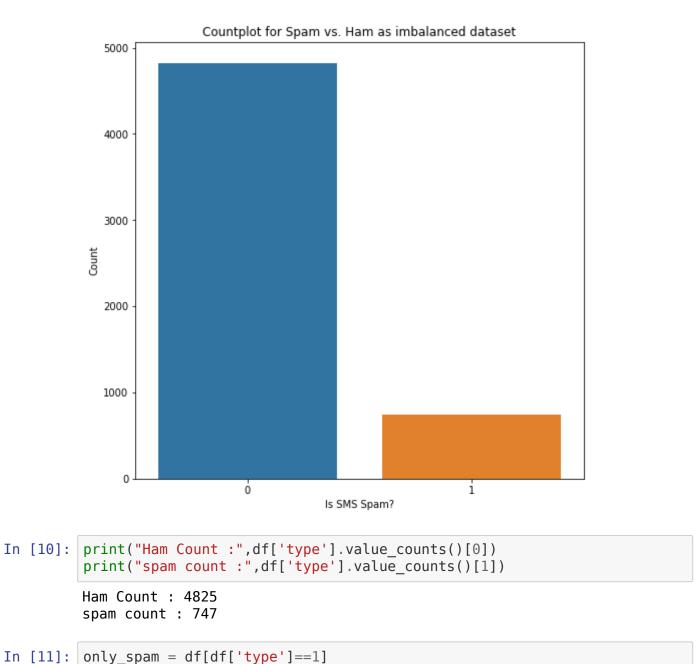
```
In [1]: import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         import seaborn as sns
         %matplotlib inline
         import nltk
         import re
         #nltk.download ('wordnet')
         from nltk.corpus import stopwords
         from nltk.stem import WordNetLemmatizer
         from sklearn.feature extraction.text import TfidfVectorizer
         from sklearn.model selection import cross val score
         from sklearn.metrics import classification report, confusion matrix
         from sklearn.model selection import train test split
         from sklearn.naive bayes import MultinomialNB
         from sklearn.tree import DecisionTreeClassifier
         from sklearn.ensemble import RandomForestClassifier
In [2]: df=pd.read csv('D:\DP Neurotech internship\spam sms collection', sep='
         \t', names=['type', 'message'])
In [3]: df.head()
Out[3]:
             type
                                             message
            ham
                    Go until jurong point, crazy.. Available only ...
             ham
                                  Ok lar... Joking wif u oni...
          2 spam Free entry in 2 a wkly comp to win FA Cup fina...
                  U dun say so early hor... U c already then say...
                    Nah I don't think he goes to usf, he lives aro...
             ham
```

```
In [4]: df.shape
Out[4]: (5572, 2)
In [5]: df.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5572 entries, 0 to 5571
         Data columns (total 2 columns):
                      5572 non-null object
         type
                      5572 non-null object
         message
         dtypes: object(2)
         memory usage: 87.2+ KB
In [6]: df.describe()
Out[6]:
                 type
                             message
            count 5572
                                5572
                    2
                                5169
           unique
                 ham Sorry, I'll call later
             freq 4825
                                  30
In [7]:
         df['type'] = df['type'].map({'ham': 0, 'spam': 1})
In [8]: df
Out[8]:
                type
                                                  message
                       Go until jurong point, crazy.. Available only ...
             0
                  0
                  0
             1
                                      Ok lar... Joking wif u oni...
                  1 Free entry in 2 a wkly comp to win FA Cup fina...
             2
                  0 U dun say so early hor... U c already then say...
             3
```

type messag	type	type	
0 Nah I don't think he goes to usf, he lives aro	0	4	
1 This is the 2nd time we have tried 2 contact u	1	5567	
0 Will ü b going to esplanade fr home	0	5568	
0 Pity, * was in mood for that. Soany other s	0	5569	
0 The guy did some bitching but I acted like i'd	0	5570	
0 Rofl. Its true to its nam	0	5571	

5572 rows × 2 columns

```
In [9]: plt.figure(figsize=(8,8))
   g = sns.countplot(x='type', data=df)
   p = plt.title('Countplot for Spam vs. Ham as imbalanced dataset')
   p = plt.xlabel('Is SMS Spam?')
   p = plt.ylabel('Count')
```



count = int((df.shape[0]-only_spam.shape[0])/only_spam.shape[0])

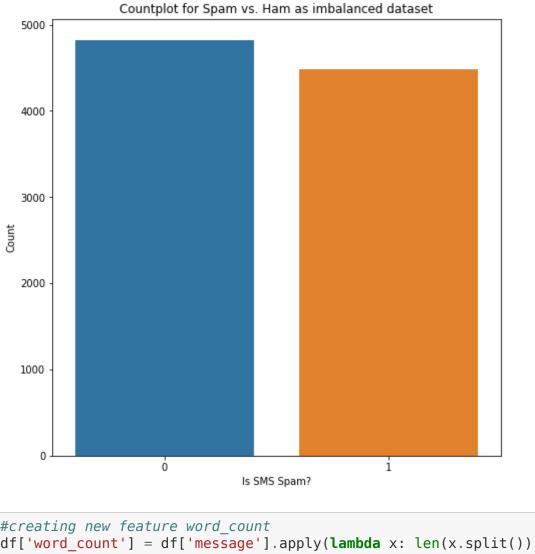
for i in range(0, count-1):

```
df = pd.concat([df,only_spam])

df.shape

Out[11]: (9307, 2)

In [12]: plt.figure(figsize=(8,8))
    g = sns.countplot(x='type', data=df)
    p = plt.title('Countplot for Spam vs. Ham as imbalanced dataset')
    p = plt.xlabel('Is SMS Spam?')
    p = plt.ylabel('Count')
```



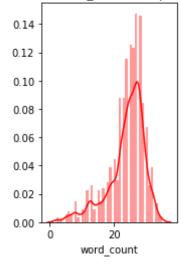
type		message	word_count
1	0	Ok lar Joking wif u oni	6
2	1	Free entry in 2 a wkly comp to win FA Cup fina	28
3	0	U dun say so early hor U c already then say	11
4	0	Nah I don't think he goes to usf, he lives aro	13

```
In [15]: plt.figure(figsize=(12, 6))
```

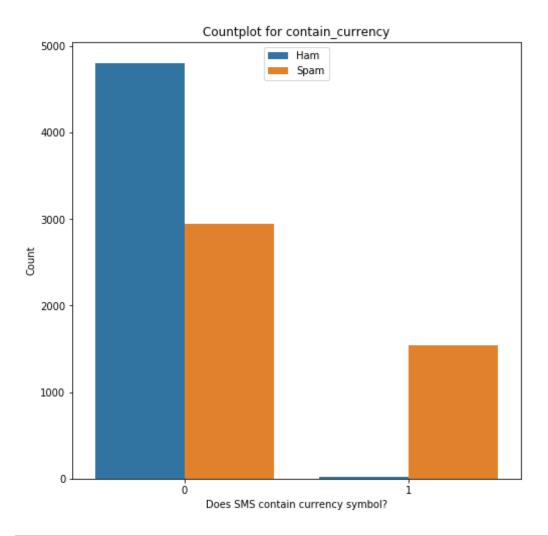
```
In [16]: plt.subplot(1,2,2)
    g = sns.distplot(a=df[df['type']==1].word_count, color='red')
    p = plt.title('Distribution of word_count for Spam messages')

plt.tight_layout()
    plt.show()
```

Distribution of word_count for Spam messages



```
In [17]: def currency(x):
             currency_symbols = ['€', '$', '¥', '₹', '£']
             for i in currency symbols:
               if i in x:
                  return 1
             return 0
          df['contains currency symbol'] = df['message'].apply(currency)
In [18]: df.head()
Out[18]:
                                               message word count contains currency symbol
              type
                                                                                        0
                     Go until jurong point, crazy.. Available only ...
                                                                20
           1
                0
                                   Ok lar... Joking wif u oni...
                                                                6
                1 Free entry in 2 a wkly comp to win FA Cup fina...
           2
                                                                28
           3
                0 U dun say so early hor... U c already then say...
                                                                11
                                                                                        0
                     Nah I don't think he goes to usf, he lives aro...
                                                                13
                0
                                                                                        0
In [19]: plt.figure(figsize=(8,8))
          g = sns.countplot(x='contains currency symbol', data=df, hue='type')
          p = plt.title('Countplot for contain currency')
          p = plt.xlabel('Does SMS contain currency symbol?')
          p = plt.ylabel('Count')
          p = plt.legend(labels=['Ham', 'Spam'], loc=9)
```



```
In [20]: #cleaning the messages
    corpus = []
    wnl = WordNetLemmatizer()

for sms_string in list(df.message):
    #cleaning special character from the sms
    message = re.sub(pattern='[^a-zA-Z]', repl=' ', string=sms_string)
```

```
#converting the entire sms into lower case
           message = message.lower()
           #Tokenizing the sms by words
           words = message.split()
           #Removing the stop words
           filtered words = [word for word in words if word not in set (stopword
         s.words('english'))]
           #lemmatizing the words
           lemmatized words = [wnl.lemmatize(word) for word in filtered words]
           #joining the lemmatized words
           message = ' '.join(lemmatized words)
           #building a corpus of messages
           corpus.append(message)
In [21]: corpus[0:5]
Out[21]: ['go jurong point crazy available bugis n great world la e buffet cine
         got amore wat',
          'ok lar joking wif u oni',
          'free entry wkly comp win fa cup final thts st may text fa receive ent
         ry question std txt rate c apply',
          'u dun say early hor u c already say',
          'nah think go usf life around though']
In [22]: tfidf = TfidfVectorizer(max features=500)
         vectors = tfidf.fit transform(corpus).toarray()
         feature names = tfidf.get feature names()
In [23]: #Extracting independent and dependent variables from the dataset
         x = pd.DataFrame(vectors, columns=feature names)
         y = df['type']
In [24]: x train, x test, y train, y test = train test split(x, y, test size=0.3
```

```
, random state=42)
In [25]: mnb = MultinomialNB()
         cv = cross val score(mnb, x, y, scoring='f1', cv=10)
         print('--- Average F1-Score for MNB model: {}---'.format(round(cv.mean
         (), 3)))
         print('Standard Deviation: {}'.format(round(cv.std(), 3)))
         --- Average F1-Score for MNB model: 0.943---
         Standard Deviation: 0.004
In [28]: mnb.fit(x train, y train)
         y pred = mnb.predict(x test)
         print('--- Classification report for MNB model ---')
         print(classification report(y test, y pred))
         --- Classification report for MNB model ---
                       precision
                                    recall f1-score
                                                       support
                    0
                            0.95
                                      0.94
                                                0.95
                                                          1457
                            0.94
                                      0.94
                                                0.94
                    1
                                                          1336
                                                0.94
                                                          2793
             accuracy
                            0.94
                                      0.94
                                                0.94
                                                          2793
            macro avq
         weighted avg
                            0.94
                                                          2793
                                      0.94
                                                0.94
```