```
In [ ]:
         #Tushar Holkar
          # Roll no: A28
 In [ ]:
          #Problem Statement
          #Emails Spam detection using Binary Classification
         import pandas as pd
 In [1]:
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
          import plotly.express as px
          import warnings
          from sklearn.model_selection import train_test_split
          from sklearn.neighbors import KNeighborsClassifier
          from sklearn.svm import SVC
          from sklearn.utils import resample
          from sklearn import metrics
          from tqdm.notebook import tqdm
          %matplotlib inline
          warnings.filterwarnings("ignore")
 In [5]: df = pd.read_csv("C:/Users/KJCOEMR/Downloads/emails.csv")
         df.head()
 In [7]:
 Out[7]:
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                                    1
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                                                57
                                                            9
                                                                                      0
                                                                                           0
         5 rows × 3002 columns
         df.columns
In [27]:
Out[27]: Index(['the', 'to', 'ect', 'and', 'for', 'of', 'a', 'you', 'hou', 'in',
                  'connevey', 'jay', 'valued', 'lay', 'infrastructure', 'military',
                 'allowing', 'ff', 'dry', 'Prediction'],
                dtype='object', length=3001)
In [29]: df.info()
```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5172 entries, 0 to 5171
Columns: 3001 entries, the to Prediction

dtypes: int64(3001)
memory usage: 118.4 MB

In [33]: df.isnull().sum()

Out[33]: the 0 to 0 ect 0 and 0 for 0

military 0
allowing 0
ff 0
dry 0
Prediction 0

Length: 3001, dtype: int64

In [35]: df.head(10)

Out[35]: the to ect and for of a you hou in ... connevey

	the	to	ect	and	tor	of	а	you	hou	in	•••	connevey	jay	valued	lay	ini
0	0	0	1	0	0	0	2	0	0	0		0	0	0	0	
1	8	13	24	6	6	2	102	1	27	18		0	0	0	0	
2	0	0	1	0	0	0	8	0	0	4		0	0	0	0	
3	0	5	22	0	5	1	51	2	10	1		0	0	0	0	
4	7	6	17	1	5	2	57	0	9	3		0	0	0	0	
5	4	5	1	4	2	3	45	1	0	16		0	0	0	0	
6	5	3	1	3	2	1	37	0	0	9		0	0	0	0	
7	0	2	2	3	1	2	21	6	0	2		0	0	0	0	
8	2	2	3	0	0	1	18	0	0	3		0	0	0	0	
9	4	4	35	0	1	0	49	1	16	9		0	0	0	0	

10 rows × 3001 columns

←

In [39]: df.tail(10)

Out[39]:		the	to	ect	and	for	of	а	you	hou	in	•••	connevey	jay	valued	lay
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	5163	0	0	1	0	0	0	1	0	0	0		0	0	0	0
	5164	21	18	3	1	6	4	106	1	2	18		0	0	0	0
	5165	1	0	1	0	3	1	12	1	0	2		0	0	0	1
	5166	1	0	1	1	0	0	4	0	0	0		0	0	0	0
	5167	2	2	2	3	0	0	32	0	0	5		0	0	0	0
	5168	35	27	11	2	6	5	151	4	3	23		0	0	0	0
	5169	0	0	1	1	0	0	11	0	0	1		0	0	0	0
	5170	2	7	1	0	2	1	28	2	0	8		0	0	0	0
	5171	22	24	5	1	6	5	148	8	2	23		0	0	0	0
	10 rows × 3001 columns															
	4															•
In [75]:	<pre>df.drop(columns=["Email No."], inplace=True, errors='ignore')</pre>															
In [71]:	df.ta:	il()														
Out[71]:		the	to	ect	and	for	of	а	you	hou	in	•••	connevey	jay	valued	lay
	5167	2	2	2	3	0	0	32	0	0	5		0	0	0	0
	5168	35	27	11	2	6	5	151	4	3	23		0	0	0	0
	5169	0	0	1	1	0	0	11	0	0	1		0	0	0	0
	5170	2	7	1	0	2	1	28	2	0	8		0	0	0	0
	5171	22	24	5	1	6	5	148	8	2	23		0	0	0	0
	5 rows × 3001 columns															
	4														•	
In [84]:	df.is	null().an	y().	value_	_coun	its())								
Out[84]:	False Name:		8001 nt, c	ltype	: int	64										
In [86]:	<pre>X=df.iloc[:, :df.shape[1]-1] y=df.iloc[:, -1] X.shape,y.shape</pre>															
Out[86]:	((517	2, 30	00),	(51	72,))											
In [88]:	from	sklea	rn.m	odel_	_seled	ction	imp	ort	train ₋	_test_	_spl:	it				
In [90]:	x_tra	in,x_	test	,y_tı	rain,y	_tes	t=tr	rain_	test_	split((X, y	, tes	t_size=0.1	L5)		
In [94]:	from	sklea	rn.l	inea	r_mode	el i m	port	t Log	istic	Regres	ssion	า				

```
from sklearn.svm import SVC, LinearSVC
         from sklearn.neural_network import MLPClassifier
In [98]: models = {
             "Logistic Regression": LogisticRegression(solver='lbfgs', max_iter=2000),
             "Linear SVM": LinearSVC(max_iter=3000),
             "Polynomial SVM": SVC(kernel='poly', degree=2),
             "RBF SVM": SVC(kernel='rbf'),
             "Sigmoid SVM": SVC(kernel='sigmoid'),
             "Multi-layer Perceptron Classification": MLPClassifier(hidden_layer_sizes=[2
In [100... from sklearn.metrics import accuracy_score
In [105... for model_name, model in models.items():
             y_pred = model.fit(x_train, y_train).predict(x_test)
             print(f"Accuracy for {model_name} model is: {accuracy_score(y_test, y_pred)}
        Accuracy for Logistic Regression model is: 0.9716494845360825
        Accuracy for Linear SVM model is: 0.9484536082474226
        Accuracy for Polynomial SVM model is: 0.7435567010309279
        Accuracy for RBF SVM model is: 0.8118556701030928
        Accuracy for Sigmoid SVM model is: 0.5927835051546392
        Accuracy for Multi-layer Perceptron Classification model is: 0.979381443298969
 In [ ]:
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