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Batch : FEBRUARY P-2

Task 4 - EMAIL SPAM DETECTION WITH MACHINE LEARNING

Problem Statement:

- Use Python to develop Email Spam Detector.
- Use Machine Learning to train Spam Detector to recognize and classify Emails into Spam and Non- Spam mails.

```
In [1]: #importing necessary libraries  
import numpy as np  
import pandas as pd  
import warnings  
warnings.filterwarnings('ignore')
```

```
In [2]: #importing Data  
data_frame = pd.read_csv('C:/Users/sinun/OneDrive/Documents/oasis infobyte/email_spam/spam.csv')
```

Performing descriptive analysis. Understand the variables and their corresponding values.

```
In [3]: # Understanding the dimensions of data  
data_frame.shape
```

```
Out[3]: (5572, 5)
```

```
In [4]: # Understanding the Data Variables  
data_frame.info()
```

```
<class 'pandas.core.frame.DataFrame'>  
RangeIndex: 5572 entries, 0 to 5571
```

```
Data columns (total 5 columns):
#      Column      Non-Null Count  Dtype
---  -
0     v1          5572 non-null    object
1     v2          5572 non-null    object
2     Unnamed: 2    50 non-null     object
3     Unnamed: 3    12 non-null     object
4     Unnamed: 4     6 non-null      object
dtypes: object(5)
memory usage: 217.8+ KB
```

```
In [5]: data_frame.columns
```

```
Out[5]: Index(['v1', 'v2', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], dtype='object')
```

* Column v2 contains texts of emails and v1 contains whether those emails are spam or not

```
In [6]: # Show the top 10 Rows of data
data_frame.head(10)
```

```
Out[6]:
```

	v1	v2	Unnamed: 2	Unnamed: 3	Unnamed: 4
0	ham	Go until jurong point, crazy.. Available only ...	NaN	NaN	NaN
1	ham	Ok lar... Joking wif u oni...	NaN	NaN	NaN
2	spam	Free entry in 2 a wkly comp to win FA Cup fina...	NaN	NaN	NaN
3	ham	U dun say so early hor... U c already then say...	NaN	NaN	NaN
4	ham	Nah I don't think he goes to usf, he lives aro...	NaN	NaN	NaN
5	spam	FreeMsg Hey there darling it's been 3 week's n...	NaN	NaN	NaN
6	ham	Even my brother is not like to speak with me. ...	NaN	NaN	NaN
7	ham	As per your request 'Melle Melle (Oru Minnamin...	NaN	NaN	NaN
8	spam	WINNER!! As a valued network customer you have...	NaN	NaN	NaN
9	spam	Had your mobile 11 months or more? U R entitle...	NaN	NaN	NaN

```
In [7]: # Checking for any Duplicated Entries
data_frame.duplicated().sum()
```

Out[7]: 403

```
In [8]: data_frame.drop_duplicates(keep = 'first', inplace = True)
data_frame.shape
```

Out[8]: (5169, 5)

```
In [9]: #Detecting missing values
data_frame.isna().sum()
```

```
Out[9]: v1          0
v2          0
Unnamed: 2    5126
Unnamed: 3    5159
Unnamed: 4    5164
dtype: int64
```

```
In [10]: #Dropping unwanted Columns from data
data_frame.drop(columns=['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], inplace=True )
data_frame.head()
```

```
Out[10]:
```

	v1	v2
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...

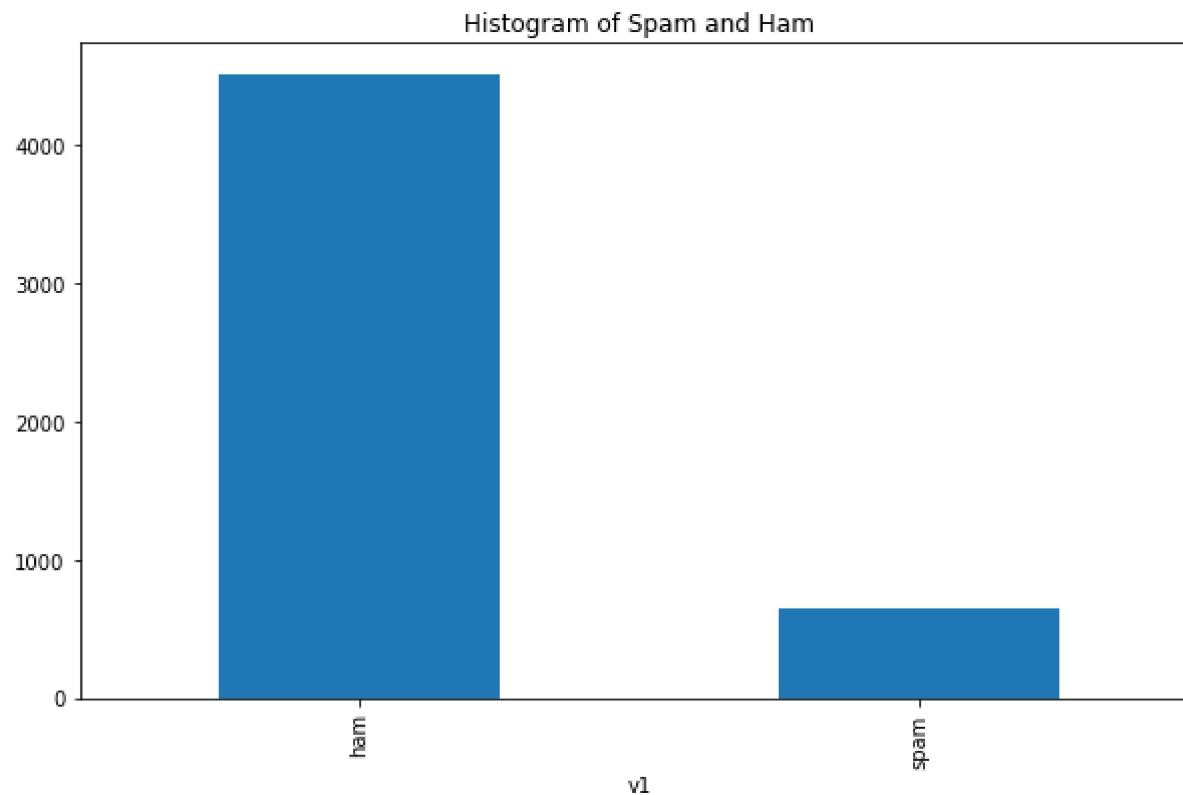
Data Visualization

```
In [11]: # Finding the number of spam and non-spam mails in dataset  
data_frame['v1'].value_counts()
```

```
Out[11]: ham      4516  
spam       653  
Name: v1, dtype: int64
```

```
In [12]: data_frame['v1'].value_counts().plot(kind='bar',figsize=(10,6),title="Histogram of Spam and Ham",xlabel="v1")
```

```
Out[12]: <AxesSubplot:title={'center':'Histogram of Spam and Ham'}, xlabel='v1'>
```



```
In [13]: # Printing examples of ham messages  
print("Some examples of ham messages:")  
print(data_frame[data_frame['v1'] == 'ham']['v2'].head())
```

```

Some examples of ham messages:
0    Go until jurong point, crazy.. Available only ...
1                Ok lar... Joking wif u oni...
3    U dun say so early hor... U c already then say...
4    Nah I don't think he goes to usf, he lives aro...
6    Even my brother is not like to speak with me. ...
Name: v2, dtype: object

```

```

In [14]: # Printing examples of spam messages
print(" Some examples of spam messages:")
print(data_frame[data_frame['v1'] == 'spam']['v2'].head())

```

```

Some examples of spam messages:
2    Free entry in 2 a wkly comp to win FA Cup fina...
5    FreeMsg Hey there darling it's been 3 week's n...
8    WINNER!! As a valued network customer you have...
9    Had your mobile 11 months or more? U R entitle...
11   SIX chances to win CASH! From 100 to 20,000 po...
Name: v2, dtype: object

```

```

In [15]: # Labelling Spam mails as 1 and Non-Spam Mails as 0
data_frame['v1'] = np.where(data_frame['v1'] == 'spam',1,0)
data_frame

```

```

Out[15]:

```

	v1	v2
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 b going to esplanade fr home?
5569	0	Pity, * was in mood for that. So...any other s...
5570	0	The guy did some bitching but I acted like i'd...

v1	v2
5571 0	Rofl. Its true to its name

5169 rows × 2 columns

Building the Model

```
In [16]: from sklearn.model_selection import train_test_split
```

```
In [17]: #First step in building the model is to identify the Feature(Input) variables and Target (Output) variable
#Features are the emails
#Target represents the labels for spam and ham emails
features = data_frame['v2']
target = data_frame['v1']
```

* Splitting data for training and testing the model

```
In [18]: # Splitting data for training the model and testing the model
# train size taken as 0.8
X_train, X_test, y_train, y_test = train_test_split(features, target, train_size = .8)
# Dimensions of Train and Test Data sets
print('Train set of features: ', X_train.shape)
print('Test set of features: ', X_test.shape)
print('Target for train: ', y_train.shape)
print('Target for test: ', y_test.shape)
```

```
Train set of features: (4135,)
Test set of features: (1034,)
Target for train: (4135,)
Target for test: (1034,)
```

Learn the model on train data

```
In [19]: from sklearn.feature_extraction.text import CountVectorizer
```

```
In [20]: #CountVectorizer transform text into a vector on the basis of the frequency/count of each word that occurs in entire text.  
#Covert each word in Train dataset into vectors for using them in further text analysis.  
cv=CountVectorizer()  
X_train_vector=cv.fit_transform(X_train.values)  
X_train_vector.toarray()
```

```
Out[20]: array([[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, 0, ..., 0, 0, 0]], dtype=int64)
```

```
In [21]: # Here Multinomial Naive Bayes model which is a supervised learning classification  
#Used for the analysis of the categorical text data.  
from sklearn.naive_bayes import MultinomialNB  
my_model=MultinomialNB()
```

```
In [22]: # Fitting the model in train data set ie the MultinomialNB Model should Learn from the Train Data  
my_model.fit(X_train_vector, y_train)
```

```
Out[22]: MultinomialNB()
```

```
In [23]: #Coverting each word in Test dataset into vectors for using them in further text analysis.  
X_test_vector=cv.transform(X_test.values)  
X_test_vector.toarray()
```

```
Out[23]: array([[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, 0, ..., 0, 0, 0]], dtype=int64)
```

Prediction

```
In [24]: # Predicting whether the mails in Test Dataset are spam or non-spam using our model from Feature Test values  
y_pred = my_model.predict(X_test_vector)  
y_pred
```

```
Out[24]: array([0, 0, 0, ..., 0, 0, 0])
```

Testing the model

```
In [25]: from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score
```

```
In [26]: # find Accuracy Score = (TP+TN)/(TP+TN+FP+FN)  
accuracy_score(y_test, y_pred)
```

```
Out[26]: 0.9903288201160542
```

```
In [27]: # find Precision Score= TP/(TP+FP)  
precision_score(y_test, y_pred)
```

```
Out[27]: 0.9754098360655737
```

```
In [28]: # Recall = TP/(TP+FN)  
# Recall gives the percentage of positives well predicted by our model.  
recall_score(y_test, y_pred)
```

```
Out[28]: 0.9444444444444444
```

```
In [29]: # f1_score= (Precision * Recall) / (Precision + Recall)  
f1_score(y_test, y_pred)
```

```
Out[29]: 0.9596774193548386
```