

```
In [1]: import pandas as pd
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')
warnings.simplefilter('ignore')
```

```
In [2]: data = pd.read_csv("https://raw.githubusercontent.com/plotly/datasets/master/iris-csv")
```

```
In [3]: data.shape
```

```
Out[3]: (150, 5)
```

```
In [4]: data.head()
```

```
Out[4]:
```

	sepal length	sepal width	petal length	petal width	class
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

```
In [5]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 5 columns):
 #   Column          Non-Null Count  Dtype
---  -
 0   sepal length    150 non-null    float64
 1   sepal width     150 non-null    float64
 2   petal length    150 non-null    float64
 3   petal width     150 non-null    float64
 4   class           150 non-null    object
dtypes: float64(4), object(1)
memory usage: 6.0+ KB
```

```
In [6]: data.describe()
```

```
Out[6]:
```

	sepal length	sepal width	petal length	petal width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

```
In [7]: data.isnull().sum()
```

```
Out[7]: sepal length    0
        sepal width    0
        petal length   0
        petal width    0
        class          0
        dtype: int64
```

```
In [8]: X = data.drop(['class'], axis=1)
        y = data.drop(['sepal length', 'sepal width', 'petal length', 'petal width'], axis=1)
```

```
In [9]: print(X.shape)
        print(y.shape)

(150, 4)
(150, 1)
```

```
In [12]: from sklearn.model_selection import train_test_split
        X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, shuffle=True)
        print(X_train.shape)
        print(X_test.shape)
        print(y_train.shape)
        print(y_test.shape)

(120, 4)
(30, 4)
(120, 1)
(30, 1)
```

```
In [13]: from sklearn.naive_bayes import GaussianNB
        model = GaussianNB()
        model.fit(X_train, y_train)
```

```
Out[13]: GaussianNB()
```

```
In [14]: y_pred = model.predict(X_test)
        model.score(X_test, y_test)
```

```
Out[14]: 0.9666666666666667
```

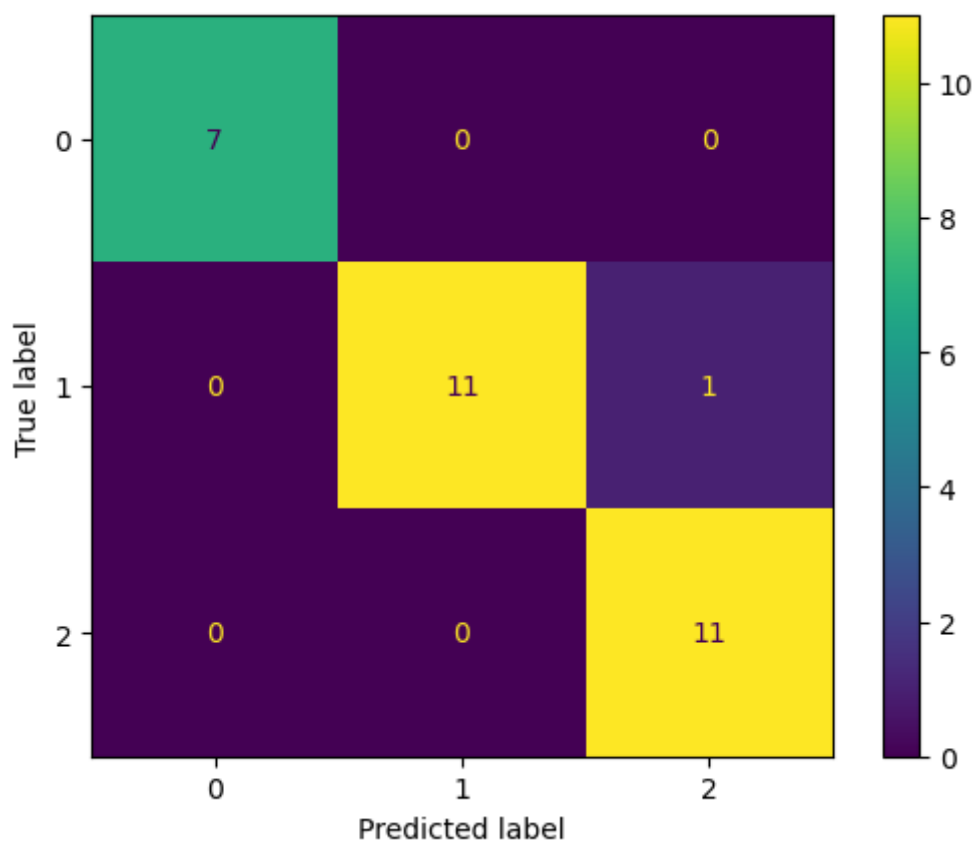
```
In [15]: from sklearn.metrics import accuracy_score, confusion_matrix, ConfusionMatrixDisplay
        print(accuracy_score(y_test, y_pred))

0.9666666666666667
```

```
In [16]: cm = confusion_matrix(y_test, y_pred)
        disp = ConfusionMatrixDisplay(confusion_matrix = cm)
        print("Confusion matrix:")
        print(cm)

Confusion matrix:
[[ 7  0  0]
 [ 0 11  1]
 [ 0  0 11]]
```

```
In [17]: disp.plot()
        plt.show()
```



```
In [19]: def get_confusion_matrix_values(y_true, y_pred):
          cm = confusion_matrix(y_true, y_pred)
          return(cm[0][0], cm[0][1], cm[1][0], cm[1][1])

          TP, FP, FN, TN = get_confusion_matrix_values(y_test, y_pred)
          print("TP: ", TP)
          print("FP: ", FP)
          print("FN: ", FN)
          print("TN: ", TN)
```

TP: 7
FP: 0
FN: 0
TN: 11

```
In [20]: print("The Accuracy is ", (TP+TN)/(TP+TN+FP+FN))
          print("The precision is ", TP/(TP+FP))
          print("The recall is ", TP/(TP+FN))
```

The Accuracy is 1.0
The precision is 1.0
The recall is 1.0

In []: