• Batch: MARCH P-1 Role:Data Science Intern • Company:OASIS INFOBYTE • TASK4: EMAIL SPAM DETECTION WITH MACHINE LEARNING • Objective: -Utilize Machine Learning techniques so that it can accurately categorize emails. -Develop an Email Spam Detection System. In [2]: #Importing required libraries import pandas as pd import numpy as np #Importing warning modules import warnings warnings.simplefilter('ignore') #Importing necessary function to split data into training and testing sets from sklearn.model_selection import train_test_split #Bringing in visualisation libraries for plotting graphs import matplotlib.pyplot as plt import seaborn as sns In [2]: #Reading the CSV file into a Dataframe import pandas as pd # Load the dataset data__3= pd.read_csv(r"C:/Users/Abhishek/Desktop/OASIS INFOBYTE/spam.csv.csv",encoding='ISO-8859-1') data__3.head(4) v2 Unnamed: 2 Unnamed: 3 Unnamed: 4 Out[2]: v1 Go until jurong point, crazy.. Available only ... NaN NaN NaN **0** ham **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 Conducting a descriptive analysis to understand the variables and their associated values In [66]: #Examining the dataset's dimensions data__3.shape Out[66]: (5572, 5) In [11]: #Display datset information data__3.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 5572 entries, 0 to 5571 Data columns (total 5 columns): # Column Non-Null Count Dtype --- -----5572 non-null object 0 v1 5572 non-null object 1 v2 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 [11]: #Display column names print(data__3.columns) Index(['Category', 'Message', 'Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], dtype='object') In [13]: #verifying for duplicate entries in dataset data__3.duplicated().sum() Out[13]: np.int64(403) In [14]: #Checking for missing values in the dataset data__3.isna().sum() Out[14]: **v1** 0 Unnamed: 2 5522 Unnamed: 3 5560 5566 Unnamed: 4 dtype: int64 In [15]: #Dropping dulicate rows data__3.drop_duplicates(keep='first',inplace=True) data__3.shape Out[15]: (5169, 5) In [16]: #Dropping irrelevant columns from the Dtaframe using their names data__3.drop(columns=['Unnamed: 2','Unnamed: 3','Unnamed: 4'],inplace=True) #Displaying the first few record after cleaning up the columns data__3.head() v1 v2 Go until jurong point, crazy.. Available only ... **0** ham Ok lar... Joking wif u oni... **1** ham 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... Nah I don't think he goes to usf, he lives aro... In [17]: #Analyzing the frequency of spam and non spam mails spam=data__3['v1'].value_counts() print(spam) v1 4516 ham spam Name: count, dtype: int64 In [18]: #Plotting the frequency of spam and ham emails in bar chart data__3['v1'].value_counts().plot(kind='bar',figsize=(12,7),title="Spam vs Ham Email Distribution",xlabel="Category",ylabel="Frequency") print(data__3['v1'].head()) ham ham spam 3 ham ham Name: v1, dtype: object Spam vs Ham Email Distribution 4000 3000 Frequency 2000 1000 Category In [21]: #converting spam labels: Spam=1,Non-spam=0 data__3['v1'] = (data__3['v1'] == 'spam').astype(int) data__3 Out[21]: v1 v2 **0** 0 Go until jurong point, crazy.. Available only ... **1** 0 Ok lar... Joking wif u oni... **2** 0 Free entry in 2 a wkly comp to win FA Cup fina... U dun say so early hor... U c already then say... Nah I don't think he goes to usf, he lives aro... ... **5567** 0 This is the 2nd time we have tried 2 contact u... Will *i*_ b going to esplanade fr home? **5568** 0 Pity, * was in mood for that. So...any other s... The guy did some bitching but I acted like i'd... **5570** 0 **5571** 0 Rofl. Its true to its name 5169 rows \times 2 columns In [55]: #Displaying sample spam messages data__3['v1']=data__3['v1'].map({'spam':1,'ham':0}) print("Examples of detected spam messages:") print(data__3[data__3['v1'] == 1]['v2'].head()) Examples of detected spam messages: v1 v2 Unnamed: 2 \ 1 Free entry in 2 a wkly comp to win FA Cup fina... 1 FreeMsg Hey there darling it's been 3 week's n... NaN 1 WINNER!! As a valued network customer you have... NaN 1 Had your mobile 11 months or more? U R entitle... NaN 1 SIX chances to win CASH! From 100 to 20,000 po... 5537 1 Want explicit SEX in 30 secs? Ring 02073162414... NaN 1 ASKED 3MOBILE IF 0870 CHATLINES INCLU IN FREE ... 5547 1 Had your contract mobile 11 Mnths? Latest Moto... NaN 5566 1 REMINDER FROM 02: To get 2.50 pounds free call... NaN 5567 1 This is the 2nd time we have tried 2 contact u... NaN Unnamed: 3 Unnamed: 4 NaN NaN 5 NaN NaN 8 NaN NaN 9 NaN NaN 11 NaN NaN 5537 NaN NaN 5540 NaN NaN 5547 NaN 5566 NaN NaN 5567 NaN NaN [747 rows x 5 columns] In [56]: #Displaying sample non-spam (ham) messages print("Here are some examples of ham messages:") print(data__3[data__3['v1'] == 0]['v2'].head()) Here are some examples of ham messages: Go until jurong point, crazy.. Available only ... Ok lar... Joking wif u oni... 3 U dun say so early hor... U c already then say... Nah I don't think he goes to usf, he lives aro... 4 6 Even my brother is not like to speak with me. ... Name: v2, dtype: object In [4]: data__3.describe() Out[4]: v2 Unnamed: 2 Unnamed: 3 Unnamed: 4 v1 **count** 5572 5572 50 12 6 43 5169 5 unique top ham Sorry, I'll call later bt not his girlfrnd... G o o d n i g h t . . . @" MK17 92H. 450Ppw 16" GNT:-)" **freq** 4825 30 2 In [10]: data__3.tail(6)#Provides Last 6 entries Out[10]: v2 Unnamed: 2 Unnamed: 3 Unnamed: 4 v1 **5566** spam REMINDER FROM O2: To get 2.50 pounds free call... NaN NaN NaN **5567** spam This is the 2nd time we have tried 2 contact u... NaN NaN NaN Will *i*_ b going to esplanade fr home? **5568** ham NaN NaN NaN **5569** ham Pity, * was in mood for that. So...any other s... NaN NaN NaN The guy did some bitching but I acted like i'd... **5570** ham NaN NaN NaN **5571** ham NaN NaN Rofl. Its true to its name NaN In [12]: #Returns a boolean series data__3.duplicated() Out[12]: 0 False False False False False 5567 False 5569 False 5570 False 5571 False Length: 5572, dtype: bool In [7]: data__3.duplicated().sum() Out[7]: np.int64(403) In [9]: #Returns the total number of elements in dataset data__3.size Out[9]: **27860** In [89]: data__3.columns=data__3.columns.str.replace("v1","Category").str.replace("v2","Message") In [90]: data__3.head() Message Unnamed: 2 Unnamed: 3 Unnamed: 4 Out[90]: Category 0 Go until jurong point, crazy.. Available only ... NaN NaN NaN ham 1 Ok lar... Joking wif u oni... NaN NaN NaN ham 2 spam Free entry in 2 a wkly comp to win FA Cup fina... NaN NaN NaN 3 U dun say so early hor... U c already then say... NaN NaN NaN Nah I don't think he goes to usf, he lives aro... NaN NaN NaN In [91]: counts= data__3["Category"].value_counts() plt.pie(counts, labels=["Safe" ,"Not Safe"],autopct="%0.4f", colors=["blue","red"]) plt.title("Distribution of message categories") plt.show() Distribution of message categories Safe 13.4063 Not Safe Developing the Model In [7]: from sklearn.model_selection import train_test_split In [9]: #Here the predictors represent mails and response denotes whether email is spam or ham predictors=data__3["v2"] response=data__3["v1"] In [11]: #Dividing the dataset into training and testing sets for model evaluation #75% of the data is allocated for training T_train,T_test,z_train,z_test=train_test_split(predictors,response,train_size=0.75) #Displaying the dimensions of the training and testing datasets print('Training set(input):', T_train.shape) print('Testing set (input):',T_test.shape) print('Training set(output):',z_train.shape) print('Testing set (output):',z_test.shape) Training set(input): (4179,) Testing set (input): (1393,) Training set(output): (4179,) Testing set (output): (1393,) Training the model on training dataset In [13]: from sklearn.feature_extraction.text import CountVectorizer from sklearn.naive_bayes import MultinomialNB In [15]: #Implementing the multinomial Naive Bayes model, a supervised learning algorithm #This model is widely used for classifying categorical text data text_classifier =MultinomialNB() In [16]: #Intialising countVectorizer cv=CountVectorizer() #Countvectorizer converts text data into numerical vector #It represents text based on the frequency of words appearing in dataset #Transforming the training dataset into a numerical vector format T_train_vector =cv.fit_transform(T_train.values) #Converting sparse matrix to an array for further analysis T_train_vector.toarray() Out[16]: 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]]) In [17]: #Training the Multinomial Naive Bayes model using the transformed training dataset #The model learns patterns from the training data to classify text effectively text_classifier.fit(T_train_vector,z_train) ▼ MultinomialNB MultinomialNB() In [18]: #Converting words in the test dataset into numerical vectors #Using the same CountVectorizer instance to ensure consistency T_test_vector=cv.transform(T_test.values) T_test_vector.toarray() Out[18]: array([[0, 0, 0, ..., 0, 0, 0], $[0, 0, 0, \ldots, 0, 0, 0],$ [0, 0, 0, ..., 0, 0, 0], $[0, 0, 0, \ldots, 0, 0, 0],$ $[0, 0, 0, \ldots, 0, 0, 0],$ [0, 0, 0, ..., 0, 0, 0]]) Estimation In [19]: #Using the trained model to predict whether emails in the test dataset are spam or not #Predictions are made on transformed test values z_pred=text_classifier.predict(T_test_vector) z_pred Out[19]: array(['ham', 'spam', 'ham', ..., 'ham', 'ham', 'ham'], dtype='<U4') Evaluating the model In [30]: from sklearn.metrics import confusion_matrix,classification_report In [21]: #Generating the confusion matrix conf_matrix=confusion_matrix(z_test,z_pred) print("Confusion Matrix:\n",conf_matrix) Confusion Matrix: [[1203 5] [13 172]] In [22]: #Generating a detailed classification report class_report=classification_report(z_test,z_pred) print("Classification Report:\n",class_report) Classification Report: recall f1-score support precision ham 0.99 1.00 0.99 1208 0.97 0.93 0.95 185 spam 0.99 1393 accuracy macro avg 0.98 0.96 0.97 1393 0.99 0.99 0.99 1393 weighted avg In [28]: #Compute balanced accuracy

from sklearn.metrics import balanced_accuracy_score
balanced_accuracy=balanced_accuracy_score(z_test,z_pred)
print("Balanced Accuracy score:",balanced_accuracy)

precision=precision_score(z_test,z_pred,pos_label="spam")

Balanced Accuracy score: 0.9627953284410238

In [29]: from sklearn.metrics import precision_score

print("Precision:",precision)

Precision: 0.9717514124293786

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