→ This program detects if an email is spam or not/ham.

Import necessary libraries

```
import pandas as pd
from sklearn.feature_extraction.text import CountVectorizer
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import MultinomialNB
from sklearn.metrics import accuracy_score

import numpy as np
import pandas as pd
import nltk
from nltk.corpus import stopwords
import string
```

Load the dataset

₽		Category	Message
	0	ham	Go until jurong point, crazy Available only
	1	ham	Ok lar Joking wif u oni
	2	spam	Free entry in 2 a wkly comp to win FA Cup fina
	3	ham	U dun say so early hor U c already then say
	4	ham	Nah I don't think he goes to usf, he lives aro

```
#Get the column names df.columns
```

```
Index(['Category', 'Message'], dtype='object')
```

```
#check for duplicates and remove them
df.drop_duplicates(inplace = True)
```

#show the new shape(new number of rows and columns)
df.shape

```
(5157, 2)
```

```
\#show the number of missing data (like - NAN, Nan, na) data for each column df.isnull().sum()
```

```
Category 0
Message 0
dtype: int64
```

Download the stopwords package

```
#Download the stopwords package
nltk.download('stopwords')
```

```
[nltk_data] Downloading package stopwords to /root/nltk_data...
[nltk_data] Unzipping corpora/stopwords.zip.
True

def process_text(text):
    #1.Remove the punctuation
    #2.Remove the stopwords
    #3.Return a list of clean text words

#1
    nopunc = [char for char in text if char not in string.punctuation]
    nopunc = ''.join(nopunc)

#2
    clean_words = [word for word in nopunc.split() if word.lower() not in stopwords.words('english')]
#3
    return clean_words
```

Tokenization

```
#Show the tokenization (a list of tokens also called lemmas)
df['Message'].head().apply(process_text)
          [Go, jurong, point, crazy, Available, bugis, n...
[Ok, lar, Joking, wif, u, oni]
          [Free, entry, 2, wkly, comp, win, FA, Cup, fin...
              [U, dun, say, early, hor, U, c, already, say]
          [Nah, dont, think, goes, usf, lives, around, t...
     4
     Name: Message, dtype: object
#Convert a collection of text to a matrix of tokens
from sklearn.feature_extraction.text import CountVectorizer
messages_bow = CountVectorizer(analyzer=process_text).fit_transform(df['Message'])
#Split the data into 80% training and 20% testing
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(messages_bow, df['Category'], test_size=0.20, random_state=0)
#Get the shape of messages_bow
messages_bow.shape
     (5157, 11422)
```

Creating and training the Naive Bayes Classifier

```
#Create and train the Naive Bayes Classifier
from sklearn.naive_bayes import MultinomialNB
classifier = MultinomialNB().fit(X_train, y_train)

#Print the predictions
print(classifier.predict(X_train))

#Print the actual values
print(y_train.values)

['ham' 'spam' 'ham' ... 'ham' 'ham' 'ham']
['ham' 'spam' 'ham' ... 'ham' 'ham' 'ham']
```

Model Evaluation on the both training and testing data

Training dataset

```
#Evaluate the model on the training dataset
from sklearn.metrics import classification_report, confusion_matrix, accuracy_score
pred = classifier.predict(X_train)
print(classification_report(y_train, pred))
print()
print('Confusion Matrix: \n', confusion_matrix(y_train, pred))
print('Accuracy:', accuracy_score(y_train, pred))
```

```
precision
                         recall f1-score
                                            support
                  1.00
                            1.00
                                      1.00
                                                3619
         ham
                  0.98
                            0.97
                                                 506
        spam
                                      0.98
    accuracy
                                      0.99
                                                4125
   macro avg
                  0.99
                            0.99
                                      0.99
                                                4125
weighted avg
                  0.99
                            0.99
                                      0.99
                                                4125
Confusion Matrix:
          81
 [[3611
 [ 13 493]]
Accuracy: 0.9949090909090909
```

After evaluating the model om the training data the model got an accuracy of 99.49%

```
#Print the predictions
print(classifier.predict(X_test))

#Print the actual values
print(y_test.values)

['ham' 'ham' 'ham' ... 'ham' 'ham' 'ham']
['ham' 'ham' 'ham' ... 'ham' 'ham' 'ham']
```

Testing dataset

```
#Evaluate the model on the testing dataset
from \ sklearn.metrics \ import \ classification\_report, \ confusion\_matrix, \ accuracy\_score
pred = classifier.predict(X_test)
print(classification_report(y_test, pred))
print()
print('Confusion Matrix: \n', confusion_matrix(y_test, pred))
print('Accuracy:', accuracy_score(y_test, pred))
                   precision
                                recall f1-score
                                                    support
                        0.99
                                  0.97
                                                        897
                                             0.98
              ham
                        0.81
                                             0.86
             spam
                                  0.93
                                                        135
         accuracy
                                             0.96
                                                       1032
        macro avg
                        0.90
                                  0.95
                                             0.92
                                                       1032
```

0.96

1032

```
Confusion Matrix:
[[867 30]
[ 10 125]]
Accuracy: 0.9612403100775194
```

0.96

weighted avg

After evaluating the model om the testing data the model got an accuracy of 96.12%

0.96