```
In [5]:
                                                                                        H
## import statements ##
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
% matplotlib inline
UsageError: Line magic function `%` not found.
Type Markdown and LaTeX: \alpha^2
                                                                                        H
In [6]:
train data = []
In [22]:
data_files =['C:\\Users\\singh\\OneDrive\\Desktop\\Youtube-Comments-Spam-Detection-maste
for file in data_files:
    data = pd.read_csv(file)
    train data.append(data)
train_data = pd.concat(train_data)
In [23]:
train_data.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1956 entries, 0 to 369
Data columns (total 5 columns):
 #
     Column
                 Non-Null Count Dtype
     ----
                 -----
                                 ----
 0
     COMMENT_ID 1956 non-null
                                 object
 1
     AUTHOR
                 1956 non-null
                                 object
 2
     DATE
                 1711 non-null
                                 object
 3
     CONTENT
                 1956 non-null
                                 object
 4
     CLASS
                 1956 non-null
                                 int64
dtypes: int64(1), object(4)
memory usage: 91.7+ KB
```

In [24]:
train_data.head()

Out[24]:

	COMMENT_ID	AUTHOR	DATE	CONTENT	CLASS
_Z	ZQPQhLyRh80UYxNuaDWhIGQYNQ96IuCg- AYWqNPjpU	Julius NM	2013-11- 07T06:20:48	Huh, anyway check out this you[tube] channel:	1
	LZQPQhLyRh_C2cTtd9MvFRJedxydaVW- 2sNg5Diuo4A	adam riyati	2013-11- 07T12:37:15	Hey guys check out my new channel and our firs	1
LZ	ZQPQhLyRh9MSZYnf8djyk0gEF9BHDPYrrK- qCczIY8	Evgeny Murashkin	2013-11- 08T17:34:21	just for test I have to say murdev.com	1
	z13jhp0bxqncu512g22wvzkasxmvvzjaz04	ElNino Melendez	2013-11- 09T08:28:43	me shaking my sexy ass on my channel enjoy ^^	1
	z13fwbwp1oujthgqj04chlngpvzmtt3r3dw	GsMega	2013-11- 10T16:05:38	watch? v=vtaRGgvGtWQ Check this out .	1
					

```
In [25]: ▶
```

```
train_data['CLASS'].value_counts()
```

Out[25]:

1 1005
 951

Name: CLASS, dtype: int64

```
In [ ]: ▶
```

#Data Cleaning

#We don't need all the features. Hence drop all the columns except CONTENT, CLASS

```
In [26]: ▶
```

```
def drop_fectures(features,data):
    data.drop(features,axis=1,inplace=True)
```

```
In [27]:
```

```
drop_fectures(['COMMENT_ID','AUTHOR','DATE'],train_data)
```

```
H
In [28]:
train_data.info()
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1956 entries, 0 to 369
Data columns (total 2 columns):
     Column
               Non-Null Count Dtype
                _____
 0
     CONTENT 1956 non-null
                                  object
               1956 non-null
 1
     CLASS
                                  int64
dtypes: int64(1), object(1)
memory usage: 45.8+ KB
In [29]:
                                                                                                  H
import re
re.findall("[A-Za-z]+", "Hey Ravi57, How are you doing?")
Out[29]:
['Hey', 'Ravi', 'How', 'are', 'you', 'doing']
In [ ]:
                                                                                                  M
#Processing the comments in such a way that they contains only alphabet
In [30]:
                                                                                                  H
def process content(content):
    return " ".join(re.findall("[A-Za-z]+",content.lower()))
In [31]:
                                                                                                  H
train_data['processed_content'] = train_data['CONTENT'].apply(process_content)
In [32]:
train_data.head()
Out[32]:
                              CONTENT CLASS
                                                                   processed_content
         Huh, anyway check out this you[tube]
                                                huh anyway check out this you tube channel
0
                              channel: ...
      Hey guys check out my new channel and
                                                hey guys check out my new channel and our
1
                               our firs...
2
         just for test I have to say murdev.com
                                             1
                                                      just for test i have to say murdev com
       me shaking my sexy ass on my channel
                                                    me shaking my sexy ass on my channel
3
                              enjoy ^ ^
                                                                               enjoy
4
      watch?v=vtaRGgvGtWQ Check this out .
                                             1
                                                        watch v vtarggvgtwq check this out
```

```
H
In [33]:
drop_fectures(['CONTENT'],train_data)
In [ ]:
                                                                                        H
##Splitting the whole data into train and test sets
In [34]:
from sklearn.model selection import train test split
x_train, x_test, y_train, y_test = train_test_split(train_data['processed_content'],trai
In [35]:
                                                                                        H
from sklearn.feature_extraction.text import CountVectorizer
In [36]:
                                                                                        H
count vect = CountVectorizer(stop words='english')
x train counts = count vect.fit transform(x train)
In [37]:
x_train_counts.shape
Out[37]:
(1564, 3345)
In [ ]:
                                                                                        H
## From occurrences to frequencies
##Occurrence count is a good start but there is an issue: longer documents will have high
##To avoid these potential discrepancies it suffices to divide the number of occurrences
##Another refinement on top of tf is to downscale weights for words that occur in many d
##This downscaling is called tf-idf for "Term Frequency times Inverse Document Frequency
#Both tf and tf-idf can be computed as follows:
In [38]:
from sklearn.feature_extraction.text import TfidfTransformer
tranformer = TfidfTransformer()
x train tfidf = tranformer.fit transform(x train counts)
x_train_tfidf.shape
Out[38]:
(1564, 3345)
```

```
H
In [39]:
x_test_counts = count_vect.transform(x_test)
x_test_counts
Out[39]:
<392x3345 sparse matrix of type '<class 'numpy.int64'>'
        with 2384 stored elements in Compressed Sparse Row format>
In [40]:
                                                                                        H
x_test_tfidf = tranformer.transform(x_test_counts)
x test tfidf
Out[40]:
<392x3345 sparse matrix of type '<class 'numpy.float64'>'
        with 2384 stored elements in Compressed Sparse Row format>
                                                                                        H
In [41]:
from sklearn.linear model import LogisticRegression
model = LogisticRegression()
model.fit(x_train_tfidf,y_train)
Out[41]:
LogisticRegression()
In [42]:
                                                                                        M
predictions = model.predict(x_test_tfidf)
In [43]:
from sklearn.metrics import confusion matrix, classification report, accuracy score
In [44]:
                                                                                        H
confusion_matrix(y_test,predictions)
Out[44]:
array([[176, 5],
       [ 24, 187]], dtype=int64)
In [ ]:
                                                                                        H
#The above confusion matrix states that we classified (176 + 187) = 363 out of 392 comme
```

```
In [45]:
                                                                                          H
print(classification_report(y_test,predictions))
                            recall f1-score
               precision
                                                support
           0
                    0.88
                              0.97
                                         0.92
                                                    181
           1
                    0.97
                              0.89
                                         0.93
                                                    211
                                         0.93
                                                    392
    accuracy
                                                    392
   macro avg
                    0.93
                              0.93
                                         0.93
weighted avg
                    0.93
                              0.93
                                         0.93
                                                    392
In [46]:
                                                                                          H
from sklearn.ensemble import RandomForestClassifier
model = RandomForestClassifier()
model.fit(x_train_tfidf,y_train)
Out[46]:
RandomForestClassifier()
In [56]:
                                                                                          H
predictions = model.predict(x_test_tfidf)
In [57]:
                                                                                          H
confusion_matrix(y_test,predictions)
Out[57]:
array([[170, 11],
       [ 47, 164]], dtype=int64)
In [58]:
                                                                                          H
print(classification_report(y_test,predictions))
               precision
                            recall f1-score
                                                support
           0
                    0.78
                              0.94
                                         0.85
                                                    181
           1
                    0.94
                              0.78
                                         0.85
                                                    211
                                         0.85
                                                    392
    accuracy
                    0.86
                              0.86
                                         0.85
                                                    392
   macro avg
weighted avg
                    0.87
                              0.85
                                         0.85
                                                    392
```

```
In [59]:
                                                                                         H
from sklearn.model_selection import GridSearchCV
parameters = {
                      'max_depth' : [1,3,4],
                      'n estimators': [10,30,50],
                      'max_features': ['sqrt', 'auto', 'log2'],
                      'min_samples_split': [10,20,30],
                      'min_samples_leaf': [1, 3, 10],
                      'bootstrap': [True, False],
model = GridSearchCV(RandomForestClassifier(),parameters)
model.fit(x train tfidf,y train)
Out[59]:
GridSearchCV(estimator=RandomForestClassifier(),
             param_grid={'bootstrap': [True, False], 'max_depth': [1, 3,
4],
                          'max_features': ['sqrt', 'auto', 'log2'],
                          'min samples leaf': [1, 3, 10],
                          'min_samples_split': [10, 20, 30],
                          'n estimators': [10, 30, 50]})
In [60]:
                                                                                         M
model.best_params_
Out[60]:
{'bootstrap': False,
 'max_depth': 3,
 'max_features': 'sqrt',
 'min samples leaf': 3,
 'min_samples_split': 30,
 'n_estimators': 50}
In [64]:
                                                                                         H
predictions = model.predict(x_test_tfidf)
                                                                                         H
In [65]:
confusion_matrix(y_test,predictions)
Out[65]:
array([[173, 8],
       [ 47, 164]], dtype=int64)
```

In [66]:	H									
<pre>print(classification_report(y_test,predictions))</pre>										
	precision	recall	f1-score	support						
0	0.79	0.96	0.86	181						
1	0.95	0.78	0.86	211						
accuracy			0.86	392						
macro avg	0.87	0.87	0.86	392						
weighted avg	0.88	0.86	0.86	392						
In []:						H				
In []:						H				