import pandas as pd
import numpy as np
import pickle
import warnings
warnings.filterwarnings("ignore")
from sklearn.metrics import ConfusionMatrixDisplay

In [2]: data=pd.read_csv(r"C:\Users\reshma_koduri\OneDrive\Documents\sentiment analysis for

In [3]: data.head(10)

Out[3]:		tweet_id	sentiment	sentiment_confidence	name	retweet_count	text	tweet_cre
	0	5.700000e+17	neutral	1.0000	cairdin	0	What @dhepburn said.	24-02-: 1
	1	5.700000e+17	positive	0.3486	jnardino	0	plus you've added commercials to the experien	24-02-; 1
	2	5.700000e+17	neutral	0.6837	yvonnalynn	0	I didn't today Must mean I need to take an	24-02-; 1
	3	5.700000e+17	negative	1.0000	jnardino	0	it's really aggressive to blast obnoxious "en	24-02-; 1
	4	5.700000e+17	negative	1.0000	jnardino	0	and it's a really big bad thing about it	24-02-; 1
	5	5.700000e+17	negative	1.0000	jnardino	0	seriously would pay \$30 a flight for seats th	24-02-; 1
	6	5.700000e+17	positive	0.6745	cjmcginnis	0	yes, nearly every time I fly VX this "ear wor	24-02-; 1
	7	5.700000e+17	neutral	0.6340	pilot	0	Really missed a prime opportunity for Men Wit	24-02-; 1
	8	5.700000e+17	positive	0.6559	dhepburn	0	Well, I didn'tbut NOW I DO! :-D	24-02-; 1

	tweet_id	sentiment	sentiment_confidence	name	retweet_count	text	tweet_cre
9 5.70	00000e+17	positive	1.0000	YupitsTate	0	it was amazing, and arrived an hour early. Yo	24-02-; 1

In [4]: data.tail(10)

tex	retweet_count	name	sentiment_confidence	sentiment	tweet_id	
Thanks! He i	0	Laurelinesblog	1.0000	positive	5.700000e+17	14630
thx fo nothing o getting us ou of the coun	0	MDDavis7	1.0000	negative	5.700000e+17	14631
@TilleyMonst George, tha doesn't loo goo	0	Chad_SMFYM	0.6760	neutral	5.700000e+17	14632
my flight wa Cancelle Flightled leaving to	0	RussellsWriting	1.0000	negative	5.700000e+17	14633
right on cu with th delays <mark>と</mark>	0	GolfWithWoody	0.6684	negative	5.700000e+17	14634
thank you w got on different fligh to Chi	0	KristenReenders	0.3487	positive	5.700000e+17	14635
leaving ove 20 minute Late Flight. N warni	0	itsropes	1.0000	negative	5.700000e+17	14636
Please brin America Airlines t #BlackBerry1	0	sanyabun	1.0000	neutral	5.700000e+17	14637
you have m money, yo change m flight, and	0	SraJackson	1.0000	negative	5.700000e+17	14638
we have 8 pp so we need know ho many seat	0	daviddtwu	0.6771	neutral	5.700000e+17	14639

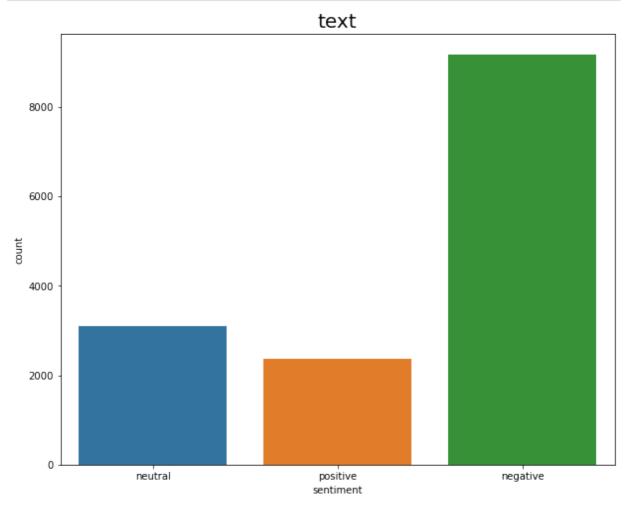
In [5]:

data.info()

```
<class 'pandas.core.frame.DataFrame'>
         RangeIndex: 14640 entries, 0 to 14639
         Data columns (total 10 columns):
                                     Non-Null Count Dtype
          #
              Column
         ---
                                      -----
          0
              tweet_id
                                      14640 non-null float64
          1
              sentiment
                                      14640 non-null object
              sentiment_confidence 14640 non-null float64
          2
          3
                                      14640 non-null object
          4
              retweet_count
                                      14640 non-null int64
          5
              text
                                      14640 non-null object
          6
              tweet created
                                      14640 non-null object
          7
              latitude
                                      14640 non-null float64
          8
              longitude
                                      14640 non-null float64
          9
              airline
                                      14640 non-null object
         dtypes: float64(4), int64(1), object(5)
         memory usage: 1.1+ MB
In [6]:
          data.describe()
Out[6]:
                    tweet_id sentiment_confidence retweet_count
                                                                   latitude
                                                                               longitude
         count 1.464000e+04
                                    14640.000000
                                                  14640.000000 14640.000000
                                                                            14640.000000
         mean 5.692602e+17
                                        0.900169
                                                      0.082650
                                                                  39.139901
                                                                              -98.687615
           std 8.100563e+14
                                        0.162830
                                                      0.745778
                                                                   3.599528
                                                                               18.731383
               5.680000e+17
                                        0.335000
                                                      0.000000
                                                                  24.295539
                                                                             -121.892659
           min
          25% 5.690000e+17
                                        0.692300
                                                      0.000000
                                                                  36.664544
                                                                             -119.106167
          50% 5.690000e+17
                                        1.000000
                                                      0.000000
                                                                  40.060817
                                                                             -101.278813
          75% 5.700000e+17
                                        1.000000
                                                      0.000000
                                                                  40.759700
                                                                              -74.002575
          max 5.700000e+17
                                        1.000000
                                                     44.000000
                                                                  59.658972
                                                                              -73.388100
In [7]:
          data.shape
         (14640, 10)
Out[7]:
In [8]:
          data.isna().sum()
         tweet id
                                  0
Out[8]:
         sentiment
                                  0
         sentiment_confidence
                                   0
         name
                                  0
                                  0
         retweet count
                                  0
         tweet_created
         latitude
         longitude
         airline
         dtype: int64
```

Visualization

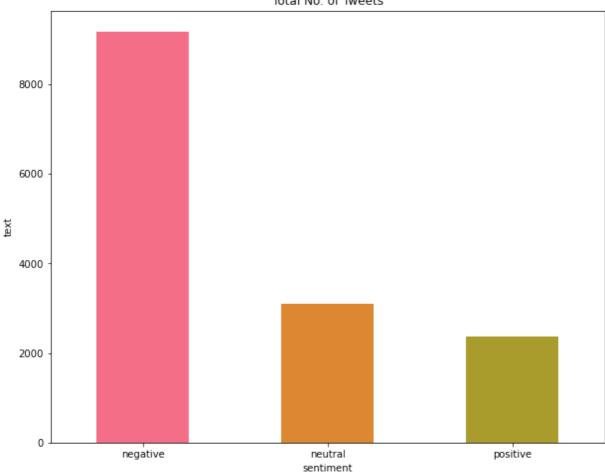
```
import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize = (10, 8))
ax = sns.countplot(x = 'sentiment', data = data)
ax.set_title(label = 'text', fontsize = 20)
plt.show()
```

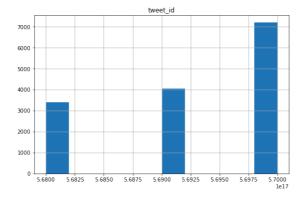


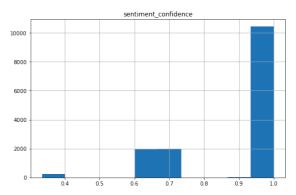
```
colors=sns.color_palette('husl',10)
pd.Series(data['sentiment']).value_counts().plot(kind="bar",color=colors,figsize=(10
plt.xlabel('sentiment',fontsize=10)
plt.ylabel('text',fontsize=10)
```

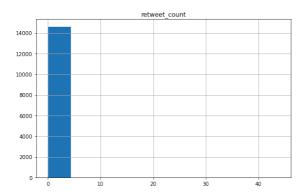
Out[10]: Text(0, 0.5, 'text')

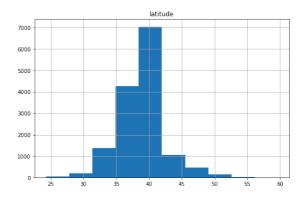
Total No. of Tweets

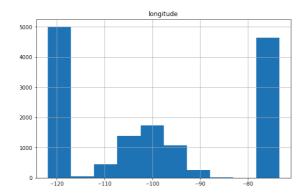






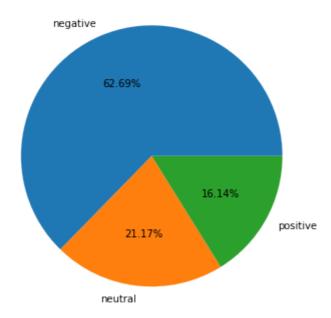






```
plt.figure(figsize=(8, 6))
    data['sentiment'].value_counts().plot.pie(autopct='%2.2f%%')
    plt.title('sentiment analysis')
    plt.ylabel('')
    plt.show()
```

sentiment analysis

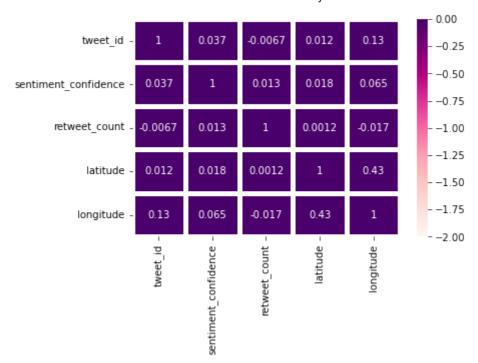


In [13]: cor=data.corr() cor

Out[13]:		tweet_id	$sentiment_confidence$	retweet_count	latitude	longitude
	tweet_id	1.000000	0.037452	-0.006650	0.011716	0.126882
	sentiment_confidence	0.037452	1.000000	0.012581	0.018142	0.064819
	retweet_count	-0.006650	0.012581	1.000000	0.001165	-0.016931
	latitude	0.011716	0.018142	0.001165	1.000000	0.425175
	longitude	0.126882	0.064819	-0.016931	0.425175	1.000000

import seaborn as sb
sb.heatmap(cor,vmax=0,vmin=-2,annot=True,linewidth=-5,cmap="RdPu")

Out[14]: <AxesSubplot:>



In [15]: #data['negativereason'] = data['negativereason'].fillna('') + ' ' + data['negativere
#data['negativereason_confidence'] = data['negativereason_confidence'].fillna('') +
#data.fillna(1,inplace=True)

Out[16]:	tweet_id	sentiment	sentiment_confidence	name	retweet_count	text
0	5.700000e+17	0	1.0000	cairdin	0	What @dhepburn said.
1	5.700000e+17	1	0.3486	jnardino	0	plus you've added commercials to the experien
2	5.700000e+17	0	0.6837	yvonnalynn	0	l didn't today Must mean l need to take an
3	5.700000e+17	-1	1.0000	jnardino	0	it's really aggressive to blast obnoxious "en
4	5.700000e+17	-1	1.0000	jnardino	0	and it's a really big bad thing about it
•••						
14635	5.700000e+17	1	0.3487	KristenReenders	0	thank you we got on a different flight to Chi

text	retweet_count	name	sentiment_confidence	sentiment	tweet_id	
leaving over 20 minutes Late Flight. No warni	0	itsropes	1.0000	-1	5.700000e+17	14636
Please bring American Airlines to #BlackBerry10	0	sanyabun	1.0000	0	5.700000e+17	14637
you have my money, you change my flight, and	0	SraJackson	1.0000	-1	5.700000e+17	14638
we have 8 ppl so we need 2 know how many seat	0	daviddtwu	0.6771	0	5.700000e+17	14639

 $14640 \text{ rows} \times 10 \text{ columns}$

```
In [17]:
          data['text']
                                                What @dhepburn said.
Out[17]:
                   plus you've added commercials to the experien...
                    I didn't today... Must mean I need to take an...
         3
                   it's really aggressive to blast obnoxious "en...
                            and it's a really big bad thing about it
         14635
                   thank you we got on a different flight to Chi...
         14636
                    leaving over 20 minutes Late Flight. No warni...
         14637
                    Please bring American Airlines to #BlackBerry10
         14638
                   you have my money, you change my flight, and ...
         14639
                   we have 8 ppl so we need 2 know how many seat...
         Name: text, Length: 14640, dtype: object
In [18]:
          x=data['text']
          y=data['sentiment']
In [19]:
                                                What @dhepburn said.
Out[19]:
                   plus you've added commercials to the experien...
                   I didn't today... Must mean I need to take an...
                    it's really aggressive to blast obnoxious "en...
         3
         4
                            and it's a really big bad thing about it
         14635
                   thank you we got on a different flight to Chi...
         14636
                    leaving over 20 minutes Late Flight. No warni...
         14637
                    Please bring American Airlines to #BlackBerry10
         14638
                   you have my money, you change my flight, and ...
         14639
                   we have 8 ppl so we need 2 know how many seat...
         Name: text, Length: 14640, dtype: object
In [20]:
          У
```

```
0
Out[20]:
                   1
         3
                  -1
                  -1
                  . .
         14635
                  1
         14636
                  -1
         14637
                  0
         14638
                  -1
         14639
                   0
         Name: sentiment, Length: 14640, dtype: int64
In [21]:
          from sklearn.feature_extraction.text import TfidfVectorizer
          tfid = TfidfVectorizer()
In [22]:
          x_final=tfid.fit_transform(x)
In [23]:
          pip install threadpoolctl==3.1.0
         Requirement already satisfied: threadpoolctl==3.1.0 in c:\users\reshma_koduri\anacond
```

a3\lib\site-packages (3.1.0)

Note: you may need to restart the kernel to use updated packages.

DEPRECATION: pyodbc 4.0.0-unsupported has a non-standard version number. pip 24.1 wil l enforce this behaviour change. A possible replacement is to upgrade to a newer vers ion of pyodbc or contact the author to suggest that they release a version with a con forming version number. Discussion can be found at https://github.com/pypa/pip/issue s/12063

```
In [24]:
          from imblearn.over_sampling import SMOTE
          smote=SMOTE()
          x_sm,y_sm=smote.fit_resample(x_final,y)
```

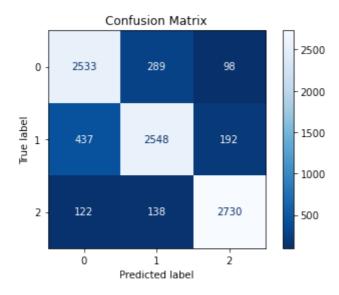
Data splitting

```
In [25]:
          from sklearn.model selection import train test split
In [26]:
          x_train,x_test,y_train,y_test=train_test_split(x_sm,y_sm,test_size=0.33,random_state
In [27]:
          print(x_train.shape)
          print(x test.shape)
          print(y_train.shape)
          print(y_test.shape)
          (18447, 15045)
          (9087, 15045)
          (18447,)
          (9087,)
```

Logistic Regression

```
In [28]:
          from sklearn.linear_model import LogisticRegression
          cls=LogisticRegression()
          cls=cls.fit(x_train,y_train)
In [29]:
          y_pred1=cls.predict(x_test)
          y_pred1
         array([ 0, 1, 0, ..., -1, 1, 1], dtype=int64)
Out[29]:
In [30]:
          from sklearn.metrics import accuracy_score,confusion_matrix
          confusion_matrix(y_pred1,y_test)
                               98],
         array([[2533, 289,
Out[30]:
                 [ 437, 2548, 192],
                 [ 122, 138, 2730]], dtype=int64)
In [31]:
          accuracy_score(y_pred1,y_test)
         0.8595796192362716
Out[31]:
In [32]:
          cm1=confusion_matrix(y_pred1,y_test)
          ConfusionMatrixDisplay(cm1).plot(cmap='Blues_r')
          plt.title('Confusion Matrix')
```

Out[32]: Text(0.5, 1.0, 'Confusion Matrix')



from sklearn.metrics import classification_report
report1=classification_report(y_pred1,y_test)
print("Classification Report:\n", report1)

Classification Report:

	precision	recall	f1-score	support
-1	0.82	0.87	0.84	2920
0	0.86	0.80	0.83	3177
1	0.90	0.91	0.91	2990
accuracy			0.86	9087
macro avg	0.86	0.86	0.86	9087

weighted avg

0.86

0.86

0.86

9087

RandomForest Classifier

```
In [34]:
          from sklearn.ensemble import RandomForestClassifier
In [35]:
          cls=RandomForestClassifier()
          cls.fit(x_train,y_train)
Out[35]:
          ▼ RandomForestClassifier
         RandomForestClassifier()
In [36]:
          y_pred2=cls.predict(x_test)
          y_pred2
          array([ 0, 1, 0, ..., -1, 1, 1], dtype=int64)
Out[36]:
In [37]:
          from sklearn.metrics import accuracy_score,confusion_matrix
          confusion_matrix(y_pred2,y_test)
          array([[2786, 318, 140],
Out[37]:
                 [ 239, 2583,
                               88],
                 [ 67,
                          74, 2792]], dtype=int64)
In [38]:
          accuracy_score(y_pred2,y_test)
          0.898096181357984
Out[38]:
In [39]:
          cm2=confusion_matrix(y_pred2,y_test)
          ConfusionMatrixDisplay(cm2).plot(cmap='Blues_r')
          plt.title('Confusion Matrix')
          Text(0.5, 1.0, 'Confusion Matrix')
Out[39]:
                      Confusion Matrix
                                                   2500
                  2786
                            318
                                       140
            0
                                                   2000
                                                   1500
                  239
                            2583
                                        88
                                                  1000
                  67
                             74
                                       2792
            2 ·
                                                   500
                   Ó
                                        2
```

Predicted label

```
In [40]:
          from sklearn.metrics import classification_report
          report2=classification_report(y_pred2,y_test)
          print("Classification Report:\n", report2)
         Classification Report:
                         precision
                                      recall f1-score
                                                          support
                             0.90
                                       0.86
                                                  0.88
                    -1
                                                            3244
                                                  0.88
                     0
                             0.87
                                        0.89
                                                            2910
                             0.92
                                        0.95
                                                  0.94
                                                            2933
             accuracy
                                                  0.90
                                                            9087
```

0.90

0.90

0.90

0.90

9087

9087

DecisionTree Classifier

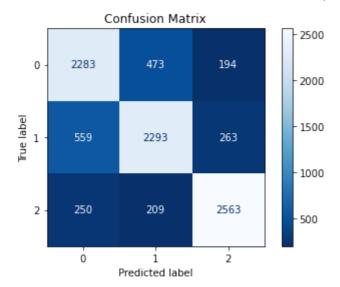
0.90

0.90

macro avg

weighted avg

```
In [41]:
          from sklearn.tree import DecisionTreeClassifier
In [42]:
          tree=DecisionTreeClassifier()
          tree.fit(x_train,y_train)
Out[42]:
         ▼ DecisionTreeClassifier
         DecisionTreeClassifier()
In [43]:
          y_pred3=tree.predict(x_test)
          y pred3
         array([1, 1, 0, ..., 1, 1, 1], dtype=int64)
Out[43]:
In [44]:
          confusion_matrix(y_pred3,y_test)
         array([[2283, 473, 194],
Out[44]:
                 [ 559, 2293, 263],
                       209, 2563]], dtype=int64)
In [45]:
          accuracy_score(y_pred3,y_test)
          0.7856278199625839
Out[45]:
In [46]:
          cm3=confusion_matrix(y_pred3,y_test)
          ConfusionMatrixDisplay(cm3).plot(cmap='Blues_r')
          plt.title('Confusion Matrix')
         Text(0.5, 1.0, 'Confusion Matrix')
Out[46]:
```



```
from sklearn.metrics import classification_report
report3=classification_report(y_pred3,y_test)
print("Classification Report:\n", report3)
```

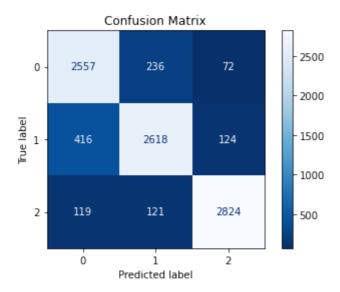
Classification	Report:			
	precision	recall	f1-score	support
-1	0.74	0.77	0.76	2950
0	0.77	0.74	0.75	3115
1	0.85	0.85	0.85	3022
accuracy			0.79	9087
macro avg	0.79	0.79	0.79	9087
weighted avg	0.79	0.79	0.79	9087

Support Vector Classifier

```
In [48]:
          from sklearn.svm import SVC
          svm_classifier = SVC(kernel='linear', random_state=42)
          svm_classifier.fit(x_train, y_train)
Out[48]:
                            SVC
         SVC(kernel='linear', random_state=42)
In [49]:
          y_pred4 = svm_classifier.predict(x_test)
          confusion_matrix(y_pred4,y_test)
         array([[2557, 236,
Out[49]:
                [ 416, 2618, 124],
                [ 119, 121, 2824]], dtype=int64)
In [50]:
          accuracy_score(y_test,y_pred4)
          0.8802685154616485
Out[50]:
In [51]:
          cm4=confusion_matrix(y_pred4,y_test)
          ConfusionMatrixDisplay(cm4).plot(cmap='Blues_r')
```

```
plt.title('Confusion Matrix')
```

Out[51]: Text(0.5, 1.0, 'Confusion Matrix')



```
from sklearn.metrics import classification_report
report4=classification_report(y_pred4,y_test)
print("Classification Report:\n", report4)
```

Classification Report:

```
precision
                          recall f1-score
                                             support
                  0.83
                           0.89
                                     0.86
         -1
                                               2865
          0
                  0.88
                            0.83
                                     0.85
                                               3158
          1
                  0.94
                            0.92
                                     0.93
                                               3064
                                     0.88
                                               9087
   accuracy
                0.88
                                     0.88
                                               9087
  macro avg
                           0.88
                                     0.88
weighted avg
                 0.88
                           0.88
                                               9087
```

```
In [53]:
    from prettytable import PrettyTable
    table=PrettyTable()
    table.field_names = ["TEST_SIZE","0.33"]
    table.add_row(["Logistic Regression",0.8595796192362716])
    table.add_row(["Random Forest Classifier",0.898096181357984])
    table.add_row(["DecisionTree Classifier",0.7856278199625839])
    table.add_row(["Support Vector Classifier",0.8802685154616485])
    print(table)
```

```
import tkinter as tk
from tkinter import ttk
from vaderSentiment.vaderSentiment import SentimentIntensityAnalyzer
import pandas as pd

def analyze_sentiment():
```

```
text = text_entry.get("1.0", "end-1c").strip() # Get text from the Text widget
    if text:
        analyzer = SentimentIntensityAnalyzer()
        sentiment_scores = analyzer.polarity_scores(text)
        sentiment = sentiment_scores['compound']
        if sentiment >= 0.05:
            result_label.config(text="Positive", foreground="green")
        elif sentiment <= -0.05:</pre>
            result_label.config(text="Negative", foreground="red")
        else:
            result_label.config(text="Neutral", foreground="blue")
    else:
        result_label.config(text="Please enter text!", foreground="orange")
data = pd.read_csv(r"C:\Users\reshma_koduri\OneDrive\Documents\sentiment analysis fo
window = tk.Tk()
window.title('Sentiment Analysis')
text_label = ttk.Label(window, text='Enter text:')
text_label.pack()
text_entry = tk.Text(window, height=5, width=50)
text entry.pack()
analyze_button = ttk.Button(window, text='Analyze Sentiment', command=analyze_sentiment')
analyze_button.pack()
result_label = ttk.Label(window, text='')
result_label.pack()
window.mainloop()
```

In []: