IMPORTS

```
import numpy as np
import pandas as pd
import re
import nltk
import random
from nltk.corpus import stopwords
import string
from textblob import Word
from sklearn import metrics
nltk.download('omw-1.4')
nltk.download('stopwords')
from sklearn.model_selection import train_test_split,GridSearchCV, cross_val_score, cross_val_predict, StratifiedKFold, cross_val
from sklearn.naive_bayes import MultinomialNB
from sklearn.linear_model import LogisticRegression
from \ sklearn. feature\_extraction. text \ import \ TfidfVectorizer, TfidfTransformer, CountVectorizer, TfidfTransformer, TfidfTransformer, CountVectorizer, TfidfTransformer, CountVectorizer, TfidfTransformer, T
from sklearn.pipeline import Pipeline
from nltk.stem import PorterStemmer
from nltk.stem import WordNetLemmatizer
nltk.download('wordnet')
from xgboost import XGBClassifier
from sklearn.svm import SVC
from sklearn.preprocessing import LabelEncoder
from sklearn.tree import DecisionTreeClassifier
from sklearn.preprocessing import StandardScaler
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import fl_score, accuracy_score, precision_score, recall_score, roc_auc_score, classification_report, confusi-
from sklearn.ensemble import RandomForestClassifier
from sklearn.datasets import make_classification,make_regression
from sklearn.cluster import KMeans
import matplotlib.pyplot as plt
%matplotlib inline
           [nltk_data] Downloading package omw-1.4 to /root/nltk_data...
[nltk_data] Downloading package stopwords to /root/nltk_data...
            [nltk data]
                                           Unzipping corpora/stopwords.zip.
           [nltk_data] Downloading package wordnet to /root/nltk_data...
```

→ DATA LOADING

```
tweets = pd.read_csv("tweet_emotions.csv")
tweets['new_content']=np.nan
tweets=tweets.rename(columns={'sentiment':'emotion'})
tweets.info()
tweets.head()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 40000 entries, 0 to 39999
     Data columns (total 4 columns):
                       Non-Null Count Dtype
          Column
      #
          tweet_id
                         40000 non-null
                                          int64
      0
          emotion
                        40000 non-null object
                        40000 non-null object
          content
     3   new_content 0 non-null float
dtypes: float64(1), int64(1), object(2)
                                          float64
     memory usage: 1.2+ MB
          tweet id emotion
                                                                  content new content
     0 1956967341
                                     @tiffanylue i know i was listenin to bad habi...
                                                                                    NaN
                        empty
      1 1956967666
                                                                                    NaN
                       sadness
                                  Layin n bed with a headache ughhhh...waitin o...
     2 1956967696
                       sadness
                                             Funeral ceremony...gloomy friday...
                                                                                    NaN
     3 1956967789 enthusiasm
                                           wants to hang out with friends SOON!
                                                                                    NaN
      4 1956968416
                        neutral @dannycastillo We want to trade with someone w...
                                                                                    NaN
#Size of the tweet dataset
tweets.shape
     (40000, 4)
#data types for tweets data set
tweets.dtypes
     tweet id
                        int64
     emotion
                      object
     content
                      object
     new_content
     dtype: object
#Create data frame for music
music = pd.read_csv("music.csv")
music.shape
     (13, 4)
```

genres

ongs

dtype: object

```
#data types for music data set
music.dtypes

sr_number int64
emotion object
```

→ DATA PREPROCESSING

object

object

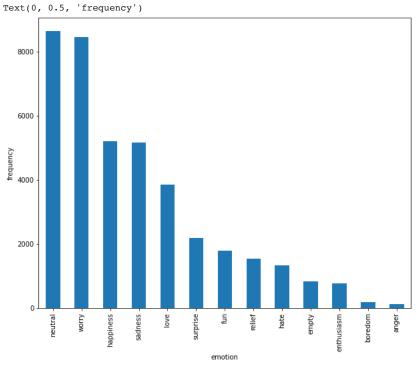
```
tweets.columns

Index(['tweet_id', 'emotion', 'content', 'new_content'], dtype='object')

#checking duplicates for each colum
[(i,tweets[i].duplicated().sum()) for i in tweets.columns]

[('tweet_id', 0), ('emotion', 39987), ('content', 173), ('new_content', 39999)]

#Finding the distribution of the response variable
plt.figure(figsize=(10, 8))
tweets['emotion'].value_counts().plot(kind='bar')
plt.xlabel('emotion')
plt.ylabel('frequency')
```



```
#Data Preprocessing
lemmatizer=WordNetLemmatizer()
stop_words = stopwords.words("english")
def run_test(y):
    for j,i in enumerate(y):
        match=re.search(r'(^@\w+)', i)
        if match:
        match1=re.sub(match.group(),'',i).lower()
        new_word=re.findall(r'(\w+)', match1)
        x=" ".join(lemmatizer.lemmatize(item) for item in new_word if item not in stop_words)
        tweets['new_content'][j]=x
    else:
        new_word=re.findall(r'(\w+)', i.lower())
        x=" ".join(lemmatizer.lemmatize(item) for item in new_word if item not in stop_words)
        tweets['new_content'][j]=x
        continue
```

tweets.shape

(40000, 4)

```
tweets['emotion'].value_counts()
     neutral
                    8638
     worry
                    8459
     happiness
                    5209
     sadness
                    5165
     love
                    3842
     surprise
                    2187
     fun
                    1776
     relief
                    1526
     hate
                    1323
     empty
     enthusiasm
                     759
     boredom
                     179
                     110
     anger
     Name: emotion, dtype: int64
```

→ DATA SPLITTING

```
X = tweets['new_content']
y = tweets["emotion"]

cv=CountVectorizer()
X=cv.fit_transform(X)
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=10,stratify=y)

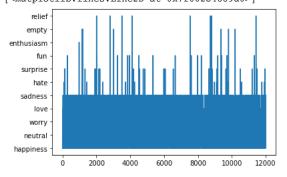
#Multinomial Naive Bayes
model=MultinomialNB()
model.fit(X_train,y_train)
NB_predicted=model.predict(X_test)
print(classification_report(y_test,NB_predicted))
plt.plot(NB_predicted)
```

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result)) precision recall f1-score support

	precision	recurr	11-50016	Bupport
anger	0.00	0.00	0.00	33
boredom	0.00	0.00	0.00	54
empty	0.33	0.01	0.02	248
enthusiasm	0.00	0.00	0.00	228
fun	0.00	0.00	0.00	533
happiness	0.32	0.30	0.31	1563
hate	0.43	0.02	0.03	397
love	0.44	0.29	0.35	1152
neutral	0.33	0.40	0.36	2591
relief	0.12	0.00	0.00	458
sadness	0.29	0.15	0.20	1549
surprise	0.04	0.00	0.00	656
worry	0.28	0.64	0.39	2538
accuracy			0.31	12000
macro avg	0.20	0.14	0.13	12000
weighted avg	0.28	0.31	0.26	12000

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))
[<matplotlib.lines.Line2D at 0x7f002c4889a0>]



```
#using TF-IDF Transformer
X = tweets['new_content']
y = tweets["emotion"]
X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=10,stratify=y)
tf = TfidfVectorizer(analyzer='word',max_features=1000,ngram_range=(1,3))
x_tf = tf.fit_transform(X_train)
x_val_tf = tf.transform(X_test)

#Logistic Regression
model =LogisticRegression(solver='liblinear',C=1)
model.fit(x_tf,y_train)
Log_predicted=model.predict(x_val_tf)
print(classification_report(y_test,Log_predicted))
plt.plot(Log_predicted)
```

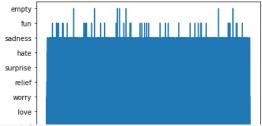
	precision	recall	il-score	support
anger	0.00	0.00	0.00	33
boredom	0.00	0.00	0.00	54
empty	0.14	0.00	0.01	248
enthusiasm	0.00	0.00	0.00	228
fun	0.20	0.02	0.03	533
happiness	0.33	0.35	0.34	1563
hate	0.44	0.15	0.23	397
love	0.47	0.37	0.42	1152
neutral	0.34	0.58	0.43	2591
relief	0.33	0.03	0.06	458
sadness	0.35	0.23	0.28	1549
surprise	0.30	0.03	0.06	656
worry	0.32	0.47	0.38	2538
accuracy			0.34	12000
macro avg	0.25	0.17	0.17	12000
weighted avg	0.33	0.34	0.31	12000

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

[<matplotlib.lines.Line2D at 0x7f002be7a190>]



cv=CountVectorizer()

X=cv.fit_transform(X)

X_train,X_test,y_train,y_test=train_test_split(X,y,test_size=0.3,random_state=10)

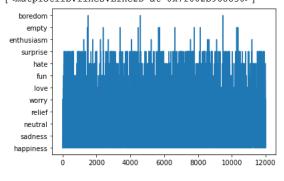
```
#SVC Model
ymodel = SVC()
model.fit(X_train,y_train)
pred_svm = model.predict(X_test)
print(classification_report(pred_svm, y_test))
plt.plot(pred_svm)
```

	precision	recall	f1-score	support
anger	0.00	0.00	0.00	0
boredom	0.00	0.00	0.00	3
empty	0.00	0.00	0.00	11
enthusiasm	0.00	0.00	0.00	12
fun	0.03	0.12	0.05	146
happiness	0.32	0.32	0.32	1542
hate	0.17	0.37	0.23	186
love	0.39	0.45	0.42	997
neutral	0.59	0.34	0.43	4476
relief	0.06	0.20	0.09	132
sadness	0.27	0.31	0.29	1289
surprise	0.05	0.15	0.07	204
worry	0.40	0.35	0.37	3002
accuracy			0.34	12000
macro avg	0.17	0.20	0.18	12000
weighted avg	0.43	0.34	0.37	12000

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall and F-score a warn prf(average, modifier, msg start, len(result))

_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall and F-score a
_warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall and F-score a

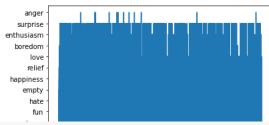
/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Recall and F-score a _warn_prf(average, modifier, msg_start, len(result))
[<matplotlib.lines.Line2D at 0x7f002b968850>]



```
#Decision Tree
model = DecisionTreeClassifier()
model.fit(X_train,y_train)
pred_decision = model.predict(X_test)
print(classification_report(pred_decision, y_test))
plt.plot(pred_decision)
```

	precision	recall	il-score	support
anger	0.04	0.08	0.05	12
boredom	0.02	0.04	0.03	23
empty	0.04	0.05	0.05	224
enthusiasm	0.03	0.04	0.03	165
fun	0.07	0.09	0.08	409
happiness	0.26	0.25	0.25	1628
hate	0.21	0.22	0.22	391
love	0.33	0.34	0.33	1130
neutral	0.44	0.32	0.37	3475
relief	0.07	0.11	0.08	288
sadness	0.23	0.25	0.24	1370
surprise	0.06	0.09	0.07	412
worry	0.30	0.31	0.30	2473
accuracy			0.27	12000
macro avg	0.16	0.17	0.16	12000
weighted avg	0.29	0.27	0.28	12000

[<matplotlib.lines.Line2D at 0x7f002bd35c10>]



```
#KNN Classifier
knn = KNeighborsClassifier()
k_range = list(range(1, 15))
param_grid = dict(n_neighbors=k_range)
# parameter tuning
grid = GridSearchCV(knn, param grid, cv=10, scoring='accuracy', return train score=False, verbose=1)
# model fitting
grid_search=grid.fit(X_train, y_train)
print(grid_search.best_params_)
```

Fitting 10 folds for each of 14 candidates, totalling 140 fits {'n_neighbors': 11}

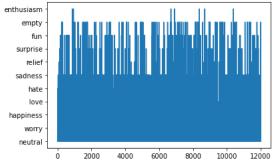
```
# For the best neighbors
knn = KNeighborsClassifier(n_neighbors=11)
knn.fit(X, y)
y_test_knn=knn.predict(X_test)
print(classification_report( y_test,y_test_knn))
plt.plot(y_test_knn)
```

 $/usr/local/lib/python 3.8/dist-packages/sklearn/metrics/_classification.py: 1318: \ Undefined Metric Warning: \ Precision \ and \ F-scorner \ F-scor$ _warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))
/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor

	_			
anger	0.00	0.00	0.00	27
boredom	0.00	0.00	0.00	47
empty	0.16	0.05	0.08	258
enthusiasm	0.14	0.00	0.01	215
fun	0.41	0.06	0.11	509
happiness	0.47	0.22	0.30	1576
hate	0.53	0.14	0.22	408
love	0.48	0.39	0.43	1158
neutral	0.27	0.90	0.42	2578
relief	0.64	0.05	0.09	444
sadness	0.44	0.19	0.26	1504
surprise	0.23	0.02	0.04	660
worry	0.55	0.16	0.25	2616
accuracy			0.33	12000
macro avg	0.33	0.17	0.17	12000
weighted avg	0.42	0.33	0.28	12000
-				

[<matplotlib.lines.Line2D at 0x7f002b3725b0>1



```
#XGB Classifier
model = XGBClassifier()
model.fit(X_train, y_train)
prod_xgbclassifier = model.predict(X_test)
```

```
print(classification_report(y_test,prod_xgbclassifier))
plt.plot(prod_xgbclassifier)
```

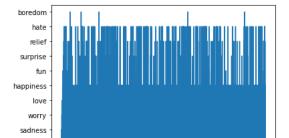
/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

precision recall f1-score support

	precision	recall	II-score	support
anger	0.00	0.00	0.00	27
boredom	0.00	0.00	0.00	47
empty	0.00	0.00	0.00	258
enthusiasm	0.00	0.00	0.00	215
fun	0.18	0.01	0.01	509
happiness	0.35	0.23	0.28	1576
hate	0.44	0.17	0.24	408
love	0.53	0.37	0.43	1158
neutral	0.29	0.77	0.42	2578
relief	0.27	0.04	0.07	444
sadness	0.40	0.18	0.25	1504
surprise	0.15	0.01	0.02	660
worry	0.34	0.30	0.32	2616
accuracy			0.33	12000
macro avg	0.23	0.16	0.16	12000
weighted avg	0.33	0.33	0.28	12000

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))

/usr/local/lib/python3.8/dist-packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning: Precision and F-scor _warn_prf(average, modifier, msg_start, len(result))
[<matplotlib.lines.Line2D at 0x7f002b88e250>]



4000

6000

8000

10000

12000

neutral

2000

```
#Random Forest Classifier
rfc = RandomForestClassifier(random_state = 1)
param_grid = {'n_estimators': [35, 40, 45, 50]}
# parameter tuning
grid = GridSearchCV(rfc, param_grid, cv=3, scoring='accuracy', return_train_score=False, verbose=1)
# model fitting
grid_search=grid.fit(X_train, y_train)
print(grid_search.best_params_)
# test data
rfc = RandomForestClassifier(n_estimators = 50)
rfc.fit(X, y)
y test rf=rfc.predict(X test)
cm = confusion_matrix(y_test,y_test_rf)
print(cm)
print(classification_report(y_test, y_test_rf))
plt.plot(y_test_rf)
```

```
0
                                     0
17
                                                     0]
                      0
                              0
                                  0
                                          0
                                             0
3
                                                 0
       0
          47
               0
                  0
                          0
           0
             238
                211
                              0
           0
              0
                          0
2
                                  0
1
                                      1
3
                                          0
                                             1
                                                 0
                                                     1]
0]
       0
                      1
                     502
                                                     4]
0]
       0
           0
                  0
                      1 1532
                              0
                                 15
                                     21
                             405
       0
           0
               0
                      0
                          0
len(y_test_rf)
   12000
```

```
#Final Results
df=pd.DataFrame(y_test_rf, columns = ['emotion'])
df['emotion'].nunique()

new_df = pd.merge(df, music, on='emotion', how='inner')
new_df = new_df.drop('sr_number', axis=1)
new_df
```

	emotion	genres	songs	0
0	neutral	['Country music']	['The Wolves','Heading South','Never Leave']	
1	neutral	['Country music']	['The Wolves','Heading South','Never Leave']	
2	neutral	['Country music']	['The Wolves','Heading South','Never Leave']	
3	neutral	['Country music']	['The Wolves','Heading South','Never Leave']	
4	neutral	['Country music']	['The Wolves','Heading South','Never Leave']	
11995	anger	['Soothing']	['River flows in you','Love yourself','I wont	
11996	anger	['Soothing']	['River flows in you','Love yourself','I wont	
11997	anger	['Soothing']	['River flows in you','Love yourself','I wont	
11998	anger	['Soothing']	['River flows in you','Love yourself','I wont	
11999	anger	['Soothing']	['River flows in you','Love yourself','I wont	
12000 rd	ows × 3 colu	ımns		

neutral 0 2000 4000 6000 8000 10000 12000

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