```
In [1]:
          import pandas as pd
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
          from sklearn.metrics import classification_report, confusion_matrix
          data = pd.read_csv("emails.csv")
In [2]:
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                                                148
                                                       8
                 5172
          5172 rows × 3002 columns
In [3]: data = data.dron('Fmail No.'. axis=1)
In [4]: data.shape
Out[4]: (5172, 3001)
In [5]: data.describe()
Out[5]:
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            max
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                                                                        77.000000 189
 In [6]: data.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 [7]: data['Prediction'].value counts()
 Out[7]: 0
               3672
               1500
          1
          Name: Prediction, dtype: int64
 In [8]: X = data.drop('Prediction', axis = 1)
          v = data['Prediction']
 In [9]: from sklearn.model selection import train test split
          X train. X test. v train. v test = train test split(X, v, test size
In [10]: from sklearn.neighbors import KNeighborsClassifier
          neigh = KNeighborsClassifier(n neighbors = 2)
          neigh.fit(X train. v train)
Out[10]: KNeighborsClassifier(n neighbors=2)
In [11]: v pred = neigh.predict(X test)
In [12]: neigh.score(X_train, y_train)
          neiah.score(X test. v test)
Out[12]: 0.8695652173913043
In [13]: |print("Confusion Matrix: ")
          cm = confusion matrix(y test, y pred)
          Confusion Matrix:
Out[13]: array([[684, 48],
                 [ 87, 216]], dtype=int64)
```

```
In [14]: mat = ConfusionMatrixDisplay(confusion_matrix = cm)
          mat.plot()
          plt.show()
                                               600
                    684
                                  48
            0
                                              - 500
          Frue label
                                              - 400
                                              - 300
                    87
            1 .
                                              - 200
                                               100
                     Ó
                                   i
                       Predicted label
In [15]: print(classification report(v test. v pred))
                         precision
                                       recall
                                               f1-score
                                                           support
                      0
                              0.89
                                         0.93
                                                    0.91
                                                                732
                      1
                              0.82
                                         0.71
                                                    0.76
                                                                303
                                                    0.87
                                                               1035
              accuracy
                                         0.82
                                                    0.84
                                                               1035
                              0.85
             macro avg
          weighted avg
                              0.87
                                         0.87
                                                    0.87
                                                               1035
In [16]: print("accuracy_score: ")
          accuracv score(v test. v pred)
          accuracy score:
Out[16]: 0.8695652173913043
In [17]: print("precision_score: ")
          precision score(v test. v pred)
          precision_score:
Out[17]: 0.8181818181818182
In [18]: print("recall score: ")
          recall score(v test. v pred)
          recall_score:
Out[18]: 0.7128712871287128
In [19]: print("Error: ")
         1-accuracv score(v test. v pred)
          Error:
Out[19]: 0.13043478260869568
In [23]:
         from sklearn.svm import SVC
          SVM = SVC(gamma = 'auto')
```

```
SVM.fit(X train. v train)
Out[23]: SVC(gamma='auto')
In [22]: v pred = SVC.predict(v test)
          TypeError
                                                       Traceback (most recent
          call last)
          <ipython-input-22-dcf354cc4f53> in <module>
          ----> 1 y_pred = SVC.predict(y_test)
         TypeError: predict() missing 1 required positional argument: 'X'
In [44]: | SVM.score(X_train, y_train)
         SVM.score(X test. v test)
Out[44]: 0.8995169082125604
In [45]: print("Confusion Matrix: ")
         cm = confusion matrix(y test, y pred)
         Confusion Matrix:
Out[45]: array([[697, 31],
                 [ 73, 234]], dtype=int64)
         mat = ConfusionMatrixDisplay(confusion_matrix = cm)
In [46]:
         mat.plot()
         nlt.show()
                                              600
            0
                    697
                                  31
                                              500
          Frue label
                                              400
                                              - 300
                    73
                                              - 200
            1
                                              100
                     0
                                  1
                       Predicted label
In [47]: print(classification report(v test. v pred))
                         precision
                                       recall
                                               f1-score
                                                           support
                     0
                              0.91
                                         0.96
                                                    0.93
                                                                728
                      1
                              0.88
                                         0.76
                                                    0.82
                                                                307
                                                    0.90
                                                               1035
              accuracy
                              0.89
                                         0.86
                                                    0.87
                                                               1035
             macro avg
                              0.90
                                         0.90
                                                    0.90
                                                               1035
         weighted avg
```

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