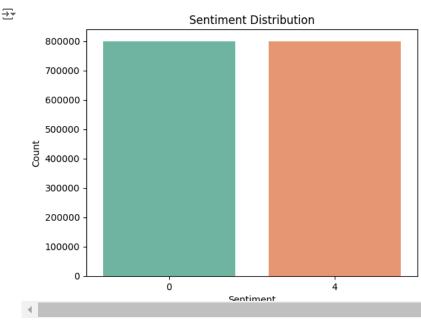
Exploratory Data Analysis

```
print(twitter_data.head())
print("\n\n")
print(twitter_data.describe())
print("\n\n")
print(twitter_data.info())
print("\n\n")
print(twitter_data.shape)
                                                                flag \
        target
     0
               1467810369
                            Mon Apr 06 22:19:45 PDT 2009
                                                           NO OUERY
     1
               1467810672
                            Mon Apr 06 22:19:49 PDT 2009
                                                           NO QUERY
     2
               1467810917
                            Mon Apr 06 22:19:53 PDT 2009
     3
               1467811184
                            Mon Apr 06 22:19:57 PDT 2009
                                                           NO QUERY
               1467811193 Mon Apr 06 22:19:57 PDT 2009
     4
                                                           NO_QUERY
                   user
     0
                         @switchfoot <a href="http://twitpic.com/2y1zl">http://twitpic.com/2y1zl</a> - Awww, t...
        TheSpecialOne
     1
          scotthamilton \, is upset that he can't update his Facebook by \dots
     2
               mattycus
                         @Kenichan I dived many times for the ball. Man...
     3
                ElleCTF
                           my whole body feels itchy and like its on fire
     4
                 Karoli @nationwideclass no, it's not behaving at all....
                  target
                                     id
     count 1.600000e+06 1.600000e+06
            2.000000e+00
                          1.998818e+09
     mean
     std
            2.000001e+00 1.935761e+08
     min
            0.000000e+00 1.467810e+09
     25%
            0.000000e+00
                          1.956916e+09
     50%
            2.000000e+00 2.002102e+09
     75%
            4.000000e+00 2.177059e+09
            4.000000e+00 2.329206e+09
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 1600000 entries, 0 to 1599999
     Data columns (total 6 columns):
         Column Non-Null Count
          target 1600000 non-null
      0
                                    int64
                                    int64
          id
                  1600000 non-null
          date
                  1600000 non-null
                                    object
          flag
                  1600000 non-null
      3
                                    object
          user
                  1600000 non-null
                                    object
          text
                  1600000 non-null
                                    object
     dtypes: int64(2), object(4)
```

```
memory usage: 73.2+ MB
None

(1600000, 6)

sns.countplot(x='target', data=twitter_data, palette='Set2')
plt.title('Sentiment Distribution')
plt.xlabel('Sentiment')
plt.ylabel('Count')
plt.show()
```



Data Cleaning

```
print(twitter_data.isnull().sum())
print("\n\n")
print(twitter_data.duplicated().sum())
print("\n\n")
twitter_data = twitter_data.drop_duplicates()
→ target
               0
     id
     date
               0
     flag
               0
     user
     text
     dtype: int64
     0
```

```
print(twitter_data['target'].value_counts())
print("\n\n")
twitter_data.replace({'target': {4: 1}}, inplace=True)
print("\n\n")
print(twitter_data['target'].value_counts())

target
0 800000
4 800000
Name: count, dtype: int64
```

```
target
0 800000
1 800000
Name: count, dtype: int64
```

Importing Stop Words

Using PorterStemmer for Word Stemming

```
port_stem = PorterStemmer()
```

Stemming

```
def stemming(content):
    stemmed_content = re.sub('[^a-zA-Z]',' ', content)
    stemmed_content = stemmed_content.lower()
    stemmed_content = stemmed_content.split()
    stemmed_content = [port_stem.stem(word) for word in stemmed_content if not word in stopwords.words('english')]
    stemmed_content = ' '.join(stemmed_content)
    return stemmed_content

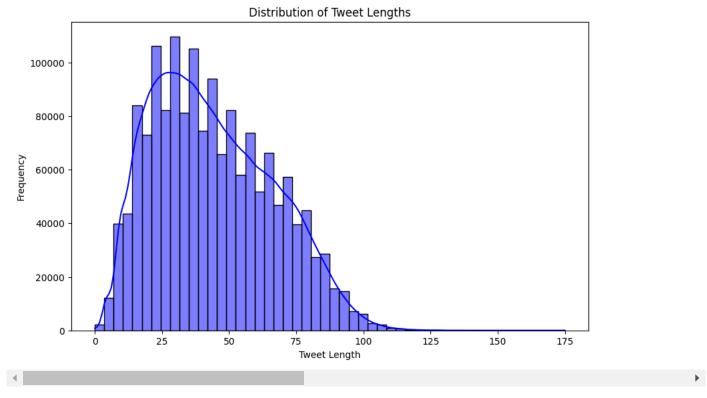
twitter_data['stemmed_content'] = twitter_data['text'].apply(stemming)
```

Analyzing Tweet Lengths

```
twitter_data['text_length'] = twitter_data['stemmed_content'].apply(len)

plt.figure(figsize=(10, 6))
sns.histplot(twitter_data['text_length'], bins=50, kde=True, color='blue')
plt.title('Distribution of Tweet Lengths')
plt.xlabel('Tweet Length')
plt.ylabel('Frequency')
plt.show()
```

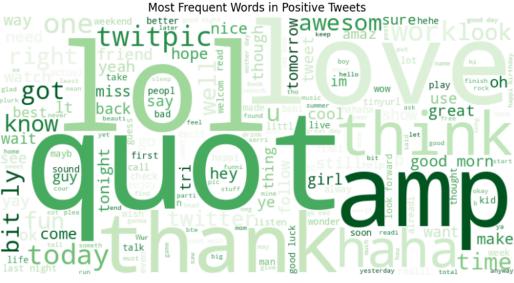
/opt/conda/lib/python3.10/site-packages/seaborn/_oldcore.py:1119: FutureWarning: use_inf_as_na option is deprecated and will be removed with pd.option_context('mode.use_inf_as_na', True):

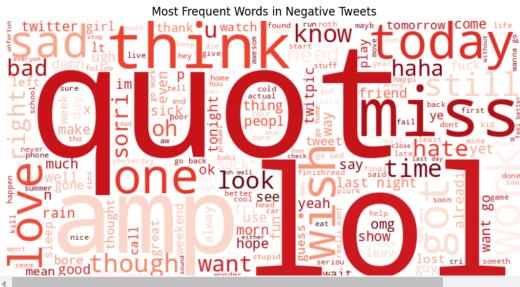


Generating Word Clouds for Positive and Negative Tweets

```
positive_tweets = twitter_data[twitter_data['target'] == 1]
negative_tweets = twitter_data[twitter_data['target'] == 0]
positive_words = ' '.join(positive_tweets['stemmed_content'])
negative_words = ' '.join(negative_tweets['stemmed_content'])
wordcloud_pos = WordCloud(width=800, height=400, background_color='white', colormap='Greens').generate(positive_words)
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud_pos, interpolation='bilinear')
plt.axis('off')
plt.title('Most Frequent Words in Positive Tweets')
plt.show()
wordcloud_neg = WordCloud(width=800, height=400, background_color='white', colormap='Reds').generate(negative_words)
plt.figure(figsize=(10, 6))
plt.imshow(wordcloud_neg, interpolation='bilinear')
plt.axis('off')
plt.title('Most Frequent Words in Negative Tweets')
plt.show()
```







twitter_data.head()

₹		target	id	date	flag	user	text	stemmed_content	text_length
	0	0	1467810369	Mon Apr 06 22:19:45 PDT 2009	NO_QUERY	_TheSpecialOne_	@switchfoot http://twitpic.com/2y1zl - Awww, t	switchfoot http twitpic com zl awww bummer sho	75
	1	0	1467810672	Mon Apr 06 22:19:49 PDT 2009	NO_QUERY	scotthamilton	is upset that he can't update his Facebook by	upset updat facebook text might cri result sch	65
	2	0	1467810917	Mon Apr 06 22:19:53 PDT 2009	NO_QUERY	mattycus	@Kenichan I dived many times for the ball. Man	kenichan dive mani time ball manag save rest g	53
	Î	^	4.467044404	Mon Apr 06	NO OUEDV	Ell ATE	my whole body feels itchy and like	1 1 1 07 15 150 6	•

Preparing Data for Model Training

twitter_data.drop(['text'], axis =1,inplace = True)

X = twitter_data['stemmed_content'].values

Y = twitter_data['target'].values

Splitting Data into Training and Testing Sets

```
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size = 0.01, stratify = Y, random_state = 42)
```

Vectorizing the Data (TF-IDF)

```
vectorizer = TfidfVectorizer(max_features=5000, ngram_range=(1, 2))
X_train = vectorizer.fit_transform(X_train)
X_test = vectorizer.transform(X_test)
```

Training the Logistic Regression Model

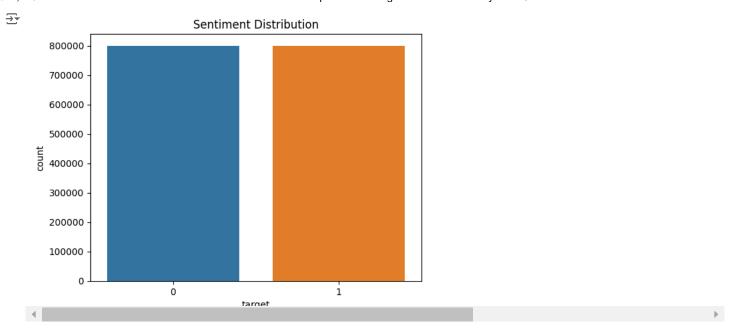
Evaluating the Model Performance

```
y_pred = model.predict(X_test)
print('Accuracy Score', accuracy_score(y_test, y_pred))
print('\nPrecision Score', precision_score(y_test, y_pred))
print('\nRecall Score', recall_score(y_test, y_pred))
print('\nf1 Score', f1_score(y_test, y_pred))
print('\nClassification Report\n',classification_report(y_test, y_pred))
Accuracy Score 0.7743125
     Precision Score 0.7588769611890999
     Recall Score 0.804125
     f1 Score 0.7808460277963222
     Classification Report
                               recall f1-score
                                                   support
               0
                       0.79
                                 0.74
                                           0.77
                                                     8000
               1
                       0.76
                                0.80
                                           0.78
                                                     8000
        accuracy
                                           0.77
                                                    16000
                       0.78
                                 0.77
                                           0.77
                                                    16000
        macro avg
     weighted avg
                       0.78
                                 0.77
                                           0.77
                                                    16000
```

Double-click (or enter) to edit

```
sns.countplot(x='target', data=twitter_data)
plt.title('Sentiment Distribution')
plt.show()

def plot_confusion_matrix(y_true, y_pred, title='Confusion Matrix'):
    cm = confusion_matrix(y_true, y_pred)
    sns.heatmap(cm, annot=True, fmt='d', cmap='Blues')
    plt.xlabel('Predicted')
    plt.ylabel('Actual')
    plt.title(title)
    plt.show()
```



Visualizing Confusion Matrix

Predicting Sentiment on New Text

```
def predict_sentiment(text):
    preprocessed_text = stemming(text)

    text_vector = vectorizer.transform([preprocessed_text])

prediction = model.predict(text_vector)

sentiment = 'Positive' if prediction[0] == 1 else 'Negative'

return sentiment

example_texts = [
    "I love this product! It's amazing.",
    "This is the worst experience I have ever had.",
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