Saurav Thakur 1810110302

# **Twitter Sentiment Analysis**

I have made a model based on twitter sentiment analysis which can determine whether the tweets are positive or negative .

### Data Preprocessing:

The Data I used here is available on kaggle.

The Link for the dataset is:

https://www.kaggle.com/kazanova/sentiment140

```
In [1]: import pandas as pd
         import numpy as np
         import re
         import matplotlib.pyplot as plt
         from spacy.lang.en.stop_words import STOP_WORDS
In [3]: df = pd.read_csv("twitter_data.csv",encoding='latin1',header=None)
In [4]: df.head()
Out[4]:
          0 0 1467810369 Mon Apr 06 22:19:45 PDT 2009 NO_QUERY _TheSpecialOne_ @switchfoot http://twitpic.com/2y1zl - Awww, t...
          1 0 1467810672 Mon Apr 06 22:19:49 PDT 2009 NO_QUERY
                                                                  scotthamilton is upset that he can't update his Facebook by
          2 0 1467810917 Mon Apr 06 22:19:53 PDT 2009 NO_QUERY mattycus @Kenichan I dived many times for the ball. Man...
          3 0 1467811184 Mon Apr 06 22:19:57 PDT 2009 NO_QUERY
                                                                                     my whole body feels itchy and like its on fire
          4 0 1467811193 Mon Apr 06 22:19:57 PDT 2009 NO_QUERY
                                                                      Karoli @nationwideclass no, it's not behaving at all....
In [5]: df = df[[5,0]]
In [6]: df.columns = ['tweets','sentiment']
In [7]: df.head()
Out[7]:
                                              tweets sentiment
          0 @switchfoot http://twitpic.com/2y1zl - Awww, t...
          1 is upset that he can't update his Facebook by ...
          2 @Kenichan I dived many times for the ball. Man...
                my whole body feels itchy and like its on fire
          4 @nationwideclass no, it's not behaving at all.... 0
In [8]: df.sentiment.value_counts()
Out[8]: 4
              800000
              800000
         Name: sentiment, dtype: int64
```

## **Exploratory Data Analysis:**

I explored the whole data and did a lot data analysis and I got following results.

### **Word Count**

```
In [10]: df["word_counts"] = df['tweets'].apply(lambda x: len(str(x).split()))
In [11]: df.head()
Out[11]:
```

		woru_counts
@switchfoot http://twitpic.com/2y1zl - Awww, t	0	19
is upset that he can't update his Facebook by $\dots$	0	21
@Kenichan I dived many times for the ball. Man	0	18
my whole body feels itchy and like its on fire	0	10
@nationwideclass no, it's not behaving at all	0	21
	@switchfoot http://twitpic.com/2y1zl - Awww, t is upset that he can't update his Facebook by @Kenichan I dived many times for the ball. Man my whole body feels itchy and like its on fire	Switchfoot http://twitpic.com/2y1zl - Awww, t  is upset that he can't update his Facebook by      Wenichan I dived many times for the ball. Man  my whole body feels itchy and like its on fire

In [12]: df["char\_counts"] = df['tweets'].apply(lambda x: len(x))
df.head()

#### Out[12]:

	tweets	sentiment	word_counts	char_counts
0	@switchfoot http://twitpic.com/2y1zl - Awww, t	0	19	115
1	is upset that he can't update his Facebook by $\dots$	0	21	111
2	@Kenichan I dived many times for the ball. Man	0	18	89
3	my whole body feels itchy and like its on fire	0	10	47
4	@nationwideclass no, it's not behaving at all	0	21	111

#### If Numeric digits are present in tweets In [21]: df["numeric\_count"] = df["tweets"].apply(lambda x : len([t for t in x.split() if t.isdigit()])) Out[22]: tweets sentiment word\_counts char\_counts avg\_word\_len stop\_words\_len hashtag\_count mention\_count numeric\_count @switchfoot http://twitpic.com/2y1zl -0 19 115 5.052632 0 0 is upset that he can't update his 4.285714 0 21 111 0 Facebook by .. @Kenichan I dived many times for the 3.944444 0 my whole body feels itchy and like its on 0 3.700000 0 0 0 @nationwideclass no, it's not behaving at 21 111 4.285714 0 **Upper Case Word Count** In [23]: df["UpperCase\_count"] = df["tweets"].apply(lambda x : len([t for t in x.split() if t.isupper() and len(x)>3])) In [24]: df.head() Out[24]: sentiment word\_counts char\_counts avg\_word\_len stop\_words\_len hashtag\_count mention\_count numeric\_count UpperCase\_coun @switchfoot 5 052632 0 http://twitpic.com/2y1zl 115 0 - Awww, t... is upset that he can't 111 4 285714 update his Facebook @Kenichan I dived 18 89 3 944444 0 0 many times for the ball. Man... my whole body feels 47 3.700000 0 0 itchy and like its on @nationwideclass no, 111 4.285714 0 it's not behaving at

# Data Cleaning:

For data cleaning I removed Urls , removed accented characters, punctuation and special characters and more.

#### Remove URLs

```
In [32]: import re
In [33]: df['urls_flag'] = df['tweets'].apply(lambda x: len(re.findall(r'(http|ftp|https)://([\w_-]+(?:(?:\.[\w_-]+)+))([\w.,@?^=%&:/~+#-]
In [34]: df['tweets'] = df['tweets'].apply(lambda x: re.sub(r'(http|ftp|https)://([\w_-]+(?:(?:\.[\w_-]+)+))([\w_-,@?^=%&:/~+#-]*[\w@?^=%&:/~+#-]*[\w]?^=%&/
In [35]: df.head()
Out[35]:
                          tweets sentiment word_counts char_counts avg_word_len stop_words_len hashtag_count mention_count numeric_count UpperCase_count u
                @switchfoot -
awww, that is a
bummer. you sh...
                                                                     115
                                                                               5.052632
                  is upset that he
            1 cannot update his facebook by...
                                                                               4.285714
                                                                                                        9
                                                                                                                                                                            0
                                          0
                                                        21
                                                                     111
                                                                                                                        0
                                                                                                                                        0
                                                                                                                                                         0
            @kenichan i
2 dived many times
for the ball. man...
                                          0
                                                        18
                                                                      89
                                                                               3.944444
                                                                                                                                                         0
                   my whole body
feels itchy and
like its on fire
                                                                               3.700000
                @nationwideclass
                 no, it is not
behaving at all...
                                          0
                                                        21
                                                                     111
                                                                               4.285714
           4
```

#### Removing Retweets

```
In [36]: df['tweets'] = df['tweets'].apply(lambda x: re.sub('RT', "", x))
```

Demoving Chariel Characters and Bunquetions

#### **Removing Accented Characters**

```
In [39]: import unicodedata
In [40]: def remove_accented_chars(x):
              x = unicodedata.normalize('NFKD', x).encode('ascii', 'ignore').decode('utf-8', 'ignore')
              return x
In [41]: x = 'Áccented text'
          remove_accented_chars(x)
Out[41]: 'Accented text'
          Removing Stop Words
In [42]: df['tweets'] = df['tweets'].apply(lambda x: " ".join([t for t in x.split() if t not in STOP_WORDS]))
In [43]: df.head()
Out[43]:
                    tweets sentiment word_counts char_counts avg_word_len stop_words_len hashtag_count mention_count numeric_count UpperCase_count
                 switchfoot -
           o awww bummer
                                                         115
                                                                  5.052632
                                                                                                                                 0
                                                                  4.285714
                  texting cry
               result schoo...
              kenichan dived
              times ball
managed save
50 rest...
                                                                  3.944444
             body feels itchy
like fire
                                                          47
                                                                                                                                                  0
                                                                  3.700000
           4 nationwideclass
                                                          111
                                                                  4.285714
```

## **Model Building:**

For Model building I used Tfidf, Logistic regression and I created pipeline which will execute tfidf and logistic regression sequentially.

#### **Model Building**

```
In [44]: X = df["tweets"]
       y = df["sentiment"]
In [45]: from sklearn.model_selection import train_test_split
In [46]: X_train,X_test,y_train,y_test = train_test_split(X,y,test_size=0.2,random_state=42)
In [47]: from sklearn.feature_extraction.text import TfidfVectorizer
        from sklearn.linear_model import LogisticRegression
In [48]: tvec = TfidfVectorizer()
       log = LogisticRegression()
In [49]: #it executes all the steps one by one
        from sklearn.pipeline import Pipeline
In [60]: # this will first create a vectorizer and then create a model
       model = Pipeline([('vectorizer',tvec),('classifier',log)])
In [61]: model.fit(X_train,y_train)
        (status=1):
        STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
        Increase the number of iterations (max iter) or scale the data as shown in:
           https://scikit-learn.org/stable/modules/preprocessing.html
        Please also refer to the documentation for alternative solver options:
          https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression
         n_iter_i = _check_optimize_result(
In [52]: from sklearn.metrics import confusion_matrix
In [53]: predictions = model.predict(X test)
```

### **Predictions:**

After Building the model I got the predictions 0 and 4 where 0 means negative and 4 means positive. I used data from my twitter account tweets for predicting the model. The model has an accuracy of 77.84%.

#### **Model Predictions**

```
In [55]: from sklearn.metrics import accuracy_score,precision_score,recall_score

In [56]: print("Accuracy : ",accuracy_score(predictions,y_test))
    print("Precision : ",precision_score(predictions,y_test,average='weighted'))
    print("Recall : ",recall_score(predictions,y_test,average='weighted'))

Accuracy : 0.77846875
Precision : 0.7793796503192831
Recall : 0.77846875
```

#### **Predicting**

#### 0 - Negative

#### 4 - Positive

```
In [57]: example = ["I hate you"] model.predict(example)

Out[57]: array([0], dtype=int64)

In [66]: model.predict(["So happy the Greatest Of All Time will meet again tonight It's gonna be a showdown Watch out Ronaldo"])

Out[66]: array([4], dtype=int64)

In [62]: model.predict(["I need to say this so people know how big of a mistake this was, I was traumatized by Human Centipede back in 200

Out[62]: array([0], dtype=int64)

In [64]: model.predict(["As cases of Covid-19 continue to rise across the country, a poll of firefighters in the Fire Department of New You will be a pray([4], dtype=int64)

In [65]: model.predict(["way too much money invested by these pharmaceuticals than to create a faulty fatal vaccine that would be financiated by a predict(["way too much money invested by these pharmaceuticals than to create a faulty fatal vaccine that would be financiated by a predict(["way too much money invested by these pharmaceuticals than to create a faulty fatal vaccine that would be financiated by a predict(["way too much money invested by these pharmaceuticals than to create a faulty fatal vaccine that would be financiated by a predict of the predict of the
```