Email spam Detection with Machine Learning

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In [1]:
         import pandas as pd
            import nltk
            from nltk.corpus import stopwords
            import string
            df = pd.read_csv("emails.csv")
            df.head()
   Out[1]:
                                                  text spam
                    Subject: naturally irresistible your corporate...
                                                          1
             1
                    Subject: the stock trading gunslinger fanny i...
                                                          1
               Subject: unbelievable new homes made easy im ...
                                                          1
             3
                    Subject: 4 color printing special request add...
                                                          1
                 Subject: do not have money, get software cds ...
                                                          1
            df.shape
In [2]:
   Out[2]: (5728, 2)
         ▶ df.columns
In [3]:
   Out[3]: Index(['text', 'spam'], dtype='object')
         In [4]:
            print(df.shape)
            (5695, 2)
In [5]:
         # to show the number of missing data
            print(df.isnull().sum())
            text
                     0
                     0
            spam
            dtype: int64
         ▶ # downLoad the stopwords package
In [6]:
            nltk.download("stopwords")
            [nltk_data] Downloading package stopwords to
                             C:\Users\Aniket\AppData\Roaming\nltk_data...
            [nltk data]
            [nltk data]
                           Package stopwords is already up-to-date!
   Out[6]: True
```

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In [7]:
         ▶ !pip install stopwords
             Requirement already satisfied: stopwords in c:\users\aniket\anaconda3\lib
             \site-packages (1.0.1)
 In [8]: ▶ def process(text):
                 nopunc = [char for char in text if char not in string.punctuation]
                 nopunc = ''.join(nopunc)
                 clean = [word for word in nopunc.split() if word.lower() not in stopword
                 return clean
             # to show the tokenization
             df['text'].head().apply(process)
    Out[8]: 0
                  [Subject, naturally, irresistible, corporate, ...
                  [Subject, stock, trading, gunslinger, fanny, m...
                  [Subject, unbelievable, new, homes, made, easy...
                  [Subject, 4, color, printing, special, request...
                  [Subject, money, get, software, cds, software,...
             Name: text, dtype: object
 In [9]: ▶ from sklearn.feature_extraction.text import CountVectorizer
             message = CountVectorizer(analyzer=process).fit_transform(df['text'])
         #split the data into 80% training and 20% testing
In [10]:
             from sklearn.model_selection import train_test_split
             xtrain, xtest, ytrain, ytest = train_test_split(message, df['spam'], test_s
             print(message.shape)
             (5695, 37229)
In [11]: ▶ # create and train the Naive Bayes Classifier
             from sklearn.naive_bayes import MultinomialNB
             classifier = MultinomialNB().fit(xtrain, ytrain)
In [12]: | print(classifier.predict(xtrain))
             [0 0 0 ... 0 0 0]
In [13]: | print(ytrain.values)
             [0 0 0 ... 0 0 0]
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In [14]:

    ₩ Evaluating the model on the training data set

             from sklearn.metrics import classification_report, confusion_matrix, accura
             pred = classifier.predict(xtrain)
             print(classification report(ytrain, pred))
             print()
             print("Confusion Matrix: \n", confusion_matrix(ytrain, pred))
             print("Accuracy: \n", accuracy score(ytrain, pred))
                            precision
                                         recall f1-score
                                                            support
                        0
                                 1.00
                                           1.00
                                                     1.00
                                                                3457
                        1
                                 0.99
                                                     0.99
                                                                1099
                                           1.00
                                                     1.00
                                                               4556
                 accuracy
                                 0.99
                                           1.00
                                                     1.00
                                                                4556
                macro avg
             weighted avg
                                 1.00
                                           1.00
                                                     1.00
                                                                4556
             Confusion Matrix:
              [[3445
                       12]
                  1 1098]]
             Accuracy:
              0.9971466198419666
          #print the predictions
In [15]:
             print(classifier.predict(xtest))
             #print the actual values
             print(ytest.values)
             [1 0 0 ... 0 0 0]
             [1 0 0 ... 0 0 0]
          # Evaluating the model on the training data set
In [16]:
             from sklearn.metrics import classification_report, confusion_matrix, accura
             pred = classifier.predict(xtest)
             print(classification_report(ytest, pred))
             print()
             print("Confusion Matrix: \n", confusion_matrix(ytest, pred))
             print("Accuracy: \n", accuracy_score(ytest, pred))
                            precision
                                         recall f1-score
                                                            support
                        0
                                 1.00
                                           0.99
                                                     0.99
                                                                 870
                                 0.97
                        1
                                           1.00
                                                     0.98
                                                                 269
                                                     0.99
                                                                1139
                 accuracy
                                 0.98
                                           0.99
                                                     0.99
                                                                1139
                macro avg
             weighted avg
                                 0.99
                                           0.99
                                                     0.99
                                                                1139
             Confusion Matrix:
                      8]
              [[862
              [ 1 268]]
```

Accuracy:

0.9920983318700615

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