

Libraries and classes

```
In [1]: import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.svm import SVC
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import ConfusionMatrixDisplay, accuracy_score
from sklearn.metrics import classification_report
import matplotlib.pyplot as plt
import seaborn as sns
```

```
In [2]: df = pd.read_csv("emails.csv")
```

```
In [3]: df.head()
```

Out[3]:

	Email No.	the	to	ect	and	for	of	a	you	hou	...	connevey	jay	valued	lay	infrastr
0	Email 1	0	0	1	0	0	0	2	0	0	...	0	0	0	0	
1	Email 2	8	13	24	6	6	2	102	1	27	...	0	0	0	0	
2	Email 3	0	0	1	0	0	0	8	0	0	...	0	0	0	0	
3	Email 4	0	5	22	0	5	1	51	2	10	...	0	0	0	0	
4	Email 5	7	6	17	1	5	2	57	0	9	...	0	0	0	0	

5 rows × 3002 columns



```
In [4]: df.isnull().sum()
```

```
Out[4]: Email No.      0
the      0
to      0
ect      0
and      0
..
military  0
allowing  0
ff      0
dry      0
Prediction  0
Length: 3002, dtype: int64
```

```
In [5]: X = df.iloc[:,1:3001]
X
```

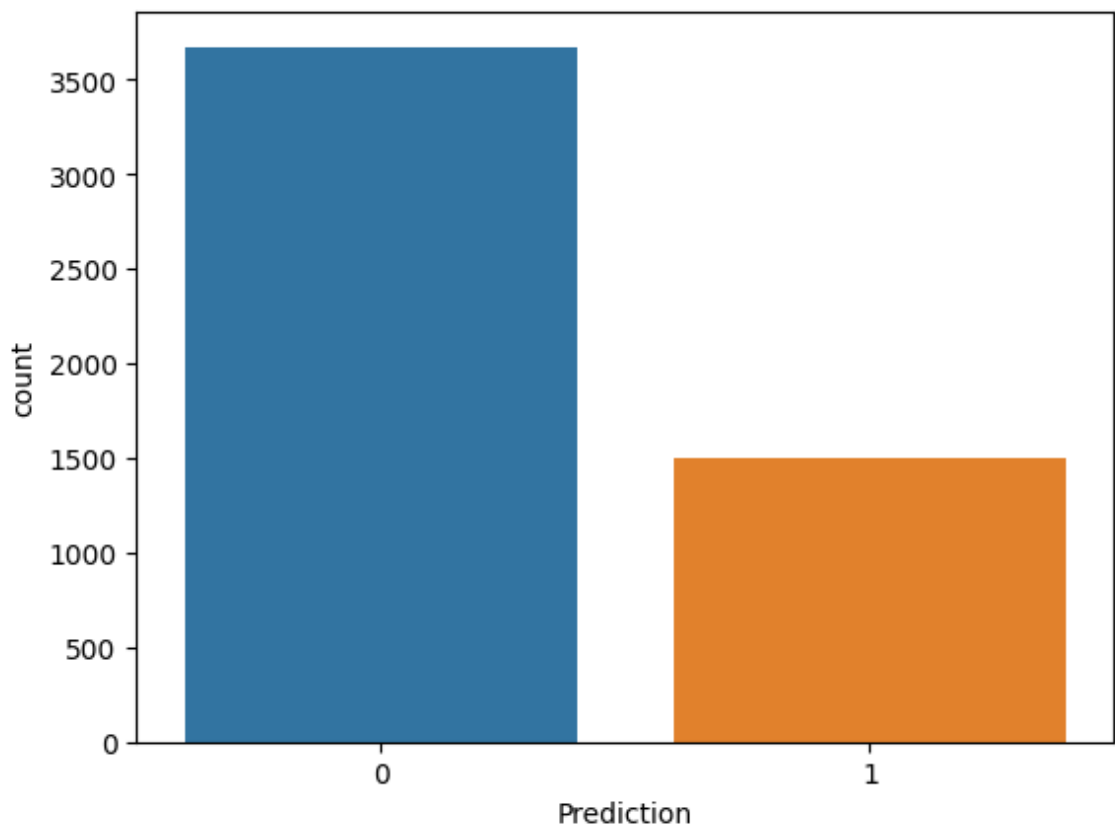
Out[5]:

	the	to	ect	and	for	of	a	you	hou	in	...	enhancements	connevey	jay	v
0	0	0	1	0	0	0	2	0	0	0	...	0	0	0	
1	8	13	24	6	6	2	102	1	27	18	...	0	0	0	
2	0	0	1	0	0	0	8	0	0	4	...	0	0	0	
3	0	5	22	0	5	1	51	2	10	1	...	0	0	0	
4	7	6	17	1	5	2	57	0	9	3	...	0	0	0	
...	
5167	2	2	2	3	0	0	32	0	0	5	...	0	0	0	
5168	35	27	11	2	6	5	151	4	3	23	...	0	0	0	
5169	0	0	1	1	0	0	11	0	0	1	...	0	0	0	
5170	2	7	1	0	2	1	28	2	0	8	...	0	0	0	
5171	22	21	5	1	2	5	142	2	2	22	...	0	0	0	

Count plot of y

```
In [6]: x=df.drop(["Email No.", "Prediction"], axis=1)
y=df["Prediction"]
print(y)
print(x.shape)
print(sns.countplot(x=y))
print(y.value_counts())
```

```
0      0
1      0
2      0
3      0
4      0
..
5167   0
5168   0
5169   1
5170   1
5171   0
Name: Prediction, Length: 5172, dtype: int64
(5172, 3000)
AxesSubplot(0.125,0.11;0.775x0.77)
0    3672
1    1500
Name: Prediction, dtype: int64
```



Scaling of values using min-max scaling technique

```
In [7]: from sklearn.preprocessing import MinMaxScaler
scaler=MinMaxScaler()
x_scaled=scaler.fit_transform(x)
x_scaled
```

```
Out[7]: array([[0.          , 0.          , 0.          , ..., 0.          , 0.          ,
                0.          ],
               [0.03809524, 0.09848485, 0.06705539, ..., 0.          , 0.00877193,
                0.          ],
               [0.          , 0.          , 0.          , ..., 0.          , 0.          ,
                0.          ],
               ...,
               [0.          , 0.          , 0.          , ..., 0.          , 0.          ,
                0.          ],
               [0.00952381, 0.0530303 , 0.          , ..., 0.          , 0.00877193,
                0.          ],
               [0.1047619 , 0.18181818, 0.01166181, ..., 0.          , 0.          ,
                0.          ]])
```

```
In [8]: Y = df.iloc[:, -1].values
Y
```

```
Out[8]: array([0, 0, 0, ..., 1, 1, 0], dtype=int64)
```

Splitting of dataset in train and test dataset

```
In [9]: x_train,x_test,y_train,y_test = train_test_split(x_scaled,y,random_state=0,
print(x_scaled.shape)
print(x_train.shape)
print(x_test.shape)
```

```
(5172, 3000)
(3879, 3000)
(1293, 3000)
```

Model training using KNN algo.

```
In [10]: knn = KNeighborsClassifier(n_neighbors=5)
knn.fit(x_train, y_train)
```

```
Out[10]: KNeighborsClassifier
KNeighborsClassifier()
```

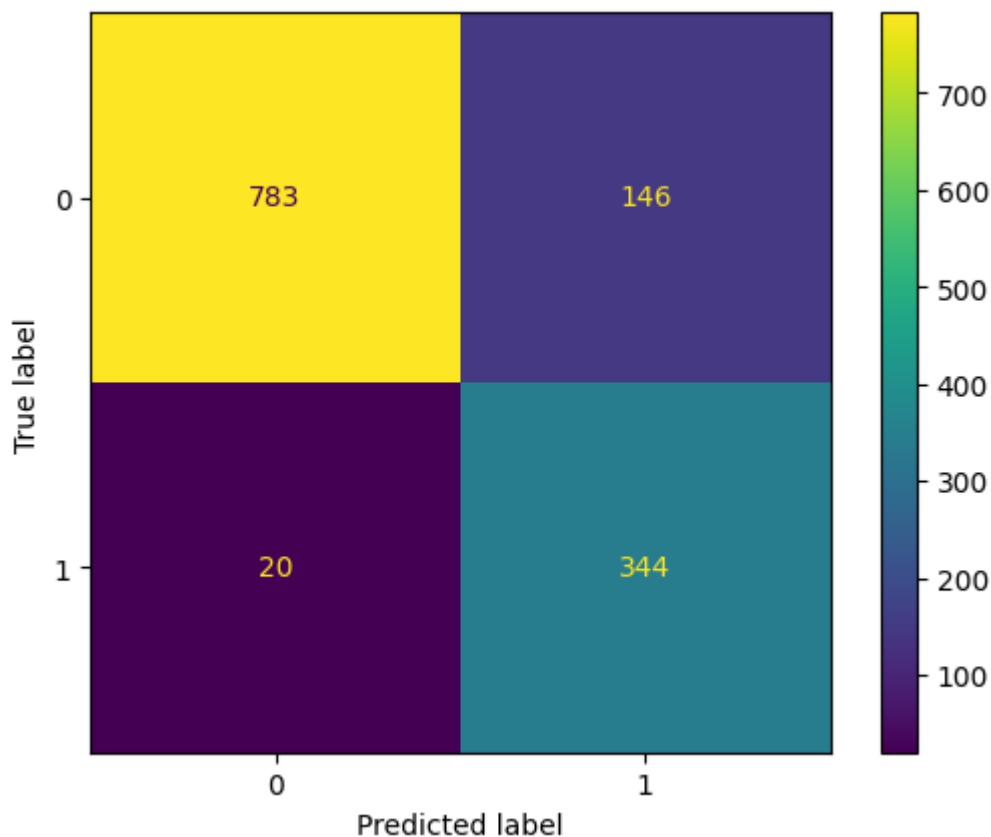
```
In [11]: y_pred=knn.predict(x_test)
print(y_pred)
```

```
[1 0 0 ... 1 0 1]
```

Performance measure of model

In [12]: `ConfusionMatrixDisplay.from_predictions(y_test,y_pred)`

Out[12]: `<sklearn.metrics._plot.confusion_matrix.ConfusionMatrixDisplay at 0x2889d9f9420>`



In [13]: `y_test.value_counts()`

Out[13]: `0 929
1 364
Name: Prediction, dtype: int64`

In [14]: `accuracy_score(y_test,y_pred)`

Out[14]: `0.871616395978345`

In [15]: `print(classification_report(y_test,y_pred))`

	precision	recall	f1-score	support
0	0.98	0.84	0.90	929
1	0.70	0.95	0.81	364
accuracy			0.87	1293
macro avg	0.84	0.89	0.85	1293
weighted avg	0.90	0.87	0.88	1293

Support vector machine model

```
In [17]: svc = SVC(C=1.0,kernel='rbf',gamma='auto')
# C here is the regularization parameter. Here, L2 penalty is used(default)
# As C increases, model overfits.
# Kernel here is the radial basis function kernel.
# gamma (only used for rbf kernel) : As gamma increases, model overfits.
svc.fit(x_train,y_train)
y_pred2 = svc.predict(x_test)
print("Accuracy Score for SVC : ", accuracy_score(y_pred,y_test))
```

Accuracy Score for SVC : 0.871616395978345

```
In [18]: svc = SVC(C=1.0,kernel='linear',gamma='auto')

svc.fit(x_train,y_train)
y_pred2 = svc.predict(x_test)
print("Accuracy Score for SVC : ", accuracy_score(y_pred,y_test))
```

Accuracy Score for SVC : 0.871616395978345

```
In [19]: svc = SVC(C=1.0,kernel='poly',gamma='auto')

svc.fit(x_train,y_train)
y_pred2 = svc.predict(x_test)
print("Accuracy Score for SVC : ", accuracy_score(y_pred,y_test))
```

Accuracy Score for SVC : 0.871616395978345