EXPT NO: 1 A python program to implement univariate regression

DATE: bivariate regression and multivariate regression.

AIM:

To write a python program to implement univariate regression, bivariate regression and multivariate regression.

PROCEDURE:

Implementing univariate, bivariate, and multivariate regression using the Iris dataset involve the following steps:

Step 1: Import Necessary Libraries

First, import the libraries that are essential for data manipulation, visualization, and model building.

```
import numpy as np
import pandas as pd
import seaborn as sns
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, r2_score
```

Step 2: Load the Iris Dataset

The Iris dataset can be loaded and display the first few rows of the dataset.

```
# Load the Iris dataset
```

```
iris = sns.load_dataset('iris')

# Display the first few rows of the dataset

print(iris.head())
```

OUTPUT:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

Step 3: Data Preprocessing

Ensure the data is clean and ready for modeling. Since the Iris dataset is clean, minimal preprocessing is needed.

```
# Check for missing values
print(iris.isnull().sum())

# Display the basic statistical details
print(iris.describe())
```

OUTPUT:

```
→ sepal length
    sepal width
    petal length
    petal width
                  0
    species
    dtype: int64
          sepal_length sepal_width petal_length petal_width
    count
            150.000000 150.000000
                                     150.000000 150.000000
              5.843333
                          3.057333
                                       3.758000
                                                  1.199333
    mean
    std
            0.828066
                          0.435866
                                      1.765298
                                                  0.762238
    min
            4.300000
                          2.000000
                                      1.000000
                                                  0.100000
                                      1.600000
    25%
             5.100000
                          2.800000
                                                  0.300000
             5.800000
    50%
                         3.000000
                                      4.350000
                                                 1.300000
    75%
              6.400000
                                      5.100000
                          3.300000
                                                 1.800000
                          4.400000
             7.900000
    max
                                     6.900000
                                                  2.500000
```

Step 4: Univariate Regression

Univariate regression involves predicting one variable based on a single predictor.

4.1: Select the Features

Choose one feature (e.g., sepal_length) and one target variable (e.g., sepal_width).

```
X_uni = iris[['sepal_length']]
y_uni = iris['sepal_width']
```

4.2: Split the Data

Split the data into training and testing sets.

Fit the linear regression model on the training data.

```
X_uni_train, X_uni_test, y_uni_train, y_uni_test = train_test_split(X_uni,
y_uni,
test_size=0.2, random_state=42)
```

4.3: Train the model

```
uni_model = LinearRegression()
uni_model.fit(X_uni_train, y_uni_train)
```



```
LinearRegression  LinearRegression()
```

4.4: Make Predictions

Use the model to make predictions on the test data.

```
y_uni_pred = uni_model.predict(X_uni_test)
```

4.5: Evaluate the Model

Evaluate the model performance using metrics like Mean Squared Error (MSE) and R-squared.

```
print(f'Univariate MSE: {mean squared error(y uni test, y uni pred)}')
print(f'Univariate R-squared: {r2_score(y_uni_test, y_uni_pred)}')
```

OUTPUT:



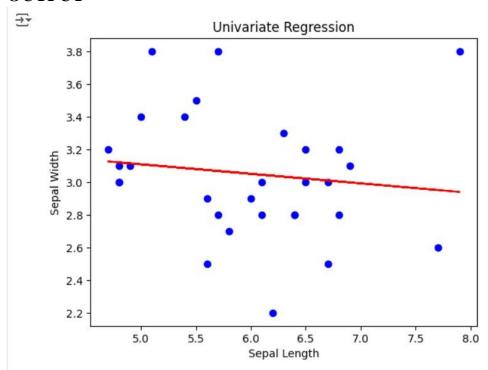
Tunivariate MSE: 0.13961895650579023 Univariate R-squared: 0.024098626473972984

4.6: Visualize the Results

Visualize the relationship between the predictor and the target variable.

```
plt.scatter(X_uni_test, y_uni_test, color='blue')
plt.plot(X uni test, y uni pred, color='red')
plt.xlabel('Sepal Length')
plt.ylabel('Sepal Width')
plt.title('Univariate Regression')
plt.show()
```

OUTPUT



Step 5: Bivariate Regression

Bivariate regression involves predicting one variable based on two predictors.

5.1: Select the Features

Choose two features (e.g., sepal_length, petal_length) and one target variable (e.g., sepal_width).

```
X_bi = iris[['sepal_length', 'petal_length']]

y_bi = iris['sepal_width']
```

5.2: Split the Data

Split the data into training and testing sets.

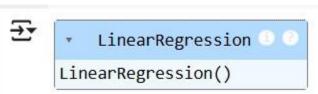
```
X_bi_train, X_bi_test, y_bi_train, y_bi_test = train_test_split(X_bi, y_bi,
test_size=0.2, random_state=42)
```

5.3: Train the Model

Fit the linear regression model on the training data.

```
bi_model = LinearRegression()
bi_model.fit(X_bi_train, y_bi_train)
```

OUTPUT:



5.4: Make Predictions

Use the model to make predictions on the test data.

```
y_bi_pred = bi_model.predict(X_bi_test)
```

5.5: Evaluate the Model

Evaluate the model performance using metrics like MSE and R-squared.

```
print(f'Bivariate MSE: {mean_squared_error(y_bi_test, y_bi_pred)}')
print(f'Bivariate R-squared: {r2_score(y_bi_test, y_bi_pred)}')
```

OUTPUT:

```
Bivariate MSE: 0.08308605032913309
Bivariate R-squared: 0.4192494152204116
```

5.6: Visualize the Results

Since visualizing in 3D is challenging, we can plot the relationships between the target and each predictor separately.

```
# Sepal Length vs Sepal Width
plt.subplot(1, 2, 1)
plt.scatter(X_bi_test['sepal_length'], y_bi_test, color='blue')
plt.plot(X_bi_test['sepal_length'], y_bi_pred, color='red')
```

```
plt.xlabel('Sepal Length')

plt.ylabel('Sepal Width')

# Petal Length vs Sepal Width

plt.subplot(1, 2, 2)

plt.scatter(X_bi_test['petal_length'], y_bi_test, color='blue')

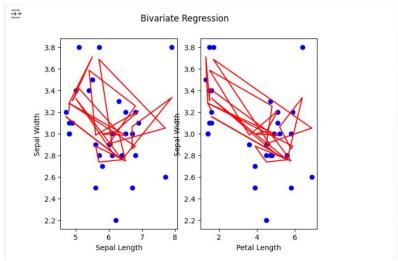
plt.plot(X_bi_test['petal_length'], y_bi_pred, color='red')

plt.xlabel('Petal Length')

plt.ylabel('Sepal Width')

plt.suptitle('Bivariate Regression')
```

OUTPUT:



Step 6: Multivariate Regression

Multivariate regression involves predicting one variable based on multiple predictors.

6.1: Select the Features

Choose multiple features (e.g., sepal_length, petal_length, petal_width) and one target variable (e.g., sepal_width).

```
X_multi = iris[['sepal_length', 'petal_length', 'petal_width']]
y_multi = iris['sepal_width']
```

6.2: Split the Data

Split the data into training and testing sets.

```
X_multi_train, X_multi_test, y_multi_train, y_multi_test =
train_test_split(X_multi,

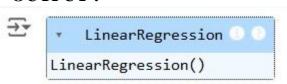
y_multi, test_size=0.2, random_state=42)
```

6.3: Train the Model

Fit the linear regression model on the training data.

```
multi_model = LinearRegression()
multi_model.fit(X_multi_train, y_multi_train)
```

OUTPUT:



6.4: Make Predictions

Use the model to make predictions on the test data.

```
y_multi_pred = multi_model.predict(X_multi_test)
```

6.5: Evaluate the Model

Evaluate the model performance using metrics like MSE and R-squared.

```
print(f'Multivariate MSE: {mean_squared_error(y_multi_test, y_multi_pred)}')
print(f'Multivariate R-squared: {r2_score(y_multi_test, y_multi_pred)}')
```

OUTPUT:

Step 7: Visualize the multivariate regression

```
plt.figure(figsize=(15,4))
plt.subplot(1, 2, 1)
plt.scatter(X_multi_test['sepal_length'], y_multi_test, color='blue')
plt.plot(X multi test['sepal length'], y multi pred, color='red')
plt.xlabel('sepal length')
plt.ylabel('sepal_width')
plt.title('Multivariate Regression-1')
plt.show()
plt.figure(figsize=(15,4))
plt.subplot(1, 2, 1)
plt.scatter(X multi test['petal length'], y multi test, color='blue')
plt.plot(X multi test['petal length'], y multi pred, color='red')
plt.xlabel('petal length')
plt.ylabel('sepal width')
plt.title('Multivariate Regression-2')
```

```
plt.show()
plt.figure(figsize=(15,4))

plt.subplot(1, 2, 2 )

plt.scatter(X_multi_test['petal_length'], y_multi_test, color='blue')

plt.plot(X_multi_test['petal_length'], y_multi_pred, color='red')

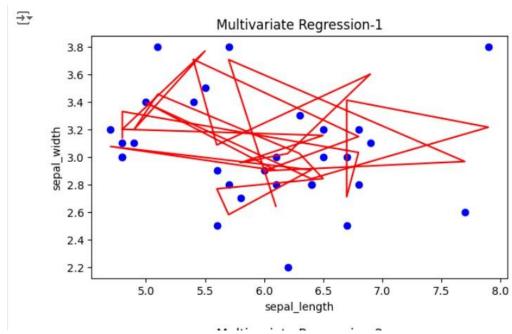
plt.xlabel('petal_length')

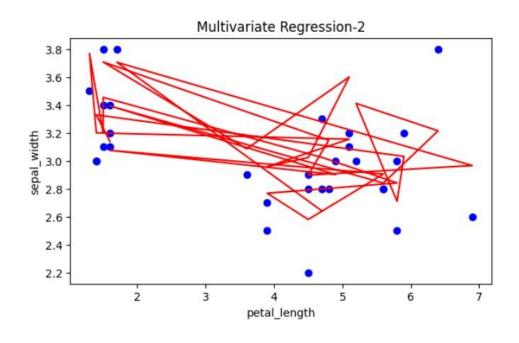
plt.ylabel('sepal_width')

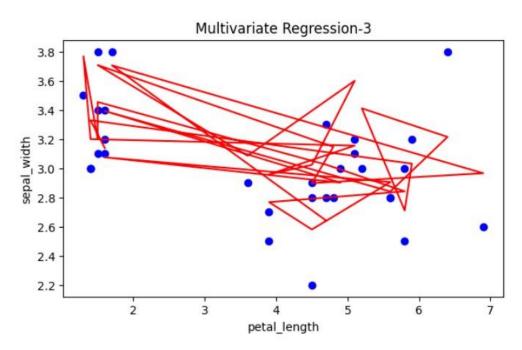
plt.title('Multivariate Regression-3')

plt.show()
```

OUTPUT:







Step 8: Interpret the Results

After implementing and evaluating the models, interpret the coefficients to understand the influence of each predictor on the target variable.

```
print('Univariate Coefficients:', uni_model.coef_)
print('Bivariate Coefficients:', bi_model.coef_)
print('Multivariate Coefficients:', multi_model.coef_)
```

OUTPUT:

```
Univariate Coefficients: [-0.05829418]

Bivariate Coefficients: [ 0.56420418 -0.33942806]

Multivariate Coefficients: [ 0.62934965 -0.63196673  0.6440201 ]
```

RESULT:

This step-by-step process will help us to implement univariate, bivariate, and multivariate regression models using the Iris dataset and analyze their performance.