1: Implement Naïve Bayes method using scikit-learn library

Use dataset available with name glass

Use train_test_split to create training and testing part Evaluate the model on test part using score and classification_report(y_true, y_pred)

```
In [2]: import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.naive_bayes import GaussianNB
from sklearn.metrics import classification report
# Load the dataset from the CSV file
glass_data = pd.read_csv("glass.csv")
# Split the dataset into features (X) and target (y)
X = glass_data.drop('Type', axis=1)
y = glass_data['Type']
# Split the dataset into training and testing parts
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
# Create and train the Naïve Bayes classifier
classifier = GaussianNB()
classifier.fit(X_train, y_train)
# Make predictions on the test set
y_pred = classifier.predict(X_test)
# Evaluate the model
score = classifier.score(X_test, y_test)
report = classification_report(y_test, y_pred)
print("Accuracy:", score)
print("Classification Report:")
print(report)
```

Output:

Accuracy: 0.5581395348837209 Classification Report: precision recall f1-score support 0.41 0.64 0.43 0.21 1 0.50 11 2 0.29 14 3 0.40 0.67 0.50 3 0.50 0.25 0.33 1.00 1.00 1.00 5 4 6 3 1.00 0.94 0.89 8 0.56 43 accuracy 0.60 0.63 0.59 43 macro avg 0.56 0.53 43 weighted avg 0.55

2: Implement linear SVM method using scikit-learn

Use the same dataset above

Use train_test_split to create training and testing part Evaluate the model on test part using score and classification_report(y_true, y_pred)

```
In [10]: import pandas as pd
 from sklearn.model selection import train test split
 from sklearn.svm import LinearSVC
 from sklearn.metrics import classification_report
 # Load the dataset from the CSV file
 glass_data = pd.read_csv("glass.csv")
 # Split the dataset into features (X) and target (y)
 X = glass_data.drop('Type', axis=1)
 y = glass_data['Type']
 # Split the dataset into training and testing parts
 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
 # Create and train the Linear SVM classifier
 classifier = LinearSVC(max_iter=10000)
 classifier.fit(X_train, y_train)
 # Make predictions on the test set
 y_pred = classifier.predict(X_test)
 # Evaluate the model
 score = classifier.score(X_test, y_test)
 report = classification_report(y_test, y_pred, zero_division=0)
 print("Accuracy:", score)
 print("Classification Report:")
 print(report)
```

Output:

Accuracy: 0.5 Classification			f1-score	support	
	•			• •	
1	0.56	0.82	0.67	11	
2	0.50	0.50	0.50	14	
3	0.00	0.00	0.00	3	
5	1.00	0.25	0.40	4	
6	0.00	0.00	0.00	3	
7	0.67	1.00	0.80	8	
accuracy			0.58	43	
accuracy	0.45	0.43	0.39	43	
macro avg					
weighted avg	0.52	0.58	0.52	43	

Which algorithm you got better accuracy? Can you justify why?

To determine which algorithm performs better on a specific dataset, it is essential to evaluate their performance using appropriate metrics such as accuracy, precision, recall, and F1-score. These metrics help to assess the model's ability to correctly classify instances from different classes and handle class imbalances. So based on the accuracy I got linear SVM method using scikit-learn algorithm got the more accuracy.

3. Implement Linear Regression using scikit-learn a) Import the given "Salary_Data.csv" b) Split the data in train_test partitions, such that 1/3 of the data is reserved as test subset. c) Train and predict the model. d) Calculate the mean_squared error. e) Visualize both train and test data using scatter plot.

```
In [14]: import pandas as pd
 import matplotlib.pyplot as plt
 from sklearn.model_selection import train_test_split
 from sklearn.linear model import LinearRegression
 from sklearn.metrics import mean squared error
 # Import the dataset from "Salary Data.csv"
 data = pd.read csv("Salary Data.csv")
 # Split the data into features (X) and target (y)
 X = data["YearsExperience"].values.reshape(-1, 1)
 y = data["Salary"].values
 # Split the data into training and testing sets
 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.33, random_state=42)
 # Create and train the Linear Regression model
 model = LinearRegression()
 model.fit(X train, y train)
 # Predict on the test set
 y pred = model.predict(X test)
 # Calculate the mean squared error
 mse = mean_squared_error(y_test, y_pred)
 print("Mean Squared Error:", mse)
 # Visualize the training and test data
 plt.scatter(X_train, y_train, color='blue', label='Training data')
 plt.scatter(X_test, y_test, color='red', label='Test data')
 plt.plot(X_test, y_pred, color='green', linewidth=2, label='Linear Regression')
 plt.xlabel('Years of Experience')
 plt.ylabel('Salary')
 plt.title('Linear Regression - Salary Prediction')
 plt.legend()
 plt.show()
```

Output:

Mean Squared Error: 35301898.88713492

