# IBM NAAN MUDHALVAN

## PHASE 5 SUBMISSION

SPAM CLASSIFIER USING AI

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## INTRODUCTION

In today's globalized world, email is a primary source of communication. This communication can vary from personal, business, corporate to government.SPAM emails, also known as junk mail involves nearly identical messages sent to numerous recipients by email. We try to identify patterns using Data-mining classification algorithms to enable us classifythe emails as HAM or SPAM.

### PROBLEM STATEMENT

Unlike emails, which have a variety of large datasets available, real databases for SMS spams are very limited. Additionally, due to small length of text messages, the number of features that can be used for their classification is far smaller than the corresponding number in emails

## **DATASET**

```
#Standard libraries for data analysis:
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
# sklearn modules for data preprocessi
from sklearn.model_selection import t
rain_test_split
from sklearn.naive_bayes import Multi
nomialNB
from sklearn.feature_extraction.text
import CountVectorizer
#sklearn modules for Model Evaluation
```

ecision\_score, recall\_score, fbeta\_sc ore from statsmodels.stats.outliers\_influ ence import variance\_inflation\_factor from sklearn.model\_selection import c ross\_val\_score, GridSearchCV, Shuffle Split, KFold # from sklearn import feature\_selectio n from sklearn import model\_selection # from sklearn import metrics from sklearn.metrics import classific ation\_report, precision\_recall\_curve from sklearn.metrics import auc, roc\_ auc\_score, roc\_curve from sklearn.metrics import make\_scor er, recall\_score, log\_loss from sklearn.metrics import average\_p recision\_score

#Standard libraries for data visualiza

```
import seaborn as sn
from matplotlib import pyplot
import matplotlib.pyplot as plt
import matplotlib.pylab as pylab
import matplotlib
%matplotlib inline
color = sn.color_palette()
import matplotlib.ticker as mtick
from IPython.display import display
pd.options.display.max_columns = None
from pandas.plotting import scatter_m
atrix
from sklearn.metrics import roc_curve
```

#### mport Dataset

```
In [2]:

df = pd.read_csv("/kaggle/input/spam-
email-dataset/emails.csv")

df
```

	text	spam
0	Subject: naturally irresistible your corporate	1
1	Subject: the stock trading gunslinger fanny i	1
2	Subject: unbelievable new homes made easy im	1
3	Subject: 4 color printing special request add	1
4	Subject: do not have money , get software cds	1
	***	***
5723	Subject: re: research and development charges	0
5724	Subject: re : receipts from visit jim , than	О
5725	Subject: re : enron case study update wow ! a	0
5726	Subject: re : interest david , please , call	0
5727	Subject: news : aurora 5 . 2 update aurora ve	О

```
Out[5]:
text
spam
dtype: int64
 In [6]:
df.duplicated().sum()
 Out[6]:
33
 In [7]:
# drop duplicate
df.drop_duplicates(inplace=True)
 In [8]:
# Check Target Variable Distribution
df["spam"].value_counts()
```

```
5724 Subject: re : receipts from vis
it jim , than... 0
5725 Subject: re : enron case study
update wow! a... 0
5726 Subject: re : interest david ,
please , call... 0
5727 Subject: news : aurora 5 . 2 up
date aurora ve... 0
[5728 \text{ rows x 2 columns}] >
 In [4]:
df.dtypes
 Out[4]:
text object
spam int64
dtype: object
 In [5]:
df.isna().sum()
```

```
df["spam"].value_counts()

Out[8]:

spam
0   4327
1   1368
Name: count, dtype: int64
```

In this case, we have class imbalance with few positives. In our business challenge, false negatives are costly. Hence let's keep an eye onto the Precision, Recall & F2 score besides accuracy

#### **Handling Text Data**

```
# clean the text
from nltk.corpus import stopwords
from nltk.tokenize import word_token
```

```
ze
import re
In [10]:
def clean_text(text):
    text=text.lower()
    text=re.sub('[^a-z]',' ',text)
    text=re.sub('subject',' ',text)
    text=word_tokenize(text)
    text=[word for word in text if le
n(word) > 1
    return ' '.join(text)
clean_text('Data clean')
Out[10]:
'data clean'
In [11]:
df['text']=df['text'].apply(clean_tex
t)
```

In [12]:

df

#### Out[12]:

	text	spam
0	naturally irresistible your corporate identity	1
1	the stock trading gunslinger fanny is merrill	1
2	unbelievable new homes made easy im wanting to	1
3	color printing special request additional info	1
4	do not have money get software cds from here s	1
		***
5723	re research and development charges to gpg her	0
5724	re receipts from visit jim thanks again for th	0
5725	re enron case study update wow all on the same	О
5726	re interest david please call shirley crenshaw	0
5727	news aurora update aurora version the fastest	0

```
In [3]:
df.info
 Out[3]:
<body><br/>bound method DataFrame.info of</br/>
text spam
0 Subject: naturally irresistible
your corporate... 1
1 Subject: the stock trading guns
linger fanny i... 1
2 Subject: unbelievable new homes
made easy im ... 1
3 Subject: 4 color printing speci
al request add... 1
4 Subject: do not have money, ge
t software cds ... 1
5723 Subject: re : research and deve
lopment charges... 0
5724 Subject: re : receipts from vi
```

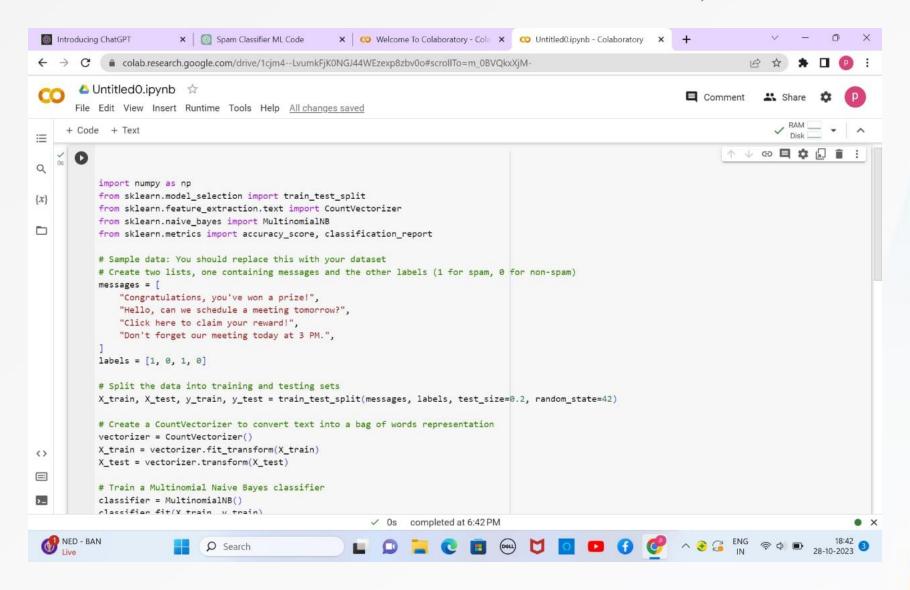
```
In [3]:
df.info
 Out[3]:
<bound method DataFrame.info of</pre>
text spam
0 Subject: naturally irresistible
your corporate... 1
1 Subject: the stock trading guns
linger fanny i... 1
2 Subject: unbelievable new homes
made easy im ... 1
3 Subject: 4 color printing speci
al request add... 1
4 Subject: do not have money, ge
t software cds ... 1
5723 Subject: re : research and deve
lopment charges... 0
5724 Subject: re : receipts from vi
```

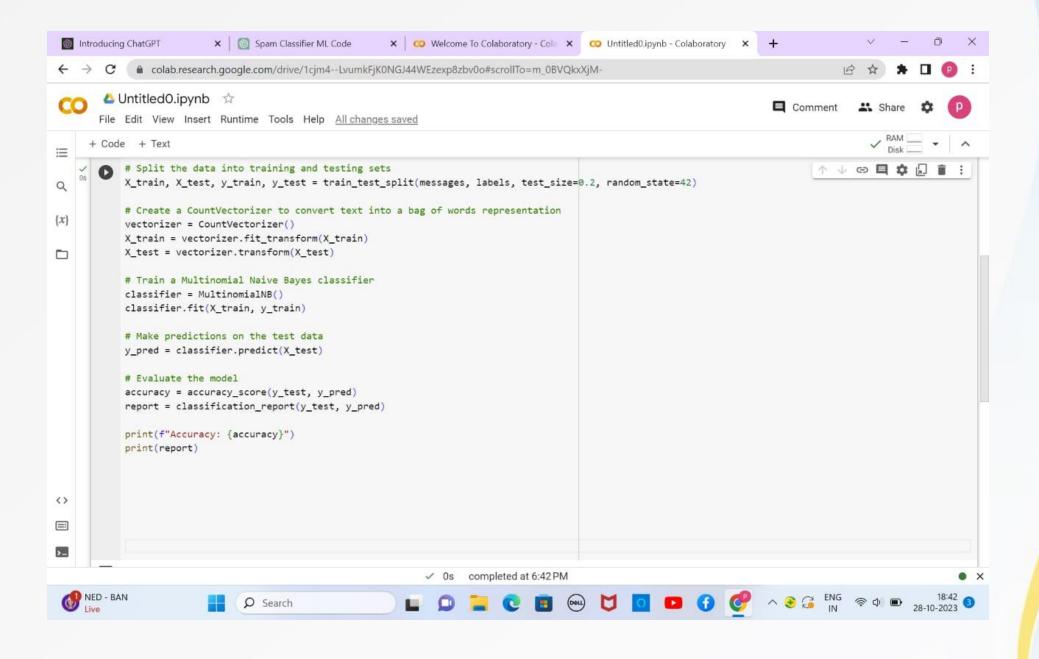
#### Data Prepocessing

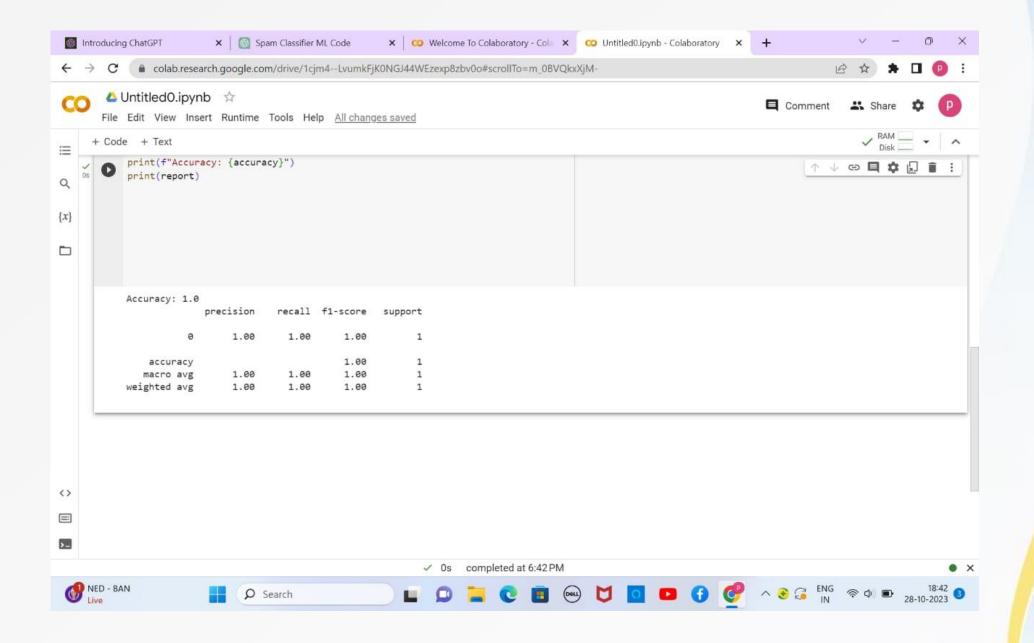
```
In [13]:
cv = CountVectorizer()
X = cv.fit_transform(df['text']).toar
ray()
y = df['spam']
In [14]:
X_{train}, X_{test}, y_{train}, y_{test} = tr
ain_test_split(X, y,stratify=y, test_
size=0.2, random_state=0)
#to resolve any class imbalance - use
stratify parameter.
print("Number transactions X_train da
taset: ", X_train.shape)
print("Number transactions y_train da
```

```
taset: ", y_train.shape)
print("Number transactions X_test dat
aset: ", X_test.shape)
print("Number transactions y_test dat
aset: ", y_test.shape)
Number transactions X_train dataset:
(4556, 33681)
Number transactions y_train dataset:
(4556,)
Number transactions X_test dataset:
(1139, 33681)
Number transactions y_test dataset:
(1139,)
```

## SIMPLE CODING USING MACHINE LEARNING







## **CONCLUSION**

By accurately identifying and filtering spam, individuals and organizations can focus on important emails and mitigate potential risks associated with malicious content. In conclusion, email spam detection using machine learning offers a promising solution to the pervasive problem of unwanted and harmful emails.

