```
In [5]:
            import pandas as pd
            import numpy as np
            import matplotlib.pyplot as plt
            import seaborn as sns
            from sklearn.metrics import accuracy_score
            from sklearn.datasets import load_iris
            from sklearn.model_selection import train_test_split
            from sklearn.naive_bayes import GaussianNB
 In [7]: | iris = load_iris()
In [11]: | data = pd.DataFrame(iris.data)
In [17]:
           data
Out[17]:
                 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
                                                             0.2
                                                                   Iris-setosa
                                                  1.4
              2
                         4.7
                                     3.2
                                                             0.2
                                                                   Iris-setosa
                                                  1.3
              3
                         4.6
                                     3.1
                                                             0.2
                                                                   Iris-setosa
                                                  1.5
                         5.0
                                     3.6
                                                  1.4
                                                             0.2
              4
                                                                   Iris-setosa
                                      ...
            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
           150 rows × 5 columns
           sepal width
                              0
```

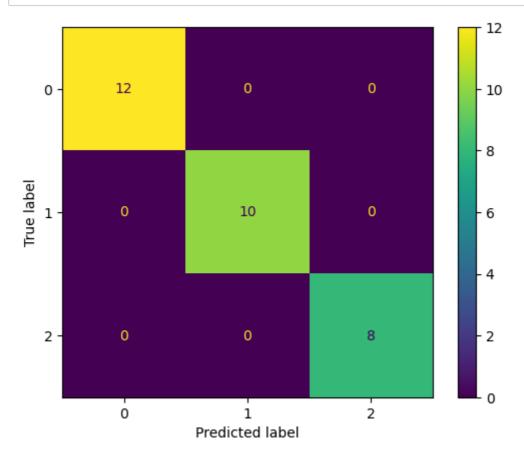
```
Untitled53 - Jupyter Notebook
In [25]:
         X = data.drop(['class'], axis=1)
         y = data.drop(['sepal length', 'sepal width', 'petal length', 'petal widt
          print(X)
          print(y)
          print(X.shape)
         print(y.shape)
               sepal length sepal width petal length petal width
         0
                        5.1
                                     3.5
                                                    1.4
                                     3.0
          1
                        4.9
                                                    1.4
                                                                  0.2
          2
                        4.7
                                      3.2
                                                    1.3
                                                                  0.2
          3
                                                                  0.2
                        4.6
                                     3.1
                                                    1.5
          4
                                                                  0.2
                        5.0
                                     3.6
                                                    1.4
                        . . .
                                      . . .
                                                    . . .
          145
                        6.7
                                     3.0
                                                    5.2
                                                                  2.3
                                                                  1.9
          146
                        6.3
                                     2.5
                                                    5.0
          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
                  Iris-setosa
         0
                  Iris-setosa
          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 [27]: 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, shu
          print(X_train.shape)
          print(X_test.shape)
          print(y_train.shape)
          print(y_test.shape)
          (120, 4)
```

```
localhost:8888/notebooks/Untitled53.ipynb?kernel_name=python3
```

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

```
In [29]:
         from sklearn.naive_bayes import GaussianNB
         model = GaussianNB()
         model.fit(X_train, y_train)
         C:\Users\System21\anaconda3\Lib\site-packages\sklearn\utils\validation.py:
         1184: DataConversionWarning: A column-vector y was passed when a 1d array
         was expected. Please change the shape of y to (n_samples, ), for example u
         sing ravel().
           y = column_or_1d(y, warn=True)
Out[29]:
         ▼ Gaus$ianNB
          GaussianNB()
In [31]: y_pred = model.predict(X_test)
         model.score(X_test,y_test)
Out[31]: 1.0
In [34]: | from sklearn.metrics import accuracy_score, confusion_matrix, ConfusionMatri
         print(accuracy_score(y_test, y_pred))
         1.0
In [36]: | cm = confusion_matrix(y_test, y_pred)
         disp = ConfusionMatrixDisplay(confusion_matrix = cm)
         print("Confusion matrix:")
         print(cm)
         Confusion matrix:
         [[12 0 0]
          [ 0 10 0]
          [0 0 8]]
```

```
In [38]: disp.plot()
plt.show()
```



```
In [40]:
         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:
               12
          FP:
               0
          FN:
               0
          TN:
               10
```

```
In [42]: print("The Accuracy is ", (TP+TN)/(TP+TN+FP+FN))
    print("The precision is ", TP/(TP+FP))
    print("Error Rate:",1-(TP+TN)/(TP+TN+FP+FN))
    print("The recall is ", TP/(TP+FN))
```

```
The Accuracy is 1.0
The precision is 1.0
error rate: 0.0
The recall is 1.0
```