```
In [70]:
import numpy as np
import pandas as pd
from sklearn.model selection import train test split
import nltk
from nltk.corpus import stopwords
from nltk.classify import SklearnClassifier
from wordcloud import WordCloud, STOPWORDS
import matplotlib.pyplot as plt
%matplotlib inline
from subprocess import check output
import nltk
import re
from sklearn.feature_extraction.text import TfidfVectorizer
import string
In [71]:
# RandomForestClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.model selection import KFold, cross val score
from sklearn.metrics import precision_recall_fscore_support as score
from sklearn.model selection import train test split
### Grid-search
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import precision recall fscore support as score
from sklearn.model selection import train test split
from sklearn.metrics import classification report, confusion matrix, accuracy score
In [72]:
#columns = ['sentiment', 'text']
data = pd.read csv('TwitterHate.csv')
In [73]:
data.shape
Out[73]:
(31962, 3)
In [74]:
data.head()
Out[74]:
  id label
                                        tweet
        0 @user when a father is dysfunctional and is s...
0 1
          @user @user thanks for #lyft credit i can't us...
  2
2
  3
        0
                             bihday your majesty
             #model i love u take with u all the time in ...
3
  4
```

4 5

0

factsguide: society now #motivation

```
In [75]:
data.label.value counts()
Out[75]:
0
     29720
      2242
1
Name: label, dtype: int64
In [76]:
data = data[['tweet', 'label']]
In [77]:
data.shape
Out[77]:
(31962, 2)
In [78]:
data.head()
Out[78]:
                                 tweet label
0 @user when a father is dysfunctional and is s...
                                         0
   @user @user thanks for #lyft credit i can't us...
                                         0
2
                      bihday your majesty
                                         0
3
     #model i love u take with u all the time in ...
                                         0
          factsguide: society now #motivation
                                         0
In [79]:
data.isnull().sum()
Out[79]:
tweet
label
dtype: int64
In [80]:
data['tweet'].head(10)
Out[80]:
0
      @user when a father is dysfunctional and is s...
1
     @user @user thanks for #lyft credit i can't us...
2
                                      bihday your majesty
               i love u take with u all the time in ...
3
     #model
4
                 factsguide: society now
                                              #motivation
5
     [2/2] huge fan fare and big talking before the...
6
      Quser camping tomorrow Quser Quser Quser Quse...
7
     the next school year is the year for exams.õ...
8
     we won!!! love the land!!! #allin #cavs #champ...
      Quser Quser welcome here ! i'm it's so #gr...
9
Name: tweet, dtype: object
In [81]:
data['tweet'] = data['tweet'].str.replace('#', '')
data['tweet'] = data['tweet'].str.replace('amp', ' ')
data['tweet'] = data['tweet'].str.replace('rt', ' ')
```

```
data['tweet'] = data['tweet'].str.replace('http\S+', ' ')
In [82]:
stopwords = nltk.corpus.stopwords.words('english')
In [83]:
ps = nltk.PorterStemmer()
In [84]:
def clean text(tweet):
    tweet = "".join([word.lower() for word in tweet if word not in string.punctuation])
    tokens = re.split('\W+', tweet)
    tweet = [ps.stem(word) for word in tokens if word not in stopwords]
    return tweet
In [85]:
tfidf vect = TfidfVectorizer(analyzer=clean text, max features =5000)
X tfidf = tfidf vect.fit transform(data['tweet'])
In [86]:
X tfidf
Out[86]:
<31962x5000 sparse matrix of type '<class 'numpy.float64'>'
with 248445 stored elements in Compressed Sparse Row format>
In [87]:
X features = pd.DataFrame(X tfidf.toarray())
In [88]:
X features.head()
Out[88]:
              2
                 3
                    4 5
                               7
                                  8 9 ... 4990 4991 4992 4993
                                                                4994 4995 4996 4997 4998 4999
0 0.073464 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 ...
                                            0.0
                                                0.0
                                                     0.0
                                                         0.0 0.000000
                                                                      0.0
                                                                          0.0
                                                                               0.0
                                                                                   0.0
                                                                                        0.0
0.0
                                                0.0
                                                     0.0
                                                         0.0 0.000000
                                                                     0.0
                                                                          0.0
                                                                              0.0
                                                                                   0.0
                                                                                        0.0
2 0.112337 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 ...
                                                     0.0
                                                         0.0 0.000000
                                                                                        0.0
                                                0.0
                                                                      0.0
                                                                          0.0
                                                                              0.0
                                                                                   0.0
3 0.095789 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 ...
                                                         0.0 0.797498
                                            0.0
                                                0.0
                                                     0.0
                                                                     0.0
                                                                          0.0
                                                                              0.0
                                                                                   0.0
                                                                                        0.0
0.0
                                                0.0
                                                     0.0
                                                         0.0 0.000000
                                                                     0.0
                                                                          0.0
                                                                              0.0
                                                                                   0.0
                                                                                        0.0
5 rows × 5000 columns
In [89]:
X features.shape
Out[89]:
(31962, 5000)
In [90]:
feauture name = tfidf vect.get feature names()
pd.DataFrame(X tfidf.toarray(), columns = feauture name)
Out[90]:
```

ï ï¼ ï½

ð ð¾ð

ñ ó¾

0 01 02 03 04 05 1 10 100 ... í

```
0 0.073464 0.0 0.10
                 0.8
                    0.8
                       0.4
                          0.5 0.6 d.0
                                  10.0 ... 0.0 0.000000 0%
                                                     Ö!⁄9
                                                       0.000000
                                                              ðijğ
                                                                 0.0 6% 0.0
                                                                           Ø.0
   0.0 ... 0.0 0.000000 0.0 0.0 0.000000
                                                               0.0 0.0 0.0 0.0 0.0
  2 0.112337 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
                                   0.0 ... 0.0 0.000000 0.0 0.0 0.000000
                                                               0.0 0.0 0.0 0.0 0.0
   3 0.095789 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
                                   0.0 ... 0.0 0.000000 0.0 0.0 0.797498
                                                               0.0 0.0 0.0 0.0 0.0
   4 0.086500 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
                                   0.0 ... 0.0 0.000000 0.0 0.0 0.000000
                                                               0.0 0.0 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0
    0.056846 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
                                   0.0 ... 0.0 0.000000 0.0 0.0 0.000000
                                                               0.0 0.0 0.0 0.0 0.0
31959
    0.0 0.0 0.0 0.0 0.0
31960 0.055602 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0
                                   0.0 ... 0.0 0.000000 0.0 0.0 0.000000
                                                               0.0 0.0 0.0 0.0 0.0
0.0 0.0 0.0 0.0 0.0
```

#### 31962 rows × 5000 columns

```
In [91]:
```

```
X_train, X_test, y_train, y_test = train_test_split(X_features, data['label'], train_siz
e = 0.8, random_state = 123)
```

## In [92]:

```
print(len(X_train))
print(len(X_test))
print(len(y_train))
print(len(y_test))
```

25569

6393

25569

6393

#### In [93]:

```
X train.head()
```

#### Out[93]:

	0	1	2	3	4	5	6	7	8	9	 4990	4991	4992	4993	4994	4995	4996	4997	4998	4999
4039	0.068867	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.000000	0.0	0.0	0.0	0.0	0.0
28391	0.057648	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.000000	0.0	0.0	0.0	0.0	0.0
7600	0.058711	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.000000	0.0	0.0	0.0	0.0	0.0
2687	0.018865	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.314118	0.0	0.0	0.0	0.0	0.0
24346	0.076532	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	 0.0	0.0	0.0	0.0	0.000000	0.0	0.0	0.0	0.0	0.0

### 5 rows × 5000 columns

#### In [94]:

```
rf = RandomForestClassifier(n_estimators=200, n_jobs = -1)
rf_mdl = rf.fit(X_train, y_train)
```

### In [95]:

```
y_pred = rf_mdl.predict(X_test)
```

#### In [96]:

```
y_pred_prob = rf_mdl.predict_proba(X_test)
```

# In [97]:

```
y pred prob
Out[97]:
array([[1. , 0.
      [0.995, 0.005],
      [0.995, 0.005],
      [1. , 0. ],
[0.99 , 0.01 ],
      [0.7 , 0.3 ]])
In [98]:
print(confusion_matrix(y_test, y_pred))
[[5937
       23]
[ 199 234]]
In [99]:
print(classification_report(y_test, y_pred))
             precision
                      recall f1-score
                                            support
                 0.97
                          1.00
                                   0.98
                                              5960
                 0.91
                          0.54
                                    0.68
                                              433
                                    0.97
                                             6393
   accuracy
                0.94 0.77
                                    0.83
                                             6393
  macro avg
                0.96
                          0.97
                                   0.96
                                             6393
weighted avg
In [100]:
y_pred_train = rf_mdl.predict(X_train)
In [101]:
print(confusion matrix(y train, y pred train))
print(classification_report(y_train, y_pred_train))
[[23760 0]
 [ 6 1803]]
             precision recall f1-score
                                           support
                  1.00
                           1.00
                                     1.00
                                             23760
                  1.00
                           1.00
                                     1.00
                                              1809
                                     1.00
                                             25569
   accuracy
                                   1.00
                1.00 1.00
                                             25569
  macro avg
                                             25569
                          1.00
                                    1.00
                1.00
weighted avg
```

# **Naive bayes**

```
In [102]:
```

```
from sklearn.naive_bayes import GaussianNB
mdl = GaussianNB()
mdl_nb = mdl.fit(X_train, y_train)
y_pred = mdl_nb.predict(X_test)
```

# **Test data metrics**

```
In [103]:
```

```
print(confusion_matrix(y_test, y_pred))
```

```
[[4564 1396]
 [ 102 331]]
              precision
                           recall f1-score
                                               support
           0
                   0.98
                             0.77
                                       0.86
                                                  5960
                             0.76
                   0.19
                                       0.31
                                                   433
                                       0.77
                                                  6393
   accuracy
                                                  6393
                   0.58
                             0.77
                                       0.58
   macro avg
                             0.77
                                       0.82
                                                  6393
                   0.92
weighted avg
Train data metrics
In [104]:
y pred train = mdl nb.predict(X train)
print(confusion matrix(y train, y pred train))
print(classification report(y train, y pred train))
[[18082 5678]
   0 1809]]
                          recall f1-score
              precision
                                               support
                            0.76
           0
                   1.00
                                       0.86
                                                 23760
           1
                   0.24
                             1.00
                                       0.39
                                                 1809
                                       0.78
                                                 25569
   accuracy
                             0.88
                                       0.63
   macro avg
                   0.62
                                                25569
weighted avg
                   0.95
                             0.78
                                       0.83
                                                25569
In [105]:
from sklearn.linear model import LogisticRegression
lr = LogisticRegression()
lr mdl = lr.fit(X train, y train)
In [106]:
y pred = lr mdl.predict(X test)
In [107]:
print(confusion matrix(y test, y pred))
print(classification report(y test, y pred))
[[5948
        12]
 [ 273 160]]
              precision
                           recall f1-score
                                              support
           0
                   0.96
                             1.00
                                       0.98
                                                  5960
                   0.93
                             0.37
                                       0.53
                                                  433
           1
                                       0.96
                                                  6393
    accuracy
                   0.94
                             0.68
                                       0.75
   macro avg
                                                  6393
weighted avg
                   0.95
                             0.96
                                       0.95
                                                  6393
In [108]:
y pred train = lr mdl.predict(X train)
print(confusion_matrix(y_train, y_pred_train))
print(classification_report(y_train, y_pred_train))
```

print(classification\_report(y\_test, y\_pred))

[[23718

[ 1063

42]

746]]

precision

recall f1-score

support

```
0.96
                         1.00
                                  0.98
                                            23760
          0
                 0.95
                         0.41
                                   0.57
                                            1809
                                            25569
                                   0.96
   accuracy
                         0.71
  macro avg
                0.95
                                   0.78
                                            25569
weighted avg
                0.96
                          0.96
                                   0.95
                                            25569
```

# In [109]:

```
from sklearn.model_selection import GridSearchCV, StratifiedKFold
clf = LogisticRegression()
param = {}
clf = GridSearchCV(clf, param, cv = 2, n_jobs = -1, verbose = 1, scoring = "recall")
#grv = GridSearchCV(estimator= 1r, cv = StratifiedKFold(5), n_jobs = -1, verbose = 1, sco
ring = "recall")
grv_mdl = clf.fit(X_train, y_train)
```

Fitting 2 folds for each of 1 candidates, totalling 2 fits

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 2 concurrent workers.

[Parallel(n_jobs=-1)]: Done 2 out of 2 | elapsed: 9.3s remaining: 0.0s

[Parallel(n_jobs=-1)]: Done 2 out of 2 | elapsed: 9.3s finished
```

# In [110]:

```
y_pred = grv_mdl.predict(X_test)
print(confusion_matrix(y_test, y_pred))
print(classification_report(y_test, y_pred))
```

[ [5948 [ 273	12] 160]]				
		precision	recall	f1-score	support
	0	0.96	1.00	0.98	5960
	1	0.93	0.37	0.53	433
accu	racy			0.96	6393
macro	avg	0.94	0.68	0.75	6393
weighted	avg	0.95	0.96	0.95	6393

# In [111]:

```
y_pred_train = grv_mdl.predict(X_train)
print(confusion_matrix(y_train, y_pred_train))
print(classification_report(y_train, y_pred_train))
```

[[23718	42	]			
[ 1063	746	]]			
		precision	recall	f1-score	support
	0	0.96	1.00	0.98	23760
	1	0.95	0.41	0.57	1809
accui	racy			0.96	25569
macro	avg	0.95	0.71	0.78	25569
weighted	avg	0.96	0.96	0.95	25569

## In [ ]: