```
In [1]: import pandas as pd
        import numpy as np
        import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn.model_selection import train_test_split
         from sklearn.metrics import confusion_matrix, accuracy_score
         from sklearn.neighbors import KNeighborsClassifier
         from sklearn.preprocessing import StandardScaler
         from sklearn.metrics import confusion_matrix, accuracy_score
         from sklearn.metrics import confusion_matrix, accuracy_score
         from sklearn.metrics import classification_report
In [2]: df=pd.read csv("emails.csv")
In [3]: print(df.head() )
        print(df.info())
        df.isnull().sum()
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        [5 rows x 3002 columns]
        <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
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        Length: 3002, dtype: int64
In [4]: X = df.iloc[:, 1:-1].values
        y = df.iloc[:, -1].values
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.30, random_state=101)
         sc X = StandardScaler()
        X_train = sc_X.fit_transform(X_train)
        X_{\text{test}} = sc_{X} \cdot transform(X_{\text{test}})
         classifier = KNeighborsClassifier(n_neighbors=5)
        classifier.fit(X_train, y_train)
y_pred = classifier.predict(X_test)
         cm = confusion matrix(y test, y pred)
         cm
        array([[866, 248],
Out[4]:
                [ 16, 422]], dtype=int64)
In [5]:
        cl_report=classification_report(y_test,y_pred)
         print(cl_report)
        print("Accuracy Score for KNN : ", accuracy_score(y_pred,y_test))
                                     recall f1-score
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                                                  0.87
                                                             1114
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                                                              438
                                                  0.83
                                                             1552
             accuracy
                             0.81
                                        0.87
                                                  0.81
                                                             1552
            macro avq
                                        0.83
                                                             1552
        weighted avg
                             0.88
                                                  0.84
        Accuracy Score for KNN : 0.8298969072164949
In [6]:
        import pandas as pd
         import numpy as np
         import seaborn as sns
```

import matplotlib.pyplot as plt

```
from sklearn.model_selection import train_test_split
         from sklearn.svm import SVC
         from sklearn.metrics import accuracy_score
         from sklearn.metrics import confusion matrix, accuracy score
        from sklearn.metrics import classification_report
In [7]: df=pd.read_csv("emails.csv")
        print(df.head() )
        print(df.info())
        print(df.isnull().sum() )
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        [5 rows x 3002 columns]
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        memory usage: 118.5+ MB
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        Length: 3002, dtype: int64
In [8]: X = df.iloc[:, 1:-1].values
         y = df.iloc[:, -1].values
        X.shape
         X\_train, \ X\_test, \ y\_train, \ y\_test = train\_test\_split(X, \ y, \ test\_size=0.30, \ random\_state=101) 
         svc = SVC(C=1.0, kernel='rbf', gamma='auto')
         svc.fit(X_train,y_train)
        y_pred2 = svc.predict(X_test)
         cm = confusion_matrix(y_test, y_pred2)
Out[8]: array([[1078, 36],
                [ 103, 335]], dtype=int64)
In [9]:
        print("Accuracy Score for SVC : ", accuracy score(y pred2,y test))
         cl_report=classification_report(y_test,y_pred2)
        print(cl_report)
        Accuracy Score for SVC : 0.9104381443298969
                       precision
                                     recall f1-score
                                                         support
                    0
                            0.91
                                       0.97
                                                  0.94
                                       0.76
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                    1
                            0.90
                                                             438
                                                  0.91
                                                            1552
             accuracy
                            0.91
                                       0.87
                                                  0.88
                                                            1552
            macro avq
        weighted avg
                                       0.91
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```