UNIVERSITY OF ASIA PACIFIC

Department of Computer Science of Engineering



Project-2:

Lab Report

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Objective: In this project the main idea is to implement the multivariable linear regression without SK-Learn and with SK-learn. In this project try to find out accurate predict result for using selected dataset. And also find out Cost, Hypothesis Function value and Updated parameters.

About(Multivariable Linear Regression): Multivariable Linear Regression also known simply as multiple regression is a statistical technique that uses several explanatory variables to predict the outcome of a response variable. The goal of Multivariable Linear Regression is to model the linear relationship between the explanatory variables and response variable.

Formula and calculation of Multivariable Linear Regression

$$y = t0 + t1x1 + t2x2 + + tpxp$$

Where,

y= dependent variable

xp= explanatory variable

t0= y-intercept

tp= slope coefficients for each explanatory variable

Multivariate Linear Regression

Hypothesis:
$$h_{\theta}(x) = \theta^T x = \theta_0 x_0 + \theta_1 x_1 + \theta_2 x_2 + \dots + \theta_n x_n$$

Parameters: $\theta_0, \theta_1, \dots, \theta_n$

Cost function:

function:

$$J(\theta_0, \theta_1, \dots, \theta_n) = \frac{1}{2m} \sum_{i=1}^m (h_{\theta}(x^{(i)}) - y^{(i)})^2$$

Gradient descent:

Repeat
$$\{$$
 $heta_j:= heta_j-lpha rac{\partial}{\partial heta_j}J(heta_0,\dots, heta_n)$ $\}$ (simultaneously update for every $j=0,\dots,n$)

<u>About(Scikit-Learn)</u>: Scikit-Learn is a Python package that simplifies the implementation of a wide range of Machine Learning(ML) methods for predictive data analysis, including linear regression.

The following are some key concepts we will come across when we work with scikit-learn linear regression methods here some method we used:

- -> **Coefficient** also known as a parameter, is the factor a variable is multiplied by. In linear regression, a coefficient represents changes in a Response Variable.
- -> **Intercept** the location where the Slope intercepts the Y-axis denoted b in the slope equation y=ax+b.
- -> **Mean** an average of a set of numbers, but in linear regression, Mean is modeled by a linear function.
- -> **Regression Model** the ideal formula for approximating a regression.

<u>Linear Regression Class Definition:</u> A scikit-learn linear regression script begins by importing the Linear Regression class

from sklearn.linear_model import LinearRegression
sklearn.linear model.LinearRegression().

<u>Dataset:</u> In this project i used a data form kaggle the dataset name are, Name: 'Petrol Consumption', In this dataset 5 column these are,

'Petrol_tax', 'Average_income', 'Paved_Highways',

'Population_Driver_Licence(%)','Petrol_Consumption'

Link->https://www.kaggle.com/anikchandra70/petrol-consumption-multivariable-linear-regression

Implementation:

For implementation I used Python programming language, Excel dataset, Scikit-learn and Colab notebooks.

In this part I discuss some points of my implementation and code.

First I discuss how to read an excel dataset in Colab. In this case i use python pandas libraries

Implement Multivariable Linear Regression without Scikit-Learn->

First import pandas libraries and read the file "pd.read_csv()".and put the excel content path in this function.

- # import data and view data
 df=pd.read_csv('/content/petrol_consumption.csv') # A simple way to store big
 df.head() # Return the first 5 rows of the DataFrame.
- ₽ Petrol_tax Average_income Paved_Highways Population_Driver_licence(%) Petrol_Consumption 0 9.0 3571 1976 0.525 541 9.0 1 4092 1250 0.572 524 2 9.0 3865 1586 0.580 561 7.5 4870 2351 0.529 414 0.544 8.0 4399 431 410

Then I rename() the data frame ['Petrol_tax' : 'x1',....] for further calculation.

```
1  # rename() Rename the row,column indexes of the DataFrame
2  df = df.rename(columns={'Petrol_tax': 'x1', 'Average_income': 'x2', 'Paved_Highways'
3  df.head() # Return the first 5 rows of the DataFrame.

x1  x2  x3  x4  y

0  9.0  3571  1976  0.525  541

1  9.0  4092  1250  0.572  524

2  9.0  3865  1586  0.580  561

3  7.5  4870  2351  0.529  414

4  8.0  4399  431  0.544  410
```

Then I define a **predict()** function for prediction.According to Multivariable Linear Regression formula

```
[] 1 # Define Predict function
2 # initialize t0-t4 for value of theta or Coefficient
3 # equation y = t0 + t1 * x1 + t2 * x2 + t3 * x3 + t4 * x4
4 def predict(t0,t1,t2,t3,t4,x1,x2,x3,x4):
5 | return t0 + t1 * x1 + t2 * x2 + t3 * x3 + t4 * x4
```

Then I define a **computeCost()** function for cost,hypothesis,parameter according to formula.

```
1 # Define computeCost function
 2 \rightarrow def computeCost(t0,t1,t2,t3,t4,x1,x2,x3,x4,y):
        # Getting number of data
 3
 4
        m = float(len(y))
 5
        loss = [] # initialize list for stroe
 6
        # Iterating over all of the data
 7 \
        for i in range(len(y)):
            # Getting prediction using the parameter [t0, t1, t2 ,t3,4]
 8
 9
            h = predict(t0,t1,t2,t3,t4,x1[i],x2[i],x3[i],x4[i])
10
            # Adding the losses to the list
            loss.append((h - y[i])**2)
11
        # Printing Hypothesis Function Value after each epoch
12
13
        print("Hypothesis Function Value is: ",loss)
        # Printing Updated parameters after each epoch
14
        print("Updated parameters are: ",t0,t1,t2,t3,t4)
15
        return (sum(loss) / (2 * m))
16
```

Then i apply **epoch = 30** for cost,hypothesis and parameter value using batch gradient descent formula.and finally predict result

```
1 # prediction result | 2 predict(425.599332,-40.016660,-0.065413,-0.004741,1341.862121,9.0,3571,1976,0.525) | 526.9689665249999
```

Implement Multivariable Linear Regression with Scikit-Learn->

First import pandas libraries and read the file "pd.read_csv()".and put the excel content path in this function.



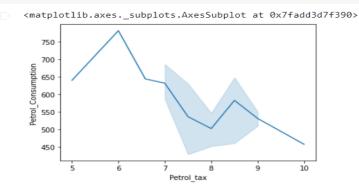
Then show the basic statistical details like count, mean, std etc using panda describe() method

0	<pre># Pandas describe() is used to view some basic statistical details like percentile, df.describe()</pre>
□→	Detail to Assess in the Development Develo

	Petrol_tax	Average_income	Paved_Highways	Population_Driver_licence(%)	Petrol_Consumption
count	48.000000	48.000000	48.000000	48.000000	48.000000
mean	7.668333	4241.833333	5565.416667	0.570333	576.770833
std	0.950770	573.623768	3491.507166	0.055470	111.885816
min	5.000000	3063.000000	431.000000	0.451000	344.000000
25%	7.000000	3739.000000	3110.250000	0.529750	509.500000
50%	7.500000	4298.000000	4735.500000	0.564500	568.500000
75%	8.125000	4578.750000	7156.000000	0.595250	632.750000
max	10.000000	5342.000000	17782.000000	0.724000	968.000000

Then draw a line plot **Petrol_tax vs Pertrol_Consumption** for data visualization in this case use **Seaborn packages** for 3d visualization.

```
1 # Draw a line plot with possibility of several semantic
2 sns.lineplot(x=df['Petrol_tax'],y=df['Petrol_Consumption'])
```



For model train purpose we creating an attribute set and label

```
# We are going to use column names for creating an attribute set and label.

X=df[['Petrol_tax','Average_income','Paved_Highways','Population_Driver_licence(%)']]

y=df['Petrol_Consumption']
```

Then dividing the data into training and test purpose

```
1  # Dividing our data into training and test sets
2  from sklearn.model_selection import train_test_split
3  X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
```

Then training the model use scikit-learn **LinearRegression()** method and fit the model

```
1  # Training the Algorithm
2  from sklearn.linear_model import LinearRegression
3  reg=LinearRegression()
4  reg.fit(X_train,y_train)
```

Using scikit-learn and find out Coefficient and Intercept

```
1 # In case of multivariable linear regression, the regression model has to find the
2 coeff_df = pd.DataFrame(reg.coef_, X.columns, columns=['Coefficient'])
3 coeff_df
```

	Coefficient
Petrol_tax	-40.016660
Average_income	-0.065413
Paved_Highways	-0.004741
Population Driver licence(%)	1341 862121

```
1 # The regression model has to find the most optimal intercept for all the attributes.
2 int_df = pd.DataFrame(reg.intercept_, X.columns, columns=['intercept'])
3 int_df
```

	intercept
Petrol_tax	425.599332
Average_income	425.599332
Paved_Highways	425.599332
Population_Driver_licence(%)	425.599332

Then making prediction and predict the result

```
[] 1 # Making Predictions
2 y_pred=reg.predict(X_test)

[] 1 # Predictions
2 y_pred=reg.predict([[9.0,3571,1976,0.525]])
3 y_pred
array([526.97067878])
```

Some visualization on the report part we discuss all visualization in code .

Result Analysis:

Formula and calculation of Multivariable Linear Regression

$$y = t0 + t1x1 + t2x2 + + tpxp$$

So in this case we found value from code result

Petrol_Consumption-> y = ?

Intercept-> t0= 425.599332

Petrol tax -> t1= -40.016660

Average income-> t2= -0.065413

Paved_Highways-> t3= -0.004741

Population_Driver_Licence(%)-> t4= 1341.862121

We assume the x1,x2,x3 and x4 form dataset these are

x1 = 9.0

x2 = 3571

x3= 1976

x4 = 0.525

```
y= t0 + t1x1 + t2x2 + t3x3 + t4x4
y= 425.599332 + 9.0*(-40.016660) + 3571*(-0.065413) +
1976*(-0.004741) + 0.525*1341.862121
y= 526.9689665249999
```

Both with Scikit-learn and without Scikit-learn prediction result and manually calculation result are the same.

Conclusion:

In this project I implement Multivariable Linear Regression with Scikit-learn and without Scikit-learn. And also visualize the necessary parameters and data using 'Seaborn packages', and Predict Petrol_Consumption use some data me already discuss and calculation the result. Both with Scikit-learn and without Scikit-learn prediction result and manually calculation result are the same. In this project my implementation result and calculation result are the same so I hopefully say that this the Petrol_Consumption prediction model is right.