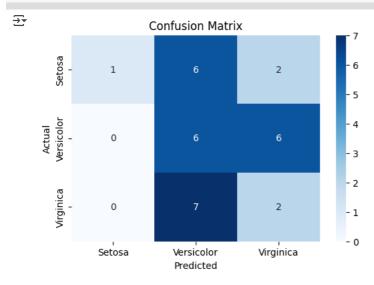
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.model selection import train test split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import confusion_matrix, accuracy_score, precision_score, recall_score
# Set random seed for reproducibility
np.random.seed(42)
# Number of samples
num_samples = 150
# Generate random synthetic features
sepal_length = np.random.uniform(4.3, 7.9, num_samples)
sepal_width = np.random.uniform(2.0, 4.4, num_samples)
petal_length = np.random.uniform(1.0, 6.9, num_samples)
petal_width = np.random.uniform(0.1, 2.5, num_samples)
# Generate class labels (3 species)
species = np.random.choice(['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'], num_samples)
# Create DataFrame
df = pd.DataFrame({
    'SepalLength': sepal_length,
    'SepalWidth': sepal_width,
    'PetalLength': petal_length,
    'PetalWidth': petal_width,
    'Species': species
})
# Display first few rows
df.head()
<del>____</del>
         SepalLength SepalWidth PetalLength PetalWidth
                                                               Species
      0
            5.648344
                        4.179838
                                      1.304922
                                                  1.965153
                                                              Iris-setosa
            7 722572
                        2 574949
                                     4 134992
                                                  1.440170 Iris-versicolor
      1
      2
            6.935178
                        2.347748
                                     4.189747
                                                  1.118133 Iris-versicolor
      3
            6 455171
                        3 174687
                                     4 760836
                                                  2 275251
                                                              Iris-setosa
            4.861667
                        4.365561
                                     5.283939
                                                  0.366874 Iris-versicolor
# Define features (X) and target (y)
X = df.drop(columns=['Species'])
y = df['Species']
# Split into training and testing sets (80% train, 20% test)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Initialize the model
model = GaussianNB()
# Train the model
model.fit(X_train, y_train)
→
     ▼ GaussianNB
     GaussianNB()
# Predict class labels
y_pred = model.predict(X_test)
# Compute confusion matrix
cm = confusion_matrix(y_test, y_pred, labels=['Iris-setosa', 'Iris-versicolor', 'Iris-virginica'])
# Compute accuracy, error rate, precision, and recall
accuracy = accuracy_score(y_test, y_pred)
error_rate = 1 - accuracy
precision = precision_score(y_test, y_pred, average='macro') # Macro because it's multi-class
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```
recall = recall_score(y_test, y_pred, average='macro')

# Print results
print(f"Confusion Matrix:\n{cm}")
print(f"Accuracy: {accuracy:.2f}")
print(f"Error Rate: {error_rate:.2f}")
print(f"Precision: {precision:.2f}")
print(f"Recall: {recall:.2f}")
```

Confusion Matrix:
[[1 6 2]
 [0 6 6]
 [0 7 2]]
 Accuracy: 0.30
 Error Rate: 0.70
 Precision: 0.51
 Recall: 0.28

```
# Plot confusion matrix
plt.figure(figsize=(6,4))
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['Setosa', 'Versicolor', 'Virginica'], yticklabels=['Setosa', 'Versicolor
plt.xlabel("Predicted")
plt.ylabel("Actual")
plt.title("Confusion Matrix")
plt.show()
```



df.to_csv("synthetic_iris.csv", index=False)
print("Dataset saved as synthetic_iris.csv")

 \Rightarrow Dataset saved as synthetic_iris.csv