```
In [1]:
```

```
# Mounting colab at drive
from google.colab import drive
drive.mount('/content/gdrive')
```

Mounted at /content/gdrive

#### In [2]:

```
[!unzip 'gdrive/MyDrive/Twitter Sentiment Analysis/archive.zip'
```

Archive: gdrive/MyDrive/Twitter Sentiment Analysis/archive.zip
inflating: sentiment tweets3.csv

## In [3]:

```
# Importing Libraries
import numpy as np
import pandas as pd
```

## In [4]:

```
# Importing dataset
df = pd.read_csv('sentiment_tweets3.csv')
```

#### In [5]:

df.head(15)

## Out[5]:

	Index	message to examine	label (depression result)
0	106	just had a real good moment. i misssssssss hi	0
1	217	is reading manga http://plurk.com/p/mzp1e	0
2	220	@comeagainjen http://twitpic.com/2y2lx - http:	0
3	288	@lapcat Need to send 'em to my accountant tomo	0
4	540	ADD ME ON MYSPACE!!! myspace.com/LookThunder	0
5	624	so sleepy. good times tonight though	0
6	701	@SilkCharm re: #nbn as someone already said, d	0
7	808	23 or 24�C possible today. Nice	0
8	1193	nite twitterville workout in the am -ciao	0
9	1324	@daNanner Night, darlin'! Sweet dreams to you	0
10	1332	Good morning everybody!	0
11	1368	Finally! I just created my WordPress Blog. The	0
12	1578	kisha they cnt get over u til they get out frm	0
13	1595	@nicolerichie Yes i remember that band, It was	0
14	1861	I really love reflections and shadows	0

### In [6]:

```
# Getting shape of dataset: (rows x columns)
df.shape
```

# Out[6]:

(10314, 3)

```
In [7]:
# Dropping Unwanted columns
df = df.drop('Index', axis = 1)
In [8]:
# New shape of dataset
df.shape
Out[8]:
(10314, 2)
In [9]:
# Changing column name
df.columns = ['Tweets', 'Polarity']
df.head(15)
Out[9]:
                                            Tweets Polarity
 0
        just had a real good moment. i misssssssss hi...
 1
             is reading manga http://plurk.com/p/mzp1e
                                                          0
 2
         @comeagainjen http://twitpic.com/2y2lx - http:...
                                                          0
 3
      @lapcat Need to send 'em to my accountant tomo...
                                                          0
    ADD ME ON MYSPACE!!! myspace.com/LookThunder
                                                          0
 5
                  so sleepy. good times tonight though
                                                          0
 6
       @SilkCharm re: #nbn as someone already said, d...
                                                          0
 7
                     23 or 24ï¿1/2C possible today. Nice
                                                          0
                nite twitterville workout in the am -ciao
 8
                                                          0
 9
         @daNanner Night, darlin'! Sweet dreams to you
                                                          0
                            Good morning everybody!
                                                          0
10
11
         Finally! I just created my WordPress Blog. The...
                                                          0
12
           kisha they cnt get over u til they get out frm...
                                                          0
13
        @nicolerichie Yes i remember that band, It was...
                                                          0
14
                   I really love reflections and shadows
In [10]:
# Checking for missing values
df.isna().sum()
Out[10]:
Tweets
               0
Polarity
               0
dtype: int64
In [11]:
# We see that there are no missing values thus we carry on with the preprocessing
# Preprocess data
```

import re
import nltk

import spacy

from nltk.corpus import stopwords

nltk.download('stopwords')

from wordcloud import WordCloud, STOPWORDS

```
# Convert Text to lower case
df['Tweets'] = df['Tweets'].str.strip().str.lower()
# Initialize Spacy 'en' model
nlp = spacy.load('en', disable = ['parser', 'ner'])
# Remove usernames and links
def remove usernames links(tweet):
   tweet = re.sub('@[^\s]+','',tweet)
   tweet = re.sub('http[^\s]+','',tweet)
   return tweet
df['Tweets'] = df['Tweets'].apply(remove_usernames_links)
# Remove punctuations
def remove punctuation(tweet):
 tweet = re.sub('[^\w\s]', '', tweet)
 return tweet
df['Tweets'] = df['Tweets'].apply(remove punctuation)
# Remove Stopwords
stop = stopwords.words('english')
df['Tweets'] = df['Tweets'].apply(lambda x: " ".join(x for x in x.split(' ') if x not in
stop))
# Lemmatization
def space(comment):
 doc = nlp(comment)
 return " ".join([token.lemma for token in doc])
df['Tweets'] = df['Tweets'].apply(space)
df.head(15)
```

[nltk\_data] Downloading package stopwords to /root/nltk\_data...
[nltk\_data] Unzipping corpora/stopwords.zip.

# Out[11]:

	Tweets	Polarity
0	real good moment misssssssss much	0
1	read manga	0
2		0
3	need send -PRON- accountant tomorrow oddly b	0
4	add myspace myspacecomlookthunder	0
5	sleepy good time tonight though	0
6	nbn someone already say fiber home mean leas	0
7	23 24ï½c possible today nice	0
8	nite twitterville workout ciao	0
9	night darlin sweet dream	0
10	good morning everybody	0
11	finally create wordpress blog there s already	0
12	kisha cnt get u til get frm u remember ur top	0
13	yes remember band awesome please reply	0
14	really love reflection shadow	0

```
In [12]:
```

```
# Outputting last 15 rows with df.tail()
```

```
df.tail(15)
Out[12]:
```

1

1

```
Tweets Polarity
10299
          engage physical activity decrease people chanc...
10300
                    depression -PRON- be always u darling
```

10302 one 20 us child teen anxiety depression 1 10303 -PRON- be go see next week chicago -PRON- be... 1

-PRON- be suffer depression -PRON- be thankful...

10304 reagan bush 43 trump gop use maga slogan r... 1 10305 rt depression could improve vitamin deficiency... 1

10306 actor symptom adult depression naked gteen por...

10307 bharatu may book court law provoke s... 10308 many sufferer depression be not sad feel nothi... 1

10309 depression g herbo mood -PRON- be do stress pe... 10310 depression succumb brain make feel like -PRON-... 1

10311 ketamine nasal spray show promise depression s...

do not mistake bad day depression everyone -10312 1 PRON-10313 0

```
In [13]:
```

10301

```
# Again checking for invalid rows
(df['Tweets'] == '').sum()
```

# Out[13]:

25

### In [14]:

```
df.shape
```

## Out[14]:

(10314, 2)

#### In [15]:

```
# Checking for rows containing just numbers
def remove_numbers(tweet):
 if tweet.isdigit():
   tweet = ''
 return tweet
df['Tweets'] = df['Tweets'].apply(remove numbers)
(df['Tweets'] == '').sum()
```

#### Out[15]:

31

# In [16]:

```
# Thus we see that we have 25 null strings + 6 more that had rows with just numbers and h
ence we must drop them
# This will change the shape of our dataset to (10283, 2)
# Dropping empty strings
```

```
nan_value = float("NaN")
df. replace('', nan_value, inplace=True)
df = df.dropna(axis = 0)
df.shape
Out[16]:
(10283, 2)
In [17]:
# Split into training and testing dataset
from sklearn.model selection import train test split
x = df['Tweets']
y = df['Polarity']
X train, X test, Y train, Y test = train test split(x, y, stratify = y, test size = 0.2,
random state = 42, shuffle = True)
In [18]:
print('Xtrain: ', X_train.shape, 'Ytrain: ', Y_train.shape)
print('Xtest: ', X_test.shape, 'Ytest: ', Y test.shape)
Xtrain: (8226,) Ytrain: (8226,)
Xtest: (2057,) Ytest: (2057,)
In [20]:
print(Y train.value counts())
print(Y test.value counts())
0
  6380
    1846
1
Name: Polarity, dtype: int64
  1595
1
     462
Name: Polarity, dtype: int64
In [21]:
from sklearn.feature extraction.text import CountVectorizer
# Vectorize text reviews to encoded numbers
vector = CountVectorizer(stop words = 'english')
X_train = vector.fit_transform(X_train).toarray()
X test = vector.transform(X test).toarray()
In [22]:
# Model Generation
# Multinomial NB
from sklearn.naive bayes import MultinomialNB
model1 = MultinomialNB()
model1.fit(X_train, Y_train)
Out[22]:
MultinomialNB()
In [23]:
# Evaluating Training score
model1.score(X_train, Y_train)
Out[23]:
0.9912472647702407
In [24]:
```

```
# Function to generate metrics
def metrics(labels, predictions):
    true_pos, true_neg, false_pos, false neg = 0, 0, 0, 0
    for i in range(len(labels)):
        true pos += int(labels.iloc[i] == 1 and predictions[i] == 1)
        true neg += int(labels.iloc[i] == 0 and predictions[i] == 0)
        false pos += int(labels.iloc[i] == 0 and predictions[i] == 1)
        false neg += int(labels.iloc[i] == 1 and predictions[i] == 0)
    precision = true pos / (true pos + false pos)
    recall = true pos / (true pos + false neg)
    Fscore = 2 * precision * recall / (precision + recall)
    accuracy = (true pos + true neg) / (true pos + true neg + false pos + false neg)
    return true pos, true neg, false pos, false neg, accuracy, precision, recall, Fscore
In [25]:
# Testing our own calculations
predictions1 = model1.predict(X test)
true pos1, true neg1, false pos1, false neg1, a1, p1, r1, f1 = metrics(Y test, predictio
print("Accuracy: ", a1)
print("Precision: ", p1)
print("Recall: ", r1)
print("F-score: ", f1)
Accuracy: 0.9795819154107924
Precision: 0.9356846473029046
Recall: 0.9761904761904762
F-score: 0.9555084745762712
In [33]:
# Print Confusion Matrix
matrix1 = np.zeros((2, 2))
matrix1[0, 0] = int(true neg1)
matrix1[0, 1] = int(false pos1)
matrix1[1, 0] = int(false neg1)
matrix1[1, 1] = int(true pos1)
print (matrix1)
        31.]
[[1564.
 [ 11. 451.]]
In [34]:
# Pitting our values against sklearn values
from sklearn.metrics import confusion_matrix, accuracy_score, recall_score, precision_sco
re, f1 score
check matrix1 = confusion matrix(Y test, predictions1)
cm1 = pd.DataFrame(check matrix1, index=['0', '1'], columns=['0', '1'])
print(cm1)
     0
0 1564
         31
    11 451
In [35]:
chk acc1 = accuracy score(Y test, predictions1)
chk prc1 = precision score(Y test, predictions1)
chk_rec1 = recall_score(Y_test, predictions1)
chk f1 = f1 score(Y test, predictions1)
print("Sklearn Accuracy: ", chk acc1)
print("Sklearn Precision: ", chk prc1)
print("Sklearn Recall: ", chk_rec1)
print("Sklearn F-score: ", chk_f1)
```

Sklearn Accuracy 0 9795819154107924

```
Sklearn Precision: 0.9356846473029046
Sklearn Recall: 0.9761904761904762
Sklearn F-score: 0.9555084745762712
```

#### In [36]:

```
# Evaluating Testing score as output by sklearn
model1.score(X_test, Y_test)
```

# Out[36]:

0.9795819154107924

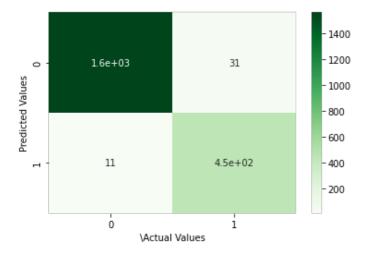
#### In [42]:

```
# Creating heatmap of confusion matrix
import seaborn as sns
import matplotlib.pyplot as plt

ax1 = sns.heatmap(check_matrix1, annot = True, cmap = 'Greens')
ax1.set_title('Seaborn Confusion Matrix for MultinomialNB \n\n')
ax1.set_xlabel('\Actual Values')
ax1.set_ylabel('Predicted Values')

ax1.set_yticklabels(['0', '1'])
ax1.set_yticklabels(['0', '1'])
```

#### Seaborn Confusion Matrix for MultinomialNB



### In [43]:

```
# Gaussian NB
from sklearn.naive_bayes import GaussianNB
model2 = GaussianNB()
model2.fit(X_train, Y_train)
```

## Out[43]:

GaussianNB()

#### In [44]:

```
# Evaluating Training score
model2.score(X_train, Y_train)
```

## Out[44]:

0.8334548991004134

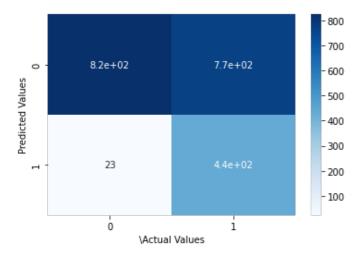
## In [45]:

```
# Testing our own calculations
predictions2 = model2.predict(X_test)
```

```
true_pos2, true_neg2, false_pos2, false_neg2, a2, p2, r2, f2 = metrics(Y_test, predictio
ns2)
print("Accuracy: ", a2)
print("Precision: ", p2)
print("Recall: ", r2)
print("F-score: ", f2)
Accuracy: 0.6140009722897424
Precision: 0.3628099173553719
Recall: 0.9502164502164502
F-score: 0.5251196172248803
In [46]:
# Print Confusion Matrix
matrix2 = np.zeros((2, 2))
matrix2[0, 0] = int(true_neg2)
matrix2[0, 1] = int(false_pos2)
matrix2[1, 0] = int(false_neg2)
matrix2[1, 1] = int(true pos2)
print (matrix2)
[[824. 771.]
[ 23. 439.]]
In [47]:
# Pitting our values against sklearn values
check matrix2 = confusion matrix(Y test, predictions2)
cm2 = pd.DataFrame(check matrix2, index=['0', '1'], columns=['0', '1'])
print(cm2)
    0
         1
 824 771
1
   23 439
In [48]:
chk_acc2 = accuracy_score(Y_test, predictions2)
chk prc2 = precision score(Y test, predictions2)
chk rec2 = recall score(Y test, predictions2)
chk f2 = f1 score(Y test, predictions2)
print("Sklearn Accuracy: ", chk acc2)
print("Sklearn Precision: ", chk prc2)
print("Sklearn Recall: ", chk rec2)
print("Sklearn F-score: ", chk f2)
Sklearn Accuracy: 0.6140009722897424
Sklearn Precision: 0.3628099173553719
Sklearn Recall: 0.9502164502164502
Sklearn F-score: 0.5251196172248803
In [49]:
# Evaluating Testing score
model2.score(X_test, Y_test)
Out[49]:
0.6140009722897424
In [50]:
ax2 = sns.heatmap(check matrix2, annot = True, cmap = 'Blues')
ax2.set title('Seaborn Confusion Matrix for MultinomialNB \n\n')
ax2.set xlabel('\Actual Values')
ax2.set_ylabel('Predicted Values')
ax2.set xticklabels(['0', '1'])
```

```
ax2.set_yticklabels(['0', '1'])
plt.show()
```

## Seaborn Confusion Matrix for MultinomialNB



## In [51]:

```
# Complement NB
from sklearn.naive_bayes import ComplementNB
model3 = ComplementNB()
model3.fit(X_train, Y_train)
```

#### Out[51]:

ComplementNB()

#### In [52]:

```
# Evaluating Training score
model3.score(X_train, Y_train)
```

#### Out[52]:

0.973741794310722

#### In [53]:

```
# Testing our own calculations
predictions3 = model3.predict(X_test)
true_pos3, true_neg3, false_pos3, false_neg3, a3, p3, r3, f3 = metrics(Y_test, predictions3)

print("Accuracy: ", a3)
print("Precision: ", p3)
print("Recall: ", r3)
print("F-score: ", f3)
```

Accuracy: 0.9470102090422946 Precision: 0.818018018018018 Recall: 0.9826839826839827 F-score: 0.8928220255653884

## In [54]:

```
# Print Confusion Matrix
matrix3 = np.zeros((2, 2))
matrix3[0, 0] = int(true_neg3)
matrix3[0, 1] = int(false_pos3)
matrix3[1, 0] = int(false_neg3)
matrix3[1, 1] = int(true_pos3)
print(matrix3)
```

```
[[1494. 101.]
```

```
In [55]:
# Pitting our values against sklearn values
check matrix3 = confusion matrix(Y test, predictions3)
cm3 = pd.DataFrame(check matrix3, index=['0', '1'], columns=['0', '1'])
print(cm3)
      0
 1494 101
1
     8 454
In [56]:
chk acc3 = accuracy score(Y test, predictions3)
chk prc3 = precision score(Y test, predictions3)
chk rec3 = recall score(Y test, predictions3)
chk f3 = f1 score(Y test, predictions3)
print("Sklearn Accuracy: ", chk acc3)
print("Sklearn Precision: ", chk prc3)
print("Sklearn Recall: ", chk_rec3)
print("Sklearn F-score: ", chk f3)
Sklearn Accuracy: 0.9470102090422946
Sklearn Precision: 0.818018018018018
Sklearn Recall: 0.9826839826839827
Sklearn F-score: 0.8928220255653884
In [57]:
# Evaluating Testing score
model3.score(X_test, Y_test)
Out[57]:
0.9470102090422946
In [58]:
ax3 = sns.heatmap(check matrix3, annot = True, cmap = 'Reds')
ax3.set title('Seaborn Confusion Matrix for MultinomialNB \n\n')
ax3.set_xlabel('\Actual Values')
ax3.set ylabel('Predicted Values')
ax3.set xticklabels(['0', '1'])
ax3.set yticklabels(['0', '1'])
plt.show()
    Seaborn Confusion Matrix for MultinomialNB
                                         1400
                                        - 1200
          1.5e+03
                           1e+02
  0
                                        - 1000
Predicted Values
                                        - 800
                                        - 600
```

- 400

- 200

4.5e+02

1

## In [59]:

// -- 77' 370

8

0

\Actual Values

```
# Bernoulli NB
from sklearn.naive bayes import BernoulliNB
model4 = BernoulliNB()
model4.fit(X_train, Y_train)
Out[59]:
BernoulliNB()
In [60]:
# Evaluating Training score
model4.score(X train, Y train)
Out[60]:
0.987478725990761
In [61]:
# Testing our own calculations
predictions4 = model4.predict(X test)
true_pos4, true_neg4, false_pos4, false_neg4, a4, p4, r4, f4 = metrics(Y_test, predictio
ns4)
print("Accuracy: ", a4)
print("Precision: ", p4)
print("Recall: ", r4)
print("F-score: ", f4)
Accuracy: 0.9708313077297035
Precision: 1.0
Recall: 0.8701298701298701
F-score: 0.93055555555556
In [62]:
# Print Confusion Matrix
matrix4 = np.zeros((2, 2))
matrix4[0, 0] = int(true_neg4)
matrix4[0, 1] = int(false pos4)
matrix4[1, 0] = int(false_neg4)
matrix4[1, 1] = int(true pos4)
print(matrix4)
[[1595.
          0.]
 [ 60. 402.]]
In [63]:
# Pitting our values against sklearn values
check matrix4 = confusion matrix(Y test, predictions4)
cm4 = pd.DataFrame(check matrix4, index=['0', '1'], columns=['0', '1'])
print(cm4)
     0
         1
0 1595
   60 402
In [64]:
chk acc4 = accuracy score(Y test, predictions4)
chk prc4 = precision score(Y test, predictions4)
chk rec4 = recall score(Y test, predictions4)
chk_f4 = f1_score(Y_test, predictions4)
print("Sklearn Accuracy: ", chk acc4)
print("Sklearn Precision: ", chk prc4)
print("Sklearn Recall: ", chk rec4)
print("Sklearn F-score: ", chk f4)
Sklearn Accuracy: 0.9708313077297035
```

```
Sklearn Precision: 1.0
Sklearn Recall: 0.8701298701298701
Sklearn F-score: 0.9305555555555556
```

## In [65]:

```
# Evaluating Testing score
model4.score(X_test, Y_test)
```

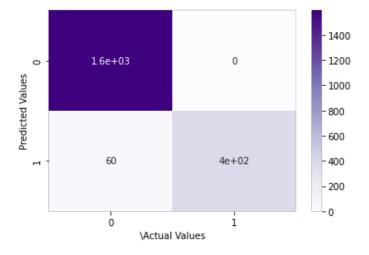
### Out[65]:

0.9708313077297035

## In [67]:

```
ax4 = sns.heatmap(check_matrix4, annot = True, cmap = 'Purples')
ax4.set_title('Seaborn Confusion Matrix for MultinomialNB \n\n')
ax4.set_xlabel('\Actual Values')
ax4.set_ylabel('Predicted Values')
ax4.set_xticklabels(['0', '1'])
ax4.set_yticklabels(['0', '1'])
```

#### Seaborn Confusion Matrix for MultinomialNB



## In [74]:

```
# Let us try to get some assumptions for each model and see
tweet1 = 'I am sad'
tweet2 = 'Today was the best day ever'
tweet3 = 'Everything will be okay'
tweet4 = 'Hello. I am amazing'
tweet5 = 'Is it sad that I am all alone'
tweet6 = 'That is sad but I am happy'
```

# In [78]:

```
# Model 1 predictions
p1 = model1.predict(vector.transform([tweet1, tweet2, tweet3, tweet4, tweet5, tweet6]))
print(p1)
print()
for index, i in enumerate(p1):
   if i == 0:
      print('Tweeter ' + str(index + 1) + ' is not depressed')
   else:
      print('Tweeter ' + str(index + 1) + ' is depressed')
```

# [1 0 0 0 1 0]

```
Tweeter 1 is depressed
Tweeter 2 is not depressed
Tweeter 3 is not depressed
```

```
Tweeter 4 is not depressed
Tweeter 5 is depressed
Tweeter 6 is not depressed
In [80]:
# Model 2 predictions
p2 = model2.predict(vector.transform([tweet1, tweet2, tweet3, tweet4, tweet5, tweet6]).t
oarray())
print(p2)
print()
for index, i in enumerate(p2):
  if i == 0:
    print('Tweeter ' + str(index + 1) + ' is not depressed')
  else:
   print('Tweeter ' + str(index + 1) + ' is depressed')
[1 \ 1 \ 1 \ 1 \ 1 \ 1]
Tweeter 1 is depressed
Tweeter 2 is depressed
Tweeter 3 is depressed
Tweeter 4 is depressed
Tweeter 5 is depressed
Tweeter 6 is depressed
In [81]:
# Model 3 predictions
p3 = model3.predict(vector.transform([tweet1, tweet2, tweet3, tweet4, tweet5, tweet6]))
print(p3)
print()
for index, i in enumerate(p3):
  if i == 0:
   print('Tweeter ' + str(index + 1) + ' is not depressed')
    print('Tweeter ' + str(index + 1) + ' is depressed')
[1 0 0 0 1 1]
Tweeter 1 is depressed
Tweeter 2 is not depressed
Tweeter 3 is not depressed
Tweeter 4 is not depressed
Tweeter 5 is depressed
Tweeter 6 is depressed
In [82]:
# Model 4 predictions
p4 = model4.predict(vector.transform([tweet1, tweet2, tweet3, tweet4, tweet5, tweet6]))
print(p4)
print()
for index, i in enumerate(p4):
  if i == 0:
   print('Tweeter ' + str(index + 1) + ' is not depressed')
  else:
    print('Tweeter ' + str(index + 1) + ' is depressed')
[0 0 0 0 0 0]
Tweeter 1 is not depressed
Tweeter 2 is not depressed
Tweeter 3 is not depressed
Tweeter 4 is not depressed
Tweeter 5 is not depressed
Tweeter 6 is not depressed
```

In [25].

```
ти [OO].
# Saving our models:
import joblib
path = 'gdrive/MyDrive/Models/'
name= 'MultinomialNB'
model_path = path + name
_a = joblib.dump(model1, model_path)
In [87]:
name= 'GaussianNB'
model_path2 = path + name
_b = joblib.dump(model2, model_path2)
In [88]:
name= 'ComplementNB'
model_path3 = path + name
_c = joblib.dump(model3, model_path3)
In [89]:
name= 'BernoulliNB'
model_path4 = path + name
_d = joblib.dump(model4, model_path4)
In [ ]:
```