In [1]: **import** numpy **as** np import pandas as pd from sklearn.model_selection import train_test_split from sklearn.naive_bayes import GaussianNB import matplotlib.pyplot as plt import seaborn as sns from sklearn.metrics import confusion_matrix,ConfusionMatrixDisplay,classification_report,accuracy_score, precision_score, recall_score, f1_score from sklearn.preprocessing import LabelEncoder In [21]: data = pd.read_csv('Iris.csv') data.head(5) Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Species Out[21]: 5.1 3.5 0.2 Iris-setosa 4.9 3.0 1.4 0.2 Iris-setosa 4.7 3.2 1.3 0.2 Iris-setosa **2** 3 4.6 3.1 1.5 0.2 Iris-setosa 5.0 3.6 1.4 0.2 Iris-setosa In [22]: data.describe(include = 'all') Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm Out[22]: 150.000000 count 150.000000 150.000000 150.000000 150.000000 150 NaN NaN NaN NaN NaN NaN NaN NaN Iris-setosa NaN NaN NaN NaN NaN 75.500000 5.843333 3.054000 3.758667 1.198667 NaN mean **std** 43.445368 0.828066 0.433594 1.764420 0.763161 NaN 1.000000 4.300000 2.000000 1.000000 0.100000 min NaN 5.100000 2.800000 0.300000 NaN **25**% 38.250000 1.600000 **50**% 75.500000 5.800000 3.000000 1.300000 4.350000 NaN **75**% 112.750000 6.400000 3.300000 5.100000 1.800000 NaN max 150.000000 7.900000 4.400000 6.900000 2.500000 NaN In [23]: data.info() <class 'pandas.core.frame.DataFrame'> RangeIndex: 150 entries, 0 to 149 Data columns (total 6 columns): # Column Non-Null Count Dtype --- ---------0 Id 150 non-null int64 1 SepalLengthCm 150 non-null float64 2 SepalWidthCm 150 non-null float64 3 PetalLengthCm 150 non-null float64 4 PetalWidthCm 150 non-null float64 5 Species 150 non-null object dtypes: float64(4), int64(1), object(1) memory usage: 7.2+ KB In [24]: print(data.shape) data['Species'].unique() (150, 6)Out[24]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object) In []: In [25]: data.isnull().sum() Out[25]: **Id** SepalLengthCm SepalWidthCm PetalLengthCm 0 PetalWidthCm Species dtype: int64 In [1]: x = data.iloc[:,1:5]y = data.iloc[:,5:]Traceback (most recent call last) /tmp/ipykernel_5714/2831424731.py in <module> ----> 1 x = data.iloc[:,1:5]**2** y = data.iloc[:,5:] NameError: name 'data' is not defined In [27]: encode = LabelEncoder() In [28]: y = encode.fit_transform(y) /home/ubuntu/.local/lib/python3.10/site-packages/sklearn/preprocessing/_label.py:115: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel(). y = column_or_1d(y, warn=True) In [29]: x_train,x_test,y_train,y_test = train_test_split(x,y,test_size = 0.3,random_state = 0) In [30]: naive_bayes = GaussianNB() naive_bayes.fit(x_train,y_train) pred = naive_bayes.predict(x_test) In [31]: pred Out[31]: array([2, 1, 0, 2, 0, 2, 0, 1, 1, 1, 2, 1, 1, 1, 1, 0, 1, 1, 0, 0, 2, 1, 0, 0, 2, 0, 0, 1, 1, 0, 2, 1, 0, 2, 2, 1, 0, 1, 1, 1, 2, 0, 2, 0, 0]) In [32]: y_test Out[32]: array([2, 1, 0, 2, 0, 2, 0, 1, 1, 1, 2, 1, 1, 1, 1, 0, 1, 1, 0, 0, 2, 1, 0, 0, 2, 0, 0, 1, 1, 0, 2, 1, 0, 2, 2, 1, 0, 1, 1, 1, 2, 0, 2, 0, 0]) In [33]: matrix = confusion_matrix(y_test, pred, labels = naive_bayes.classes_) print(matrix) tp, fn, fp, tn = confusion_matrix(y_test, pred, labels=[1,0]).reshape(-1) [[16 0 0] [0 18 0] [0 0 11]] In [34]: conf_matrix = ConfusionMatrixDisplay(confusion_matrix=matrix, display_labels=naive_bayes.classes_) conf_matrix.plot(cmap=plt.cm.YlGn) plt.show()

In [35]: print(classification_report(y_test,pred))

Predicted label

precision recall f1-score support 16 1.00 1.00 1.00 1.00 1.00 18 1.00 1.00 1.00 11 45 1.00 accuracy macro avg 1.00 1.00 1.00 45 1.00 weighted avg 1.00 1.00 45

In [36]: print('\nAccuracy: {:.2f}'.format(accuracy_score(y_test,pred)))
print('Error Rate: ',(fp+fn)/(tp+tn+fn+fp))
print('Sensitivity (Recall or True positive rate) :',tp/(tp+fn))
print('Specificity (True negative rate) :',tn/(fp+tn))
print('Precision (Positive predictive value) :',tp/(tp+fp))
print('False Positive Rate :',fp/(tn+fp))

Accuracy: 1.00 Error Rate: 0.0 Sensitivity (Recall or True positive rate) : 1.0 Specificity (True negative rate) : 1.0 Precision (Positive predictive value) : 1.0 False Positive Rate : 0.0