```
warnings.filterwarnings('ignore')
              dataset=pd.read_csv("https://github.com/XiaoyuLiu198/IMDB-Classify/blob/master/Tweets.csv")
             dataset.head()
 Out[1]:
                              tweet_id airline_sentiment airline_sentiment_confidence negativereason negativereason_confidence
                                                                                                                                             Virgin
              0 570306133677760513
                                                                                   1.0000
                                                                                                       NaN
                                                                                                                                     NaN
                                                    neutral
                                                                                                                                           America
                                                                                                                                             Virgin
              1 570301130888122368
                                                   positive
                                                                                   0.3486
                                                                                                       NaN
                                                                                                                                   0.0000
                                                                                                                                           America
                                                                                                                                             Virgin
              2 570301083672813571
                                                    neutral
                                                                                   0.6837
                                                                                                       NaN
                                                                                                                                     NaN
                                                                                                                                           America
                                                                                                                                             Virgin
                                                                                                  Bad Flight
              3 570301031407624196
                                                                                   1.0000
                                                                                                                                   0.7033
                                                  negative
                                                                                                                                           America
                                                                                                                                             Virgin
              4 570300817074462722
                                                                                   1.0000
                                                                                                                                   1.0000
                                                  negative
                                                                                                   Can't Tell
                                                                                                                                           America
 In [2]: #split and extract words
              data_senti=dataset.copy(deep=True)
              for i, comments in enumerate(dataset['text']):
                   data_senti['text'].iloc[i]=re.sub("[^a-zA-Z]"," ",comments)
                   data_senti['airline_sentiment'].iloc[i]=dataset['airline_sentiment'].iloc[i]
              data_senti.head()
 Out[2]:
                              tweet\_id \quad airline\_sentiment \quad airline\_sentiment\_confidence \quad negative reason \quad negative reason\_confidence \quad negative reason\_confi
                                                                                                                                            airline ai
                                                                                                                                             Virgin
              0 570306133677760513
                                                    neutral
                                                                                   1.0000
                                                                                                       NaN
                                                                                                                                     NaN
                                                                                                                                            America
                                                                                                                                             Virgin
              1 570301130888122368
                                                   positive
                                                                                   0.3486
                                                                                                       NaN
                                                                                                                                   0.0000
                                                                                                                                           America
                                                                                                                                             Virgin
              2 570301083672813571
                                                    neutral
                                                                                   0.6837
                                                                                                       NaN
                                                                                                                                     NaN
                                                                                                                                           America
                                                                                                                                             Virgin
              3 570301031407624196
                                                                                   1.0000
                                                                                                                                   0.7033
                                                   negative
                                                                                                  Bad Flight
                                                                                                                                           America
                                                                                                                                             Virgin
              4 570300817074462722
                                                                                                                                   1.0000
                                                  negative
                                                                                   1.0000
                                                                                                   Can't Tell
                                                                                                                                           America
 In [3]: data_senti=data_senti.replace("negative", -1)
              data_senti=data_senti.replace("positive",1)
              data_senti=data_senti.replace("neutral",0)
 In [4]: ##check if data is balanced
             import seaborn as sns
             sns.countplot(x='airline_sentiment', data=data_senti)
 Out[4]: <matplotlib.axes._subplots.AxesSubplot at 0x1c69242cfc8>
                 8000
                 6000
                 4000
                 2000
                                             airline_sentiment
 In [5]: data=data_senti.copy(deep=True)
 In [6]: | data=data[['text', 'airline_sentiment']]
              data.head()
 Out[6]:
                                                            text airline_sentiment
                                                                                  0
                               VirginAmerica What dhepburn said
              1 VirginAmerica plus you ve added commercials t...
                                                                                  1
                      VirginAmerica I didn t today Must mean I n...
                                                                                  0
                      VirginAmerica it s really aggressive to blast...
                                                                                 -1
                                                                                 -1
                      VirginAmerica and it s a really big bad thing...
 In [7]: | from sklearn.model_selection import train_test_split
              X_train, X_test, y_train, y_test = train_test_split(data['text'], data['airline_sentiment'],
              test_size=0.2, random_state=0)
             test_word_senti=testset.copy(deep=True) for i,comments in enumerate(testset['text']): test_word_senti['text'].iloc[i]=re.sub("
             [^a-zA-Z]"," ",comments).split() test_word_senti['sentiment'].iloc[i]=testset['sentiment'].iloc[i] test_word_senti.head()
 In [8]: import nltk
              #nltk.download('stopwords')
 In [9]: from nltk.corpus import stopwords
              en_stops = set(stopwords.words('english'))
             def delet_stw(sets,n_ds): for i,strings in enumerate(sets['text']): n_string=[] for words in strings: if words not in en_stops:
             n_string.append(words) n_ds[i]=n_string return n_ds x_word2_train=delet_stw(x_word2_train,x_word2_train)
             x_word2_test=delet_stw(x_word2_test,x_word2_test)
In [10]: ##td-idf transformation
              tf_x_train=X_train.copy()
              tf_x_test=X_test.copy()
              from sklearn.feature_extraction.text import TfidfVectorizer
              tfidfconverter = TfidfVectorizer(sublinear_tf=True, max_features=2000, min_df=5, max_df=0.7)
              tf_x_train =tfidfconverter.fit_transform(tf_x_train)#train the vectorizer, build the vocabul
              tf_x_test=tfidfconverter.transform(tf_x_test)
In [11]: | import matplotlib.pyplot as plt
             import matplotlib
              import seaborn as sns
              from sklearn.model_selection import GridSearchCV
              from sklearn.metrics import confusion_matrix
              from sklearn import preprocessing
              from sklearn.metrics import precision_recall_curve
              from sklearn.naive_bayes import GaussianNB
             lists=np.arange(0,1,0.05)
             scoreList = []
             accuracies = {}
              for i in lists:
                   clf = GaussianNB(var_smoothing=i)
                   clf.fit(tf_x_train.toarray(),y_train)
                   scoreList.append(clf.score(tf_x_test.toarray(),y_test))
              plt.plot(np.arange(0,1,0.05), scoreList)
              plt.xticks(np.arange(0,1,0.05))
              plt.xlabel("var_smoothing")
              plt.ylabel("Score")
              plt.show()
                 0.64
                 0.62
                 0.60
                 0.58
              ပ္တိ 0.56
                 0.54
                 0.52
                 0.50
                       0.00.09.10.19.20.29.30.39.40.49.50.59.60.69.70.79.80.89.90.95
                                             var_smoothing
In [12]: from sklearn.metrics import classification_report
              clf = GaussianNB(var_smoothing=0.01)
              clf.fit(tf_x_train.toarray(),y_train)
              predictions = clf.predict(tf_x_test.toarray())
              print(classification_report(y_test, predictions))
                                                     recall f1-score support
                                  precision
                                         0.88
                                                        0.64
                                                                      0.74
                                                                                     1870
                            -1
                             0
                                         0.44
                                                        0.54
                                                                      0.49
                                                                                      614
                                                                      0.53
                             1
                                         0.40
                                                        0.76
                                                                                      444
                                                                      0.63
                                                                                     2928
                   accuracy
                                         0.58
                                                        0.64
                                                                      0.58
                                                                                     2928
                  macro avg
             weighted avg
                                         0.72
                                                        0.63
                                                                      0.65
                                                                                     2928
In [13]: tf_x_train[1].toarray()
Out[13]: array([[0., 0., 0., ..., 0., 0., 0.]])
In [15]: from sklearn.naive_bayes import MultinomialNB
              model=MultinomialNB(alpha=1.0, fit_prior=True, class_prior=None)
              model.fit(tf_x_train,y_train)
             predictions = model.predict(tf_x_test)
             print(classification_report(y_test, predictions))
                                                     recall f1-score support
                                  precision
                                                                      0.85
                            -1
                                         0.74
                                                        0.98
                                                                                     1870
                             0
                                         0.76
                                                        0.31
                                                                      0.44
                                                                                      614
                                         0.87
                                                        0.38
                             1
                                                                      0.53
                                                                                      444
                                                                      0.75
                                                                                     2928
                   accuracy
                 macro avg
             weighted avg
                                                        0.75
                                                                      0.71
                                                                                     2928
                                         0.76
In [16]: #conda install -c glemaitre imbalanced-learn
              from imblearn.over_sampling import RandomOverSampler
              from imblearn.under_sampling import RandomUnderSampler
              #import RandomUnderSampler from imblearn
             undersample = RandomUnderSampler(sampling_strategy='majority')
             X_under, y_under = undersample.fit_resample(data, data['airline_sentiment'])
In [17]: | X_train, X_test, y_train, y_test = train_test_split(data['text'], data['airline_sentiment'],
              test_size=0.2, random_state=0)
              ##td-idf transformation
              tf_x_train=X_train.copy()
              tf_x_test=X_test.copy()
              tfidfconverter = TfidfVectorizer(sublinear_tf=True, max_features=2000, min_df=5, max_df=0.7)
              tf_x_train =tfidfconverter.fit_transform(tf_x_train)#train the vectorizer, build the vocabul
              tf_x_test=tfidfconverter.transform(tf_x_test)
             lists=np.arange(0,1,0.05)
              scoreList = []
             accuracies = {}
              for i in lists:
                   clf = GaussianNB(var_smoothing=i)
                   clf.fit(tf_x_train.toarray(),y_train)
                   scoreList.append(clf.score(tf_x_test.toarray(),y_test))
              plt.plot(np.arange(0,1,0.05), scoreList)
              plt.xticks(np.arange(0,1,0.05))
              plt.xlabel("var_smoothing")
             plt.ylabel("Score")
             plt.show()
             clf = GaussianNB(var_smoothing=0.01)
              clf.fit(tf_x_train.toarray(),y_train)
              predictions = clf.predict(tf_x_test.toarray())
              print(classification_report(y_test, predictions))
              model=MultinomialNB(alpha=1.0, fit_prior=True, class_prior=None)
              model.fit(tf_x_train,y_train)
              predictions = model.predict(tf_x_test)
             print(classification_report(y_test, predictions))
                 0.64
                 0.62
                 0.60
                 0.58
               ပ္တိ 0.56
                 0.54
                 0.52
                 0.50
                       0.00.09.10.19.20.29.30.39.40.49.50.59.60.69.70.79.80.89.90.95
                                             var_smoothing
                                  precision
                                                     recall f1-score
                                                                                support
                                         0.88
                                                        0.64
                                                                      0.74
                                                                                     1870
                            -1
                                         0.44
                                                        0.54
                                                                      0.49
                             0
                                                                                      614
                                         0.40
                                                        0.76
                                                                      0.53
                                                                                      444
                                                                                     2928
                                                                      0.63
                   accuracy
                                         0.58
                                                        0.64
                                                                      0.58
                  macro avg
                                                                                     2928
             weighted avg
                                         0.72
                                                        0.63
                                                                      0.65
                                                                                     2928
                                  precision
                                                     recall f1-score support
                            -1
                                         0.74
                                                        0.98
                                                                      0.85
                                                                                     1870
                             0
                                         0.76
                                                        0.31
                                                                      0.44
                                                                                      614
                             1
                                         0.87
                                                        0.38
                                                                      0.53
                                                                                      444
                                                                      0.75
                                                                                     2928
                   accuracy
                  macro avg
                                         0.79
                                                        0.56
                                                                                     2928
                                                                      0.61
             weighted avg
                                         0.76
                                                        0.75
                                                                      0.71
                                                                                     2928
 In [ ]:
```

In [1]: import pandas as pd

import re

import numpy as np
import warnings