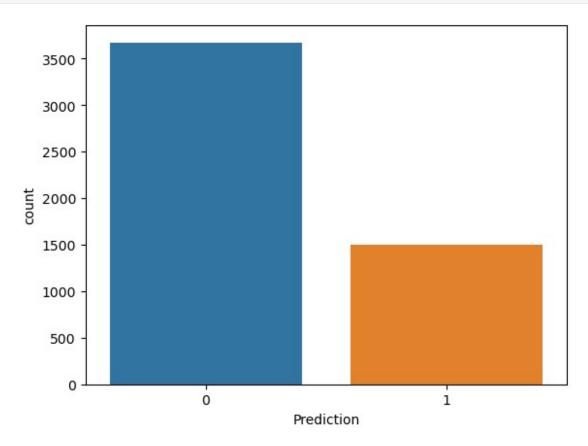
```
import pandas as pd
df=pd.read_csv("emails.csv")
df.sample(5)
       Email No. the to ect and for of
                                               a you hou ...
connevey \
4321 Email 4322
                        0
                    0
                             1
                                  0
                                       0
                                           0
                                               0
                                                    0
                                                         0 ...
4558 Email 4559 18
                       14
                             4
                                  3
                                           0
                                              93
2706 Email 2707
                    1
                        4
                             1
                                              22
2665 Email 2666
                    4
                        0
                             1
                                              20
                                                    0
                                  0
                                       5
                                           0
4834 Email 4835
                    2
                        2
                             2
                                           1 60
                               4
                        infrastructure military allowing
      jay valued
                  lay
                                                            ff
                                                                dry \
4321
                0
                                                             0
4558
                0
                     0
                                     0
                                                             0
        0
                                               0
                                                         0
                                                                   0
                0
                     0
                                     0
                                               0
                                                             1
                                                                   0
2706
        0
                                                         0
2665
        0
                0
                     0
                                     0
                                               0
                                                         0
                                                             0
                                                                   0
4834
                0
                     0
                                     0
                                               0
                                                              0
                                                                   0
        0
      Prediction
4321
4558
               0
2706
               1
               0
2665
4834
               0
[5 rows x 3002 columns]
df.shape
(5172, 3002)
# The last column has the labels for prediction : 1 for spam, 0 for
not spam.
# The remaining 3000 columns are the 3000 most common words in all the
emails, after excluding the non-alphabetical characters/words.
# split input and output data
X=df.drop(['Email No.', 'Prediction'],axis=1)
Y=df['Prediction']
X.shape
(5172, 3000)
```

```
import seaborn as sns
sns.countplot(x=Y)

<Axes: xlabel='Prediction', ylabel='count'>
```



```
from sklearn.model_selection import train_test_split
X_train,X_test,Y_train,Y_test=train_test_split(X,Y,test_size=0.3,rando
m_state=42)

# k- nearest neighbors

from sklearn.neighbors import KNeighborsClassifier
knn_m=KNeighborsClassifier(n_neighbors=5)

# the n_neighbors parameter specifies the number of nearest neighbors
to consider when making a prediction.
knn_m.fit(X_train,Y_train)

KNeighborsClassifier()

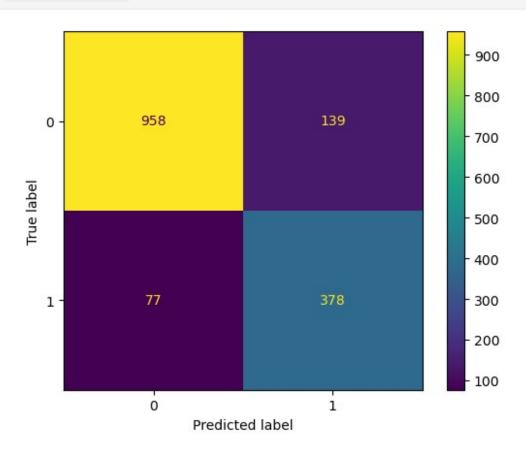
y_pred_knn=knn_m.predict(X_test)
y_pred_knn
array([0, 0, 1, ..., 0, 0, 0], dtype=int64)
```

## # evaluate

from sklearn.metrics import accuracy\_score, classification\_report, ConfusionMatrixDisplay

ConfusionMatrixDisplay.from\_predictions(Y\_test,y\_pred\_knn)

<sklearn.metrics.\_plot.confusion\_matrix.ConfusionMatrixDisplay at
0x1a8fe5c1110>



knn\_accuracy = accuracy\_score(Y\_test, y\_pred\_knn) knn\_report = classification\_report(Y\_test, y\_pred\_knn) print("Accuracy : ",knn accuracy) print("Report : ",knn\_report) Accuracy: 0.8608247422680413 Report: precision recall f1-score support 0 0.93 0.90 1097 0.87 1 0.73 0.83 0.78 455 1552 0.86 accuracy 0.83 0.85 0.84 1552 macro avg

```
weighted avg
                   0.87
                             0.86
                                       0.86
                                                  1552
## Support Vector Machine
from sklearn.svm import SVC
svm model=SVC()
svm_model.fit(X_test,Y_test)
SVC()
y_pred_svm=svm_model.predict(X_test)
y pred svm
array([0, 0, 1, ..., 0, 0, 0], dtype=int64)
svm_accuracy = accuracy_score(Y_test, y_pred_svm)
svm_report = classification_report(Y_test, y_pred_svm)
print("Accuracy : ",svm_accuracy)
print("Report : ",svm_report)
Accuracy: 0.7744845360824743
Report :
                        precision
                                      recall f1-score
                                                         support
           0
                   0.76
                             0.99
                                        0.86
                                                  1097
           1
                   0.89
                             0.26
                                        0.41
                                                   455
                                        0.77
                                                  1552
    accuracy
                                        0.63
                                                  1552
   macro avq
                   0.83
                             0.63
weighted avg
                   0.80
                             0.77
                                        0.73
                                                  1552
```