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,classifica
 from sklearn.preprocessing import LabelEncoder

In [8]: data = pd.read_csv('Iris.csv')
 data.head(5)

Out[8]:

	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
0	1	5.1	3.5	1.4	0.2	Iris-setosa
1	2	4.9	3.0	1.4	0.2	Iris-setosa
2	3	4.7	3.2	1.3	0.2	Iris-setosa
3	4	4.6	3.1	1.5	0.2	Iris-setosa
4	5	5.0	3.6	1.4	0.2	Iris-setosa

In [9]: data.describe(include = 'all')

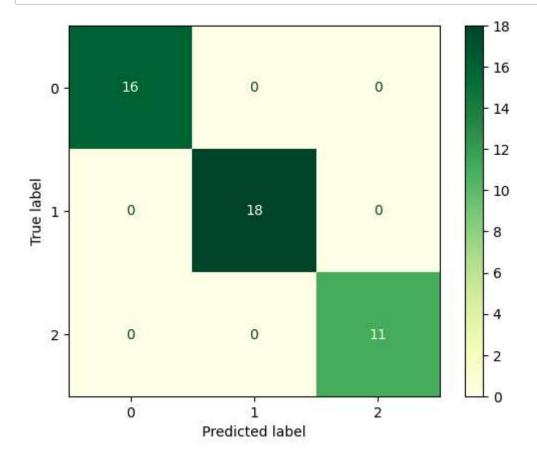
Out[9]:

	ld	SepalLengthCm	SepalWidthCm	PetalLengthCm	PetalWidthCm	Species
count	150.000000	150.000000	150.000000	150.000000	150.000000	150
unique	NaN	NaN	NaN	NaN	NaN	3
top	NaN	NaN	NaN	NaN	NaN	Iris-setosa
freq	NaN	NaN	NaN	NaN	NaN	50
mean	75.500000	5.843333	3.054000	3.758667	1.198667	NaN
std	43.445368	0.828066	0.433594	1.764420	0.763161	NaN
min	1.000000	4.300000	2.000000	1.000000	0.100000	NaN
25%	38.250000	5.100000	2.800000	1.600000	0.300000	NaN
50%	75.500000	5.800000	3.000000	4.350000	1.300000	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 [10]: data.info()
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 150 entries, 0 to 149
         Data columns (total 6 columns):
                             Non-Null Count Dtype
              Column
              -----
                             -----
                                             ----
              Ιd
                                             int64
          0
                             150 non-null
          1
              SepalLengthCm 150 non-null
                                             float64
              SepalWidthCm
                             150 non-null
                                             float64
          2
          3
              PetalLengthCm 150 non-null
                                             float64
          4
              PetalWidthCm
                             150 non-null
                                             float64
          5
              Species
                             150 non-null
                                             obiect
         dtypes: float64(4), int64(1), object(1)
         memory usage: 7.2+ KB
In [11]: print(data.shape)
         data['Species'].unique()
         (150, 6)
Out[11]: array(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], dtype=object)
In [12]: data.isnull().sum()
Out[12]: Id
                          0
         SepalLengthCm
                          0
         SepalWidthCm
         PetalLengthCm
         PetalWidthCm
                          0
         Species
                          0
         dtype: int64
In [14]: x = data.iloc[:,1:5]
         y = data.iloc[:,5:]
In [15]: encode = LabelEncoder()
         y = encode.fit transform(y)
         C:\Users\Lenovo\anaconda3\lib\site-packages\sklearn\preprocessing\ label.py:1
         15: DataConversionWarning: A column-vector y was passed when a 1d array was e
         xpected. Please change the shape of y to (n_samples, ), for example using rav
         el().
           y = column_or_1d(y, warn=True)
In [16]: x train,x test,y train,y test = train test split(x,y,test size = 0.3,random st
```

```
In [17]: naive_bayes = GaussianNB()
         naive_bayes.fit(x_train,y_train)
         pred = naive_bayes.predict(x_test)
In [18]: pred
Out[18]: 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,
                01)
In [19]: y_test
Out[19]: 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 [20]: 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]]
```

conf_matrix = ConfusionMatrixDisplay(confusion_matrix=matrix,display_labels=na In [21]: conf_matrix.plot(cmap=plt.cm.YlGn) plt.show()



In [22]: print(classification_report(y_test,pred))

	precision	recall	f1-score	support
0	1.00	1.00	1.00	16
1	1.00	1.00	1.00	18
2	1.00	1.00	1.00	11
accuracy			1.00	45
macro avg	1.00	1.00	1.00	45
weighted avg	1.00	1.00	1.00	45

```
In [23]: 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

In []: