

```
In [7]: # Import libraries
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import urllib.request
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import accuracy_score, confusion_matrix,
precision_score, recall_score
import seaborn as sns
```

```
In [4]: # URL of the Iris dataset from UCI Repository
url = "https://archive.ics.uci.edu/ml/machine-learning-
databases/iris/iris.data"

# Column names based on dataset description
column_names = ["Sepal_Length", "Sepal_Width", "Petal_Length", "Petal_Width",
"Class"]

# Load the dataset into a Pandas DataFrame
iris = pd.read_csv(url, names=column_names)
```

```
In [6]: # Initialize the data frame
iris.head()
```

```
Out[6]:
```

	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

Perform Data Preprocessing

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In [9]: # Convert categorical target variable to numerical
label_encoder = LabelEncoder()
iris["Class"] = label_encoder.fit_transform(iris["Class"])
```

```
In [10]: iris.head()
```

```
Out[10]:
```

	Sepal_Length	Sepal_Width	Petal_Length	Petal_Width	Class
0	5.1	3.5	1.4	0.2	0
1	4.9	3.0	1.4	0.2	0
2	4.7	3.2	1.3	0.2	0
3	4.6	3.1	1.5	0.2	0
4	5.0	3.6	1.4	0.2	0

```
In [11]: # Check for null values  
iris.isnull().sum()
```

```
Out[11]: Sepal_Length    0  
Sepal_Width    0  
Petal_Length    0  
Petal_Width    0  
Class          0  
dtype: int64
```

```
In [13]: # Divide dataset into Independent (X) and Dependent (Y) variables  
X = iris.drop(columns=["Class"]) # Features  
Y = iris["Class"] # Target variable
```

```
In [15]: # Split dataset into training and testing datasets (80% training, 20%  
testing)  
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.2,  
random_state=42)
```

```
In [17]: # Feature Scaling  
from sklearn import preprocessing  
# Apply Min-Max Scaling only on numerical columns  
scaler = preprocessing.MinMaxScaler()  
features_scaled = scaler.fit_transform(X)  
  
# Create a new DataFrame with scaled features and original labels  
iris_normalized = pd.DataFrame(features_scaled, columns=X.columns)  
iris_normalized["Class"] = Y # Add the class column back  
  
# Display first few rows  
iris_normalized.head()
```

Out[17]:	Sepal_Length	Sepal_Width	Petal_Length	Petal_Width	Class
0	0.222222	0.625000	0.067797	0.041667	0
1	0.166667	0.416667	0.067797	0.041667	0
2	0.111111	0.500000	0.050847	0.041667	0
3	0.083333	0.458333	0.084746	0.041667	0
4	0.194444	0.666667	0.067797	0.041667	0

```
In [18]: # Train the Model using Naïve Bayes
naive_bayes = GaussianNB()
naive_bayes.fit(X_train, y_train)
```

```
Out[18]: GaussianNB()
```

```
In [19]: # Make Predictions
y_pred_train = naive_bayes.predict(X_train)
y_pred_test = naive_bayes.predict(X_test)
```

```
In [20]: # Evaluate Model Performance
train_accuracy = accuracy_score(y_train, y_pred_train)
test_accuracy = accuracy_score(y_test, y_pred_test)
```

```
In [21]: # Evaluate Model Performance
train_accuracy = accuracy_score(y_train, y_pred_train)
test_accuracy = accuracy_score(y_test, y_pred_test)

precision = precision_score(y_test, y_pred_test, average="micro")
recall = recall_score(y_test, y_pred_test, average="micro")
```

```
In [23]: # Compute Confusion Matrix
cm = confusion_matrix(y_test, y_pred_test)
```

```
Out[23]: array([[10,  0,  0],
               [ 0,  9,  0],
               [ 0,  0, 11]], dtype=int64)
```

```
In [24]: # Display results
print("\nTraining Accuracy:", train_accuracy)
print("Testing Accuracy:", test_accuracy)
print("Precision:", precision)
print("Recall:", recall)
print("\nConfusion Matrix:\n", cm)
```

Training Accuracy: 0.95

Testing Accuracy: 1.0

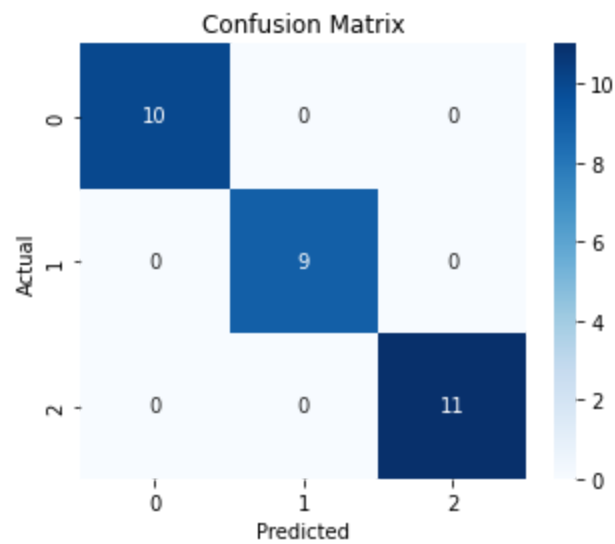
Precision: 1.0

Recall: 1.0

Confusion Matrix:

```
[[10  0  0]
 [ 0  9  0]
 [ 0  0 11]]
```

```
In [25]: # Visualizing Confusion Matrix
plt.figure(figsize=(5, 4))
sns.heatmap(cm, annot=True, fmt="d", cmap="Blues")
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.title("Confusion Matrix")
plt.show()
```



In [ ]: