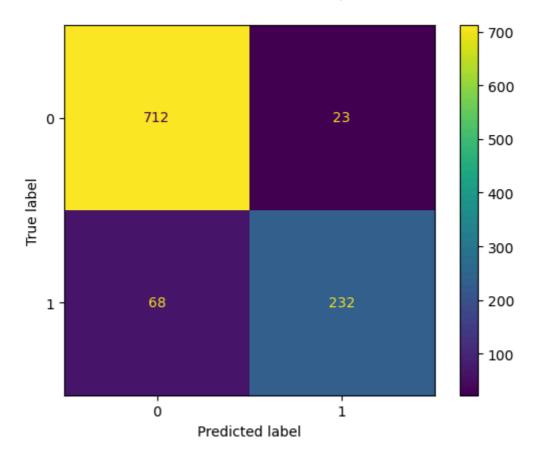
```
In [1]:
         import pandas as pd
         import numpy as np
         import seaborn as sns
         import matplotlib.pyplot as plt
         from sklearn.metrics import classification_report, confusion_matrix, ConfusionMa
         data = pd.read_csv("emails.csv")
In [2]:
         data
Out[2]:
                Email
                       the to ect and for of
                                                     a you hou ... connevey jay valued lay
                 No.
                Email
             0
                                            0
                                                0
                                                     2
                                                                                    0
                                                                                            0
                Email
             1
                                                2
                                                                                    0
                                                                                            0
                         8 13
                                 24
                                       6
                                            6
                                                   102
                                                           1
                                                               27
                                                                               0
                Email
             2
                             0
                                  1
                                       0
                                            0
                                                0
                                                     8
                                                           0
                                                                0
                                                                               0
                                                                                    0
                                                                                            0
                Email
                         0
             3
                             5
                                 22
                                       0
                                            5
                                                1
                                                    51
                                                           2
                                                               10
                                                                               0
                                                                                    0
                                                                                            0
                Email
             4
                         7
                                 17
                                       1
                                            5
                                                2
                                                    57
                                                           0
                                                                9
                                                                               0
                                                                                    0
                                                                                            0
                             6
                    5
                Email
         5167
                         2
                             2
                                  2
                                       3
                                            0
                                                    32
                                                           0
                                                                               0
                                                                                    0
                                                                                            0
                                                0
                                                                 0
                 5168
                Email
                        35
         5168
                                                5
                                                                                    0
                                                                                            0
                           27
                                 11
                                       2
                                            6
                                                   151
                                                                 3
                                                                               0
                5169
                Email
         5169
                             0
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                                                    11
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                                                                 0
                                                                               0
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                                                                                            0
                5170
                Email
         5170
                         2
                             7
                                       0
                                            2
                                                    28
                                                           2
                                                                 0
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                                                                                    0
                                                                                            0
                 5171
                Email
                        22 24
         5171
                                  5
                                       1
                                            6
                                                5 148
                                                           8
                                                                 2 ...
                                                                               0
                                                                                    0
                                                                                            0
                5172
        5172 rows × 3002 columns
         data = data.drop('Email No.', axis=1)
In [3]:
In [4]:
         data.shape
Out[4]: (5172, 3001)
In [5]:
         data.describe()
```

```
Out[5]:
                        the
                                      to
                                                  ect
                                                              and
                                                                           for
                                                                                        of
          count 5172.000000 5172.000000 5172.000000 5172.000000 5172.000000 5
                                6.188128
                                                         3.075599
                                                                      3.124710
                    6.640565
                                             5.143852
                                                                                   2.627030
          mean
                   11.745009
                                9.534576
                                            14.101142
                                                         6.045970
                                                                      4.680522
                                                                                   6.229845
            std
                    0.000000
                                0.000000
                                             1.000000
                                                         0.000000
                                                                      0.000000
                                                                                   0.000000
           min
           25%
                    0.000000
                                1.000000
                                             1.000000
                                                         0.000000
                                                                      1.000000
                                                                                  0.000000
           50%
                    3.000000
                                3.000000
                                             1.000000
                                                         1.000000
                                                                      2.000000
                                                                                   1.000000
           75%
                    8.000000
                                             4.000000
                                                         3.000000
                                7.000000
                                                                      4.000000
                                                                                   2.000000
                                           344.000000
                                                         89.000000
                  210.000000
                              132.000000
                                                                     47.000000
                                                                                  77.000000 1
           max
         8 rows × 3001 columns
 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]: Prediction
               3672
               1500
          Name: count, dtype: int64
 In [8]: X = data.drop('Prediction', axis = 1)
          y = data['Prediction']
         from sklearn.model_selection import train_test_split
 In [9]:
          X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.20)
In [10]:
         from sklearn.neighbors import KNeighborsClassifier
          neigh = KNeighborsClassifier(n_neighbors = 2)
          neigh.fit(X_train, y_train)
Out[10]:
                 KNeighborsClassifier
          KNeighborsClassifier(n_neighbors=2)
In [11]:
         y_pred = neigh.predict(X_test)
In [12]:
         neigh.score(X_train, y_train)
          neigh.score(X_test, y_test)
Out[12]: 0.8966183574879227
```

In [13]: print("Confusion Matrix: ")

```
cm = confusion_matrix(y_test, y_pred)
          cm
        Confusion Matrix:
Out[13]: array([[701, 34],
                 [ 73, 227]], dtype=int64)
In [14]: mat = ConfusionMatrixDisplay(confusion_matrix = cm)
          mat.plot()
          plt.show()
                                                                            700
                                                                            600
                                                     34
                          701
           0
                                                                           - 500
        True label
                                                                           - 400
                                                                            300
                          73
                                                     227
           1
                                                                           - 200
                                                                           - 100
                           0
                                                      1
                                  Predicted label
In [15]: print(classification_report(y_test, y_pred))
                       precision
                                    recall f1-score
                                                        support
                    0
                                      0.95
                                                 0.93
                            0.91
                                                            735
                    1
                            0.87
                                      0.76
                                                 0.81
                                                            300
                                                 0.90
                                                           1035
            accuracy
                            0.89
                                      0.86
                                                 0.87
                                                           1035
           macro avg
                            0.90
                                      0.90
                                                 0.89
        weighted avg
                                                           1035
In [16]: print("accuracy_score: ")
          accuracy_score(y_test, y_pred)
        accuracy_score:
Out[16]: 0.8966183574879227
In [17]: print("precision_score: ")
          precision_score(y_test, y_pred)
```

```
precision_score:
Out[17]: 0.8697318007662835
In [18]: print("recall_score: ")
         recall_score(y_test, y_pred)
        recall_score:
Out[18]: 0.756666666666667
In [19]: print("Error: ")
         1-accuracy_score(y_test, y_pred)
        Error:
Out[19]: 0.10338164251207727
In [20]: from sklearn.svm import SVC
         SVM = SVC(gamma = 'auto')
         SVM.fit(X_train, y_train)
Out[20]:
                SVC
         SVC(gamma='auto')
In [22]: y_pred = SVM.predict(X_test)
In [23]: SVM.score(X_train, y_train)
         SVM.score(X_test, y_test)
Out[23]: 0.9120772946859903
In [24]: print("Confusion Matrix: ")
         cm = confusion_matrix(y_test, y_pred)
        Confusion Matrix:
Out[24]: array([[712, 23],
                 [ 68, 232]], dtype=int64)
In [25]: mat = ConfusionMatrixDisplay(confusion_matrix = cm)
         mat.plot()
         plt.show()
```



print(classification_report(y_test, y_pred)) precision recall f1-score support 0 0.91 0.97 0.94 735 1 0.91 0.77 0.84 300 0.91 accuracy 1035 0.91 0.87 0.89 macro avg 1035 weighted avg 0.91 0.91 0.91 1035

In []: