Machine Learning Lab program 5



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AIM: Use any machine learning method to classify the email dataset

GitHub Link: - https://github.com/pranshuag9/machine-learning-lab/blob/main/lab6/Classify%20email%20dataset.ipynb

Program Snippets: -

1. Importing Modules

```
Problem statement :-
Use any machine learning method to classify the email dataset

importing libraries

In [152]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
```

2. Loading dataset: -

	Loading Dataset				
<pre>In [4]: df = pd.read_csv("spam_ham_dataset.csv") df.head()</pre>					
		Unnamed: 0	label	text	label_num
	0	605	ham	Subject: enron methanol ; meter # : 988291\r\n	0
	1	2349	ham	Subject: hpl nom for january 9 , 2001\r\n(see	0
	2	3624	ham	Subject: neon retreat\r\nho ho ho , we ' re ar	0
	3	4685	spam	Subject: photoshop , windows , office . cheap	1
	4	2030	ham	Subject: re : indian springs\r\nthis deal is t	0

3. Data Cleaning/ Preprocessing

```
checking count of values which are null

In [155]: df.isnull().value_counts()

Unnamed: 0 label text label_num
False False False False 5171
dtype: int64

dropping unnamed column

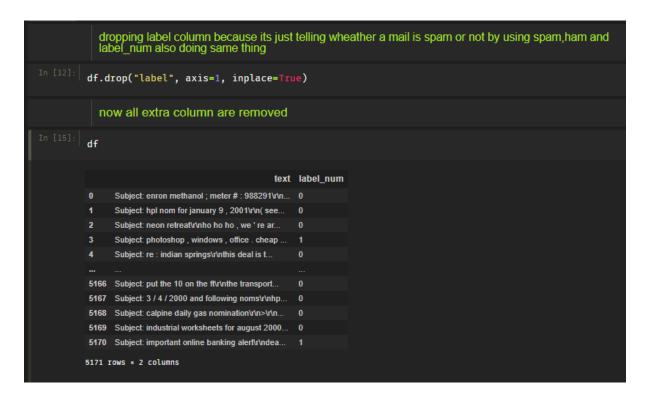
In [8]: df.drop("Unnamed: 0", axis=1,inplace=True)

checking count of each spam and ham(not spam value)

In [11]:

df.value_counts("label_num")

label_num
0 3672
1 1499
dtype: int64
```



```
appending Length to df which helps to get len of each mail
len(df.text[0])
df['Length']=df['text'].apply(len)
df
                                             text label_num Length
       Subject: enron methanol; meter #: 988291\r\n... 0
0
                                                                327
      Subject: hpl nom for january 9, 2001\r\n( see...
                                                                97
      Subject: neon retreat\r\nho ho ho , we ' re ar...
2
                                                                2524
3
      Subject: photoshop, windows, office. cheap... 1
                                                                414
      Subject: re : indian springs\r\nthis deal is t...
4
                                                                336
5166 Subject: put the 10 on the ft\r\nthe transport...
                                                                534
5167 Subject: 3 / 4 / 2000 and following noms\r\nhp... 0
                                                                1114
5168 Subject: calpine daily gas nomination\r\n>\r\n...
                                                                402
5169 Subject: industrial worksheets for august 2000... 0
                                                                573
5170 Subject: important online banking alert\r\ndea...
                                                                1114
5171 rows × 3 columns
```

```
using regular expression to clean mail and appending it to df
 def mailprocess(text):
              text= text.replace('\r','')
               text = text.replace('\n', ' ')
              text = text.replace('#', '')
              text = text.replace("Subject:","")
              text = text.replace("we 're", "we are")
              text = text.replace("ect","etc")
               text = text.replace("you 're","you are")
                return text
df['clean_text'] = df['text'].map(mailprocess)
                                                                                                                              text label_num Length
                                                                                                                                                                                                                                                                                                clean text
                Subject: enron methanol ; meter # : 988291\r\n... 0 327 enron methanol ; meter : 988291 this is a f...
            5166 Subject: put the 10 on the ff\(\text{tr\nthe transport...}\) 0 534 put the 10 on the ft the transport volumes d...

5167 Subject: 3 / 4 / 2000 and following noms\(\text{V\nhp...}\) 0 1114 3 / 4 / 2000 and following noms \(\text{hpl tr\nthe transport}\) 11568 Subject: calpine daily gas nomination\(\text{V\n...}\) 0 402 calpine daily gas nomination \(\text{> julle , a...}\)
5168 Subject: calpine daily gas nomination\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm\u00cm
                                                                                                                                                                                                           industrial worksheets for august 2000 activit...
                                                                                                                                                                                                    important online banking alert dear valued c..
5171 rows × 4 columns
```

4. Data Visualization

```
Data Visualization(most common words in clean_text column)
from wordcloud import WordCloud, STOPWORDS
comment_word = ''
stopwords= set(STOPWORDS)
for val in df.clean_text:
   tokens= val.strip().split()
    for i in range(len(tokens)):
       tokens[i] = tokens[i].lower()
   comment_word += " ".join(tokens)+" "
wc = WordCloud(width=2000, height=1000, background_color='white', stopwords=stopwords, min_font_si
plt.figure(figsize=(8,8), facecolor=None)
plt.imshow(wc)
plt.axis("off")
plt.tight_layout(pad=0)
plt.show()
                       onwenror
```

5. Converting data into X and Y

6. Dividing data into train/test splits

```
Splitting x,y into train and test split

In [176]: | from sklearn.model_selection import train_test_split

In [90]: | x_train, x_test, y_train, y_test=train_test_split(x,y,test_size=0.2, random_state=0)
```

7. Using Naïve Bayes classifying X_test

```
from sklearn.metrics import classification_report
from sklearn.metrics import confusion_matrix
from sklearn.metrics import accuracy_score
from sklearn.naive_bayes import GaussianNB
  using naive bayes classifier
gNB = GaussianNB()
y_pred=gNB.fit(x_train.toarray(), y_train).predict(x_test.toarray())
print(classification_report(y_test, y_pred))
            precision recall f1-score support
               0.96 0.99
0.97 0.90
                             0.97
               0.97
                       0.90
                               0.94
                               0.96
                                       1035
   accuracy
              0.96 0.95
                              0.95
   macro avg
              0.96
weighted avg
                      0.96
                               0.96
                                       1035
```

8. Accuracy Score:-

```
Accuracy Score

In [96]: accuracy_score(y_test, y_pred)

0.9632850241545894
```

9. Confusion matrix:-

