```
import pandas as pd
In [1]:
         dataset = pd.read csv('hate speech.csv')
         dataset.head()
           id label
Out[1]:
                                                    tweet
           1
                  0 @user when a father is dysfunctional and is s...
            2
                  0 @user @user thanks for #lyft credit i can't us...
            3
         2
                  0
                                         bihday your majesty
         3
           4
                  0
                       #model i love u take with u all the time in ...
         4 5
                             factsquide: society now #motivation
                  0
         dataset.shape
In [2]:
         (5242, 3)
Out[2]:
         dataset.label.value counts()
In [3]:
         label
Out[3]:
              3000
              2242
         Name: count, dtype: int64
In [4]: for index, tweet in enumerate(dataset["tweet"][10:15]):
             print(index+1,"-",tweet)
         1 - âDD #ireland consumer price index (mom) climbed from previous 0.2% to 0.5% in ma
             #blog #silver #gold #forex
         2 - we are so selfish. #orlando #standwithorlando #pulseshooting #orlandoshooting #bi
         ggerproblems #selfish #heabreaking #values #love #
         3 - i get to see my daddy today!! #80days #gettingfed
         4 - ouch...junior is angryð@@@#got7 #junior #yugyoem
         5 - i am thankful for having a paner. #thankful #positive
In [5]: import re
         def clean_text(text):
             text = re.sub(r'[^a-zA-Z\']', ' ', text)
             text = re.sub(r'[^\x00-\x7F]+', ' ', text)
             text = text.lower()
             return text
         dataset['clean text'] = dataset.tweet.apply(lambda x: clean text(x))
In [6]:
         dataset.head(10)
In [7]:
```

```
id label
                                                                  tweet
 Out[7]:
                                                                                                    clean text
                                                                         user when a father is dysfunctional and is
           0
               1
                      0
                              @user when a father is dysfunctional and is s...
               2
                      0
                              @user @user thanks for #lyft credit i can't us...
                                                                          user user thanks for lyft credit i can't us...
               3
                      0
           2
                                                     bihday your majesty
                                                                                            bihday your majesty
                                                                          model i love u take with u all the time in
           3
                      0
                                #model i love u take with u all the time in ...
           4
               5
                      0
                                       factsquide: society now #motivation
                                                                                factsquide society now motivation
           5
               6
                      0
                             [2/2] huge fan fare and big talking before the...
                                                                         huge fan fare and big talking before the...
                             @user camping tomorrow @user @user @user
                                                                           user camping tomorrow user user user
               7
                      0
           6
                                                                 @use...
                                                                         the next school year is the year for exams
               8
                      0
                             the next school year is the year for exams.ŏ□□...
           7
           8
               9
                      0
                            we won!!! love the land!!! #allin #cavs #champ...
                                                                          we won love the land allin cavs champ...
             10
                      0
                               @user @user welcome here! i'm it's so #gr...
                                                                            user user welcome here i'm it's so gr...
           from nltk.corpus import stopwords
 In [8]:
           # listing stop words
           len(stopwords.words('english'))
           179
 Out[8]:
           stop words = [
 In [9]:
                 'a', 'about', 'above', 'after', 'again', 'against', 'all', 'am', 'an', 'and', 'any
                'aren\'t', 'as', 'at', 'be', 'because', 'been', 'before', 'being', 'below', 'betwee'
'both', 'but', 'by', 'can\'t', 'cannot', 'could', 'couldn\'t', 'did', 'didn\'t', '
                'does', 'doesn\'t', 'doing', 'don\'t', 'down', 'during', 'each', 'few', 'for', 'fr
                'further', 'had', 'hadn\'t', 'has', 'hasn\'t', 'have', 'haven\'t'
                                                                                               , 'having', 'he',
                'he\'ll', 'he\'s', 'her', 'here', 'here\'s', 'hers', 'herself', 'him', 'himself',
                'how', 'how\'s', 'i', 'i\'d', 'i\'ll', 'i\'m', 'i\'ve', 'if', 'in', 'into', 'is',
                'it', 'it\'s', 'its', 'itself', 'let\'s', 'me', 'more', 'most', 'mustn\'t', 'my',
                'no', 'nor', 'not', 'of', 'off', 'on', 'once', 'only', 'or', 'other', 'ought',
                'ours', 'ourselves', 'out', 'over', 'own', 'same', 'shan\'t', 'she', 'she\'d', 'sh
                'she\'s', 'should', 'shouldn\'t', 'so', 'some', 'such', 'than', 'that\'s',
                'their', 'theirs', 'them', 'themselves', 'then', 'there', 'there\'s', 'these', 'the
                'they\'d', 'they\'ll', 'they\'re', 'they\'ve', 'this', 'those', 'through', 'to', '
                'under', 'until', 'up', 'very', 'was', 'wasn\'t', 'we', 'we\'d', 'we\'ll', 'we\'re
                'we\'ve', 'were', 'weren\'t', 'what', 'what\'s', 'when', 'when\'s', 'where', 'where', 'which', 'while', 'who', 'who\'s', 'whom', 'why\'s', 'with', 'won\'t', 'wou
                'wouldn\'t', 'you', 'you\'d', 'you\'ll', 'you\'re', 'you\'ve', 'your', 'yours',
                 'yourself', 'yourselves'
           ]
In [10]: #Generate word frequency
           def gen freq(text):
                #Will store the list of words
```

```
word_list = []
#Loop over all the tweets and extract words into word_list
for tw_words in text.split():
    word_list.extend(tw_words)

#Create word frequencies using word_list
word_freq = pd.Series(word_list).value_counts()

#Drop the stopwords during the frequency calculation
word_freq = word_freq.drop(stop_words, errors='ignore')
return word_freq
```

```
In [15]: #Check whether a negation term is present in the text

def any_neg(stop_words):
    for word in stop_words:
        if word in ['n', 'no', 'non', 'not'] or re.search(r"\wn't", word):
            return 1
        else:
            return 0
```

```
In [16]: #Check whether one of the 100 rare words is present in the text

def any_rare(stop_words, rare_100):
    for word in stop_words:
        if word in rare_100:
            return 1
        else:
            return 0
```

```
In [17]: #Check whether prompt words are present

def is_question(stop_words):
    for word in stop_words:
        if word in ['when', 'what', 'how', 'why', 'who', 'where']:
            return 1
        else:
```

return 0

```
word freq = gen freq(dataset.clean text.str)
In [23]:
         #100 most rare words in the dataset
         rare 100 = word freq[-100:] # last 100 rows/words
         #Number of words in a tweet
         dataset['word_count'] = dataset.clean_text.str.split().apply(lambda x: len(x))
         #Negation present or not
         dataset['any_neg'] = dataset.clean_text.str.split().apply(lambda x: any_neg(x))
         #Prompt present or not
         dataset['is question'] = dataset.clean text.str.split().apply(lambda x: is question(x)
         #Any of the most 100 rare words present or not
         dataset['any_rare'] = dataset.clean_text.str.split().apply(lambda x: any_rare(x, rare_
         #Character count of the tweet
         dataset['char_count'] = dataset.clean_text.apply(lambda x: len(x))
In [24]:
         dataset.head(10)
```

4]:	id	label	tweet	clean_text	word_count	any_neg	is_question	any_rare	char_cour
0	1	0	@user when a father is dysfunctional and is s	user when a father is dysfunctional and is s	18	0	0	0	10
1	2	0	@user @user thanks for #lyft credit i can't us	user user thanks for lyft credit i can't us	19	0	0	0	1
2	3	0	bihday your majesty	bihday your majesty	3	0	0	0	
3	4	0	#model i love u take with u all the time in 	model i love u take with u all the time in 	12	0	0	0	
4	5	0	factsguide: society now #motivation	factsguide society now motivation	4	0	0	0	
5	6	0	[2/2] huge fan fare and big talking before the	huge fan fare and big talking before the	18	0	0	0	1
6	7	0	@user camping tomorrow @user @user @user @use	user camping tomorrow user user user use	11	0	0	0	
7	8	0	the next school year is the year for exams.ð□□	the next school year is the year for exams	20	0	0	0	1
8	9	0	we won!!! love the land!!! #allin #cavs #champ	we won love the land allin cavs champ	10	0	0	0	
9	10	0	@user @user welcome here ! i'm it's so #gr	user user welcome here i'm it's so gr	8	0	0	0	

```
In [25]: from sklearn.model_selection import train_test_split
X = dataset[['word_count', 'any_neg', 'any_rare', 'char_count', 'is_question']]
y = dataset.label
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=
```

```
In [26]: from sklearn.naive_bayes import GaussianNB
    #Initialize GaussianNB classifier
    model = GaussianNB()
    #Fit the model on the train dataset
```

```
model = model.fit(X train, y train)
         #Make predictions on the test dataset
         pred = model.predict(X_test)
         model.predict(X_test[5:10])
In [28]:
         array([0, 1, 1, 1, 1], dtype=int64)
Out[28]:
In [29]:
         from sklearn.metrics import accuracy score
         print("Accuracy:", accuracy_score(y_test, pred)*100, "%")
         Accuracy: 42.99332697807436 %
In [30]: from sklearn.ensemble import RandomForestClassifier
         #Create a random forest classifier, 100 trees
         clf rf=RandomForestClassifier()
         #Train the model using the training sets
         clf rf.fit(X train,y train)
         rf_pred=clf_rf.predict(X_test).astype(int)
In [32]: from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
         print(confusion_matrix(y_test,rf_pred))
         print(classification_report(y_test,rf_pred))
         print("Accuracy:",accuracy_score(y_test, rf_pred))
         [[399 200]
          [225 225]]
                       precision
                                    recall f1-score
                                                        support
                    0
                            0.64
                                      0.67
                                                 0.65
                                                            599
                    1
                            0.53
                                      0.50
                                                 0.51
                                                            450
                                                 0.59
                                                           1049
             accuracy
            macro avg
                            0.58
                                      0.58
                                                 0.58
                                                           1049
                                      0.59
                                                           1049
         weighted avg
                            0.59
                                                0.59
         Accuracy: 0.5948522402287894
In [33]: from sklearn.linear_model import LogisticRegression
         logreg = LogisticRegression(class_weight='balanced')
         logreg.fit(X train, y train)
Out[33]:
                       LogisticRegression
         LogisticRegression(class_weight='balanced')
         y_pred = logreg.predict(X_test) #predicting the values
In [34]:
```

from sklearn.metrics import classification_report In [35]: print(classification_report(y_test, y_pred)) precision recall f1-score support 0 0.58 0.63 0.61 599 1 0.49 0.54 0.52 450 1049 accuracy 0.57 macro avg 0.56 0.56 0.56 1049 weighted avg 0.57 0.57 0.57 1049 In []: