```
In [1]:
         import nltk
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
         from sklearn.naive_bayes import MultinomialNB
         from sklearn.model selection import train test split
         from nltk.corpus import stopwords
         from nltk.stem import PorterStemmer,WordNetLemmatizer
         from nltk.tokenize import word_tokenize
         import sklearn.metrics as m
         from sklearn.linear_model import LogisticRegression
         from sklearn.svm import SVC
         from sklearn.tree import DecisionTreeClassifier
         dataset = pd.read_csv("SMSSpamCollection", sep='\t', names=['label', 'message'])
In [2]:
In [3]:
         dataset
Out[3]:
                label
                                                     message
             0
                 ham
                         Go until jurong point, crazy.. Available only ...
             1
                                        Ok lar... Joking wif u oni...
                 ham
             2
                spam Free entry in 2 a wkly comp to win FA Cup fina...
             3
                       U dun say so early hor... U c already then say...
                 ham
                        Nah I don't think he goes to usf, he lives aro...
             4
                 ham
                       This is the 2nd time we have tried 2 contact u...
          5567
                spam
          5568
                                Will ü b going to esplanade fr home?
                 ham
          5569
                        Pity, * was in mood for that. So...any other s...
                 ham
          5570
                        The guy did some bitching but I acted like i'd...
                 ham
          5571
                 ham
                                          Rofl. Its true to its name
         5572 rows × 2 columns
In [4]: | dataset.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 5572 entries, 0 to 5571
         Data columns (total 2 columns):
               Column
                         Non-Null Count Dtype
          0
               label
                         5572 non-null
                                            object
          1
               message 5572 non-null
                                            object
         dtypes: object(2)
         memory usage: 87.2+ KB
         dataset.describe()
In [5]:
Out[5]:
                  label
                              message
                  5572
                                  5572
           count
          unique
                     2
                                  5169
```

Sorry, I'll call later

30

top

freq 4825

ham

```
In [6]: dataset['label'] = dataset['label'].map({'ham': 0, 'spam': 1})
```

In [7]: dataset

Out[7]:

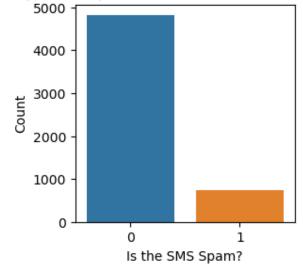
	label	message
0	0	Go until jurong point, crazy Available only
1	0	Ok lar Joking wif u oni
2	1	Free entry in 2 a wkly comp to win FA Cup fina
3	0	U dun say so early hor U c already then say
4	0	Nah I don't think he goes to usf, he lives aro
5567	1	This is the 2nd time we have tried 2 contact u
5568	0	Will ü b going to esplanade fr home?
5569	0	Pity, * was in mood for that. Soany other s
5570	0	The guy did some bitching but I acted like i'd
5571	0	Rofl. Its true to its name

5572 rows × 2 columns

```
In [8]: import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
In [9]: plt.figure(figsize=(3, 3))
    g = sns.countplot(x="label", data=dataset)
    p = plt.title('Countplot for Spam vs Ham as inmbalanced dataset')
    p = plt.xlabel('Is the SMS Spam?')
    p = plt.ylabel('Count')
```

Countplot for Spam vs Ham as inmbalanced dataset



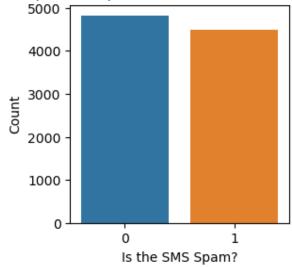
```
In [10]: only_spam = dataset[dataset["label"] == 1]
```

```
Out[11]:
                  label
                                                                 message
               2
                     1
                                 Free entry in 2 a wkly comp to win FA Cup fina...
               5
                     1
                                 FreeMsg Hey there darling it's been 3 week's n...
               8
                     1
                             WINNER!! As a valued network customer you have...
               9
                     1
                               Had your mobile 11 months or more? U R entitle...
              11
                     1
                              SIX chances to win CASH! From 100 to 20,000 po...
            5537
                     1
                              Want explicit SEX in 30 secs? Ring 02073162414...
                        ASKED 3MOBILE IF 0870 CHATLINES INCLU IN FREE ...
            5540
            5547
                     1
                               Had your contract mobile 11 Mnths? Latest Moto...
                            REMINDER FROM O2: To get 2.50 pounds free call...
            5566
                     1
            5567
                                  This is the 2nd time we have tried 2 contact u...
           747 rows × 2 columns
           print("Number of Spam SMS:", len(only_spam))
In [12]:
           print("Number of Ham SMS:", len(dataset) - len(only_spam))
           Number of Spam SMS: 747
           Number of Ham SMS: 4825
In [13]: count = int((dataset.shape[0] - only_spam.shape[0]) // only_spam.shape[0])
In [14]: count
Out[14]: 6
In [15]: for i in range(0, count-1):
                dataset = pd.concat([dataset, only_spam])
           dataset.shape
Out[15]: (9307, 2)
```

In [11]: only_spam

```
In [16]: plt.figure(figsize=(3, 3))
    g = sns.countplot(x="label", data=dataset)
    p = plt.title('Countplot for Spam vs Ham as balanced dataset')
    p = plt.xlabel('Is the SMS Spam?')
    p = plt.ylabel('Count')
```

Countplot for Spam vs Ham as balanced dataset



```
In [17]: dataset['word_count'] = dataset['message'].apply(lambda x: len(x.split()))
```

In [18]: dataset

Out[18]:

	label	message	word_count
0	0	Go until jurong point, crazy Available only	20
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
5537	1	Want explicit SEX in 30 secs? Ring 02073162414	16
5540	1	ASKED 3MOBILE IF 0870 CHATLINES INCLU IN FREE \dots	33
5547	1	Had your contract mobile 11 Mnths? Latest Moto	28
5566	1	REMINDER FROM O2: To get 2.50 pounds free call	28
5567	1	This is the 2nd time we have tried 2 contact u	30

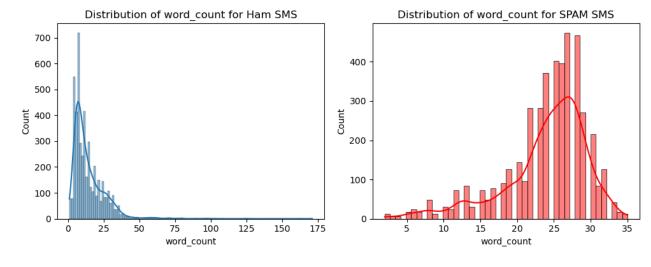
9307 rows × 3 columns

```
In [19]: plt.figure(figsize=(10,4))

# (1,1)
plt.subplot(1,2,1)
g = sns.histplot(dataset[dataset["label"]==0].word_count, kde=True)
p = plt.title('Distribution of word_count for Ham SMS')

# (1,2)
plt.subplot(1,2,2)
g = sns.histplot(dataset[dataset["label"]==1].word_count, color="red", kde=True)
p = plt.title('Distribution of word_count for SPAM SMS')

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



```
In [20]: # Creating new feature of containing currency symbol
    def currency(data):
        currency_symbols = ['€', '$', '£', '¥', '₹']
        for i in currency_symbols:
            if i in data:
                return 1
        return 0
```

In [21]: dataset["contains_currency_symbols"] = dataset["message"].apply(currency)

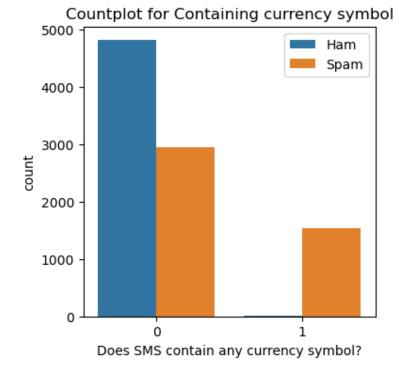
In [22]: dataset

Out[22]:

	label	message	word_count	contains_currency_symbols
0	0	Go until jurong point, crazy Available only	20	0
1	0	Ok lar Joking wif u oni	6	0
2	1	Free entry in 2 a wkly comp to win FA Cup fina	28	0
3	0	U dun say so early hor U c already then say	11	0
4	0	Nah I don't think he goes to usf, he lives aro	13	0
5537	1	Want explicit SEX in 30 secs? Ring 02073162414	16	0
5540	1	ASKED 3MOBILE IF 0870 CHATLINES INCLU IN FREE \dots	33	1
5547	1	Had your contract mobile 11 Mnths? Latest Moto	28	0
5566	1	REMINDER FROM O2: To get 2.50 pounds free call	28	0
5567	1	This is the 2nd time we have tried 2 contact u	30	1

9307 rows × 4 columns

```
In [23]: # Countplot for contains_currency_symbols
plt.figure(figsize=(4,4))
g = sns.countplot(x = 'contains_currency_symbols', data = dataset, hue = "label")
p = plt.title('Countplot for Containing currency symbol')
p = plt.xlabel('Does SMS contain any currency symbol?')
p = plt.ylabel('count')
p = plt.legend(labels=["Ham", "Spam"], loc=0)
```

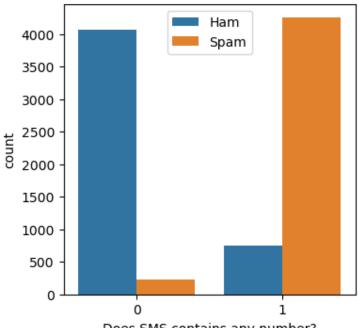


```
In [24]: # creating new feature of containing numbers
def number(data):
    for i in data:
        if ord(i) >= 48 and ord(i) <= 57:
            return 1
    return 0</pre>
```

In [25]: dataset["contains_number"] = dataset["message"].apply(number)

```
In [26]: # Countplot for containing numbers
plt.figure(figsize=(4,4))
g = sns.countplot(x = 'contains_number', data = dataset, hue = "label")
p = plt.title('Countplot for Containing Numbers')
p = plt.xlabel('Does SMS contains any number?')
p = plt.ylabel('count')
p = plt.legend(labels=["Ham", "Spam"], loc=0)
```

Countplot for Containing Numbers



Does SMS contains any number?

```
In [27]: import nltk
   import re
   nltk.download('stopwords')
   nltk.download('wordnet')
   from nltk.corpus import stopwords
   from nltk.stem import WordNetLemmatizer
```

```
In [28]: corpus = []
wnl = WordNetLemmatizer()

for sms in list(dataset.message):
    message = re.sub(pattern='[^A-Za-z]', repl= ' ', string=sms) # Filtering out special
    message = message.lower()
    words = message.split() # Tokenizer
    filtered_words = [word for word in words if word not in set(stopwords.words("english
    lenm_words = [wnl.lemmatize(word) for word in filtered_words]
    message = ' '.join(lenm_words)

corpus.append(message)
```

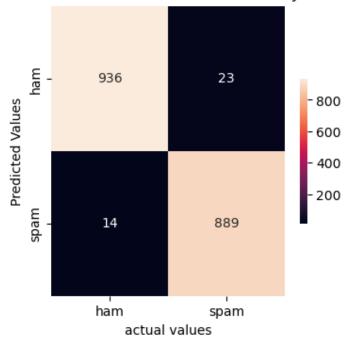
```
In [29]: corpus
Out[29]: ['go jurong point crazy available bugis n great world la e buffet cine got amore wat',
           'ok lar joking wif u oni',
          'free entry wkly comp win fa cup final tkts st may text fa receive entry question std
         txt rate c apply',
          'u dun say early hor u c already say',
          'nah think go usf life around though',
          'freemsg hey darling week word back like fun still tb ok xxx std chgs send rcv',
          'even brother like speak treat like aid patent',
          'per request melle melle oru minnaminunginte nurungu vettam set callertune caller pre
         ss copy friend callertune',
          'winner valued network customer selected receivea prize reward claim call claim code
         kl valid hour',
          'mobile month u r entitled update latest colour mobile camera free call mobile update
         co free',
           'gonna home soon want talk stuff anymore tonight k cried enough today',
          'six chance win cash pound txt csh send cost p day day tsandcs apply reply hl info',
          'urgent week free membership prize jackpot txt word claim c www dbuk net lccltd pobox
         ldnw rw',
           'searching right word thank breather promise wont take help granted fulfil promise wo
In [30]: from sklearn.feature_extraction.text import TfidfVectorizer
         tfidf=TfidfVectorizer(max_features=5000)
         features=tfidf.fit_transform(corpus)
         features=features.toarray()
         features
Out[30]: array([[0., 0., 0., ..., 0., 0., 0.],
                [0., 0., 0., ..., 0., 0., 0.]
                [0., 0., 0., \ldots, 0., 0., 0.]
                [0., 0., 0., \ldots, 0., 0., 0.]
                [0., 0., 0., \ldots, 0., 0., 0.]
                [0., 0., 0., ..., 0., 0., 0.]
In [31]: len(tfidf.get_feature_names_out())
Out[31]: 5000
In [32]: | feature_names = tfidf.get_feature_names_out()
In [33]: x = pd.DataFrame(features, columns = feature names)
         y = dataset['label']
In [34]: from sklearn.model selection import cross val score, train test split
         from sklearn.metrics import classification report, confusion matrix
In [35]: x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2, random_state=42
```

```
Out[36]:
                                     aberdeen abi ability abiola abj able ... zebra zed zero
                          aathi
                                                                                                  zf zhong zindgi
                  aa
                      aah
                                 ab
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                                                                  0.0
            3533 0.0
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                             0.0
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            2592 0.0
                      0.0
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                             0.0 0.0
            4253 0.0
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            6976 0.0
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                      0.0
                             0.0 0.0
                                           0.0
                                               0.0
                                                                  0.0
                                                                                                               0.0
            5734 0.0
                                                       0.0
                                                              0.0
                                                                        0.0
                                                                                  0.0
                                                                                       0.0
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                                                                                                 0.0
                                                                                                        0.0
            5191 0.0
                      0.0
                             0.0 0.0
                                           0.0
                                               0.0
                                                       0.0
                                                                  0.0
                                                                        0.0 ...
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                                                                                                               0.0
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                                                                                  0.0
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            5390 0.0
                      0.0
                             0.0 0.0
                                           0.0
                                               0.0
                                                       0.0
                                                              0.0
                                                                  0.0
                                                                        0.0
                                                                                  0.0
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                                                                                            0.0
                                                                                                 0.0
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             860
                 0.0
                      0.0
                             0.0 0.0
                                           0.0
                                               0.0
                                                       0.0
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                                                                  0.0
                                                                        0.0
                                                                                  0.0
                                                                                       0.0
                                                                                            0.0
                                                                                                 0.0
                                                                                                        0.0
                                                                                                               0.0
            7270 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 0.0
                                                                                                        0.0
                                                                                                               0.0
           7445 rows × 5000 columns
In [37]:
          # Naive Bayes Model
           from sklearn.naive bayes import MultinomialNB
           mnb = MultinomialNB()
           cv = cross_val_score(mnb, x, y, scoring='f1', cv=10)
           print(mnb)
           print(cv)
           MultinomialNB()
           [0.97674419 0.97900552 0.9810901 0.98447894 0.98004435 0.98218263
            0.97900552 0.98113208 0.98342541 0.9844098 ]
In [38]:
          print(cv.std)
           <built-in method std of numpy.ndarray object at 0x000002D4FAD01C30>
In [39]:
          print(round(cv.mean(),3))
           print(round(cv.std(),3))
           0.981
           0.002
In [40]:
          mnb.fit(x_train, y_train)
           y pred = mnb.predict(x test)
In [41]:
          print(classification_report(y_test, y_pred))
                           precision
                                           recall f1-score
                                                                 support
                       0
                                 0.99
                                             0.98
                                                         0.98
                                                                      959
                        1
                                 0.97
                                             0.98
                                                         0.98
                                                                      903
                                                         0.98
                                                                     1862
               accuracy
              macro avg
                                 0.98
                                             0.98
                                                         0.98
                                                                     1862
           weighted avg
                                 0.98
                                             0.98
                                                         0.98
                                                                     1862
```

In [36]:

x_train

Confusion Matrix of Multinomial Naive Bayes Model



```
In [44]: from sklearn.tree import DecisionTreeClassifier
    dt = DecisionTreeClassifier()
    cv1 = cross_val_score(dt, x, y, scoring='f1', cv=10)
    print(round(cv.mean(),3))
    print(round(cv.std(),3))

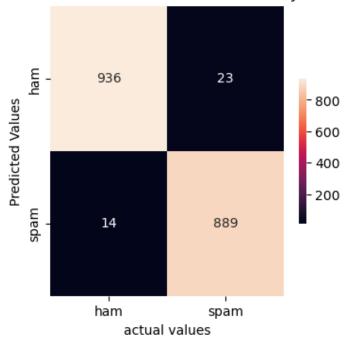
    0.981
    0.002

In [45]: dt.fit(x_train, y_train)
    y_pred1 = dt.predict(x_test)
```

<pre>In [46]: print(classification_report(y_test, y_pred))</pre>						
	precision	recall	f1-score	support		

0	0.99	0.98	0.98	959
1	0.97	0.98	0.98	903
accuracy macro avg	0.98 0.98	0.98 0.98	0.98 0.98 0.98	1862 1862 1862
weighted avg	0.98	0.98	0.98	1802

Confusion Matrix of Multinomial Naive Bayes Model



```
In [49]: def predict_spam(sms):
    message = re.sub(pattern='[^A-Za-z]', repl= ' ', string=sms) # Filtering out special
    message = message.lower()
    words = message.split() # Tokenizer
    filtered_words = [word for word in words if word not in set(stopwords.words("english
    lenm_words = [wnl.lemmatize(word) for word in filtered_words]
    message = ' '.join(lenm_words)
    temp = tfidf.fit_transform([message]).toarray()
    return dt.predict(temp)
```