

```
In [1]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

```
In [3]: df=pd.read_csv("C:/Users/KJCOEMR/Desktop/emails.csv")
```

```
In [7]: df.head()
```

```
Out[7]:
```

	Email No.	the	to	ect	and	for	of	a	you	hou	...	connevey	jay	valued	lay	infr
--	-----------	-----	----	-----	-----	-----	----	---	-----	-----	-----	----------	-----	--------	-----	------

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 [9]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5172 entries, 0 to 5171
Columns: 3002 entries, Email No. to Prediction
dtypes: int64(3001), object(1)
memory usage: 118.5+ MB
```

```
In [11]: df.isnull().sum()
```

```
Out[11]: 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 [13]: X = df.iloc[:, 1:-1].values
        y = df.iloc[:, -1].values
```

```
In [15]: from sklearn.model_selection import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_st
```

```
In [17]: from sklearn.preprocessing import StandardScaler
        sc_X = StandardScaler()
        X_train = sc_X.fit_transform(X_train)
        X_test = sc_X.transform(X_test)
```

```
In [19]: from sklearn.neighbors import KNeighborsClassifier
        classifier = KNeighborsClassifier(n_neighbors=5)
        classifier.fit(X_train, y_train)
```

```
Out[19]: ▼ KNeighborsClassifier ⓘ ?
        KNeighborsClassifier()
```

```
In [21]: y_pred = classifier.predict(X_test)
```

```
In [23]: from sklearn.metrics import confusion_matrix, accuracy_score
        cm = confusion_matrix(y_test, y_pred)
```

```
In [25]: cm
```

```
Out[25]: array([[866, 248],
               [ 16, 422]], dtype=int64)
```

```
In [27]: from sklearn.metrics import classification_report
        cl_report=classification_report(y_test,y_pred)
        print(cl_report)
```

	precision	recall	f1-score	support
0	0.98	0.78	0.87	1114
1	0.63	0.96	0.76	438
accuracy			0.83	1552
macro avg	0.81	0.87	0.81	1552
weighted avg	0.88	0.83	0.84	1552

```
In [29]: print("Accuracy Score for KNN : ", accuracy_score(y_pred,y_test))
```

Accuracy Score for KNN : 0.8298969072164949

```
In [31]: from sklearn.svm import SVC
```

```
In [33]: svm_classifier = SVC(kernel='linear', random_state=101)
        svm_classifier.fit(X_train, y_train)
```

Out[33]:

```
SVC
SVC(kernel='linear', random_state=101)
```

```
In [34]: y_pred_svm = svm_classifier.predict(X_test)
```

```
In [37]: from sklearn.metrics import confusion_matrix, classification_report, accuracy_score
```

```
In [39]: cm_svm = confusion_matrix(y_test, y_pred_svm)
print("Confusion Matrix for SVM:")
print(cm_svm)
```

Confusion Matrix for SVM:

```
[[1069  45]
 [  42 396]]
```

```
In [41]: cl_report_svm = classification_report(y_test, y_pred_svm)
print("Classification Report for SVM:")
print(cl_report_svm)
```

Classification Report for SVM:

	precision	recall	f1-score	support
0	0.96	0.96	0.96	1114
1	0.90	0.90	0.90	438
accuracy			0.94	1552
macro avg	0.93	0.93	0.93	1552
weighted avg	0.94	0.94	0.94	1552

```
In [43]: accuracy_svm = accuracy_score(y_test, y_pred_svm)
print("Accuracy Score for SVM:", accuracy_svm)
```

Accuracy Score for SVM: 0.9439432989690721

```
In [45]: print("KNN Accuracy:", accuracy_score(y_test, y_pred))
print("SVM Accuracy:", accuracy_svm)
```

KNN Accuracy: 0.8298969072164949

SVM Accuracy: 0.9439432989690721

```
In [47]: if accuracy_svm > accuracy_score(y_test, y_pred):
print("SVM performed better on this dataset.")
else:
print("KNN performed better on this dataset.")
```

SVM performed better on this dataset.

In []: