## Program No:- 7

Aim:- Program to implement multiple regression technique using any standard dataset available in the public domain and evaluate its performance.

Program-1

```
Program Code
import pandas
df = pandas.read_csv("cars.csv")
x = df[['Weight', 'Volume']]
y = df['CO2']
#splitting
from sklearn.model_selection import train_test_split
x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.20,
random state=0)
# feature scaling
from sklearn.preprocessing import StandardScaler
sc = StandardScaler()
x_train = sc.fit_transform(x_train)
x test = sc.transform(x test)
print(x_train)
print(x_test)
from sklearn import linear_model
regr = linear_model.LinearRegression()
regr.fit(x,y)
predictedCO2 = regr.predict([[2300, 1300]])
```

```
print(predictedCO2)
from sklearn.metrics import accuracy_score
ac=accuracy_score(y_test,y_test)
print(ac)
```

## **Output**

Program-2

print(predictedCO2)

```
Program Code

import pandas

df = pandas.read_csv("cars.csv")

X = df[['Weight', 'Volume']]

y = df['CO2']

from sklearn import linear_model

regr = linear_model.LinearRegression()

regr.fit(X,y)

predictedCO2 = regr.predict([[2300, 1300]])
```

## **Output**

C:\Users\mca\PycharmProjectspython\data\venv\Scripts\python.exe C:\Users\mca\PycharmProjectspython\pythonProject1/mullinear.py
[107.2087328]

```
Program-3
Aim:-Variance and Coefficient
Program Code
import matplotlib.pyplot as plt
#import numpy as np
from sklearn import datasets, linear_model, metrics
boston = datasets.load boston(return X y=False)
X = boston.data
y = boston.target
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.4,
random state=1)
reg = linear model.LinearRegression()
reg.fit(X_train, y_train)
print('Coefficients: ', reg.coef_)
print('Variance score: { }'.format(reg.score(X_test, y_test)))
```

## **Output**

```
C:\Users\mca\PycharmProjectspython\data\venv\Scripts\python.exe C:/Users/mca/PycharmProjectspython/pythonProject1/mul2.py
Coefficients: [-8.95714048e-02 6.73132853e-02 5.04649248e-02 2.18579583e+00
-1.72053975e+01 3.63606995e+00 2.05579939e-03 -1.36602886e+00
2.89576718e-01 -1.22700072e-02 -8.34881849e-01 9.40360790e-03
-5.04008320e-01]
Variance score: 0.7209056672661767
```