```
In [43]: import pandas as pd
import matplotlib.pyplot as plt
```

In [60]: data = pd.read\_csv("iris-data.csv")
 data.head()

Out[60]:

	sepal length	sepal width	petal length	petal width	class
(	5.1	3.5	1.4	0.2	Iris-setosa
	4.9	3.0	1.4	0.2	Iris-setosa
2	2 4.7	3.2	1.3	0.2	Iris-setosa
;	<b>3</b> 4.6	3.1	1.5	0.2	Iris-setosa
	<b>5</b> .0	3.6	1.4	0.2	Iris-setosa

In [61]: data.shape

Out[61]: (150, 5)

In [62]: data.head()

Out[62]:

	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 [63]: data.tail()

Out[63]:

	sepal length	sepal width	petal length	petal width	class
145	6.7	3.0	5.2	2.3	Iris-virginica
146	6.3	2.5	5.0	1.9	Iris-virginica
147	6.5	3.0	5.2	2.0	Iris-virginica
148	6.2	3.4	5.4	2.3	Iris-virginica
149	5.9	3.0	5.1	1.8	Iris-virginica

# In [64]: 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 sepal width 150 non-null float64 1 petal length 150 non-null float64 2 3 petal width 150 non-null float64 4 class 150 non-null object

dtypes: float64(4), object(1)

memory usage: 6.0+ KB

## In [65]: data.describe()

#### Out[65]:

	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 [66]: data.isnull().sum()

#### Out[66]: sepal length

0 sepal width 0 petal length 0 petal width 0 class

dtype: int64

```
In [67]: | X = data.drop(['class'], axis=1)
         y = data.drop(['sepal length', 'sepal width', 'petal length', 'petal width']
         print(X)
         print(y)
         print(X.shape)
         print(y.shape)
               sepal length sepal width petal length petal width
         0
                        5.1
                                      3.5
                                                    1.4
                                                                  0.2
         1
                        4.9
                                      3.0
                                                    1.4
                                                                  0.2
         2
                        4.7
                                      3.2
                                                    1.3
                                                                  0.2
                                                                  0.2
         3
                        4.6
                                      3.1
                                                    1.5
         4
                        5.0
                                      3.6
                                                    1.4
                                                                  0.2
                                      . . .
                        . . .
                                                     . . .
                                                                  . . .
                        6.7
                                      3.0
                                                    5.2
                                                                  2.3
         145
         146
                        6.3
                                      2.5
                                                    5.0
                                                                  1.9
         147
                        6.5
                                      3.0
                                                    5.2
                                                                  2.0
         148
                        6.2
                                      3.4
                                                    5.4
                                                                  2.3
         149
                        5.9
                                      3.0
                                                    5.1
                                                                  1.8
         [150 rows x 4 columns]
                        class
         0
                  Iris-setosa
                  Iris-setosa
         1
         2
                  Iris-setosa
         3
                  Iris-setosa
         4
                  Iris-setosa
         145 Iris-virginica
         146 Iris-virginica
         147
              Iris-virginica
         148
              Iris-virginica
         149 Iris-virginica
         [150 rows x 1 columns]
          (150, 4)
          (150, 1)
In [68]:
         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, shuff]
         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 [69]: from sklearn.naive_bayes import GaussianNB
model = GaussianNB()
model.fit(X_train, y_train)
```

C:\Users\admin\Anaconda3\lib\site-packages\sklearn\utils\validation.py:993: D
ataConversionWarning: A column-vector y was passed when a 1d array was expect
ed. Please change the shape of y to (n\_samples, ), for example using ravel().
 y = column\_or\_1d(y, warn=True)

Out[69]: GaussianNB()

```
In [70]: y_pred = model.predict(X_test)
model.score(X_test,y_test)
```

Out[70]: 0.966666666666667

In [71]: from sklearn.metrics import accuracy\_score, confusion\_matrix, ConfusionMatrixDi
print(accuracy\_score(y\_test, y\_pred))

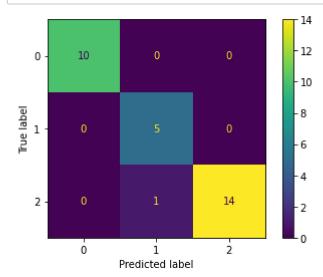
0.96666666666666

```
In [72]: cm = confusion_matrix(y_test, y_pred)
    disp = ConfusionMatrixDisplay(confusion_matrix = cm)
    print("Confusion matrix:")
    print(cm)
```

Confusion matrix:

```
[[10 0 0]
[ 0 5 0]
[ 0 1 14]]
```

```
In [73]: disp.plot()
  plt.show()
```



```
In [74]: | 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: 10
          FP:
               0
          FN:
               0
          TN: 5
          print("The Accuracy is ", (TP+TN)/(TP+TN+FP+FN))
print("The precision is ", TP/(TP+FP))
In [75]:
          print("The recall is ", TP/(TP+FN))
          The Accuracy is 1.0
          The precision is 1.0
          The recall is 1.0
In [ ]:
```