



Artificial Intelligence and Expert System Lab (CSE 404)

Department of CSE

Project: 02

Topic/Question: Multivariable Liner Regression Using Open Source Dataset  
(data\_monthly\_rainfall).

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Submitted by	Submitted to
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**Tools:**

1. Language: Python 3 (for coding)
2. IDE: jupyter notebook (text editor)
3. Google Colab

My dataset title is: Rainfall data of Bangladesh(1970-2016)

The dataset link: <https://www.kaggle.com/redikod/historical-rainfall-data-in-bangladesh>

My task is to implement the Multivariable Liner Regression Using Open Source Dataset without SK-Learn.

Here the model trained by raw coding in python 3.

I haven't used any library function to train my model.

But only to normalize the dataset, I used a normalized library function from Sk-learn.

**Now, Coding: (using Google Colab)**

the first block of code is to upload the dataset.

```
1 from google.colab import files
2 #uploading dataset
3 uploaded = files.upload()
4
5 for fn in uploaded.keys():
6     print('User uploaded file "{name}" with length {length} bytes'.format(
7         name=fn, length=len(uploaded[fn])))
```

... Choose Files data\_monthly\_rainfall.csv

- data\_monthly\_rainfall.csv(application/vnd.ms-excel) - 380394 bytes, last modified: 3/17/2021 - 81% done

Saving data\_monthly\_rainfall.csv to data\_monthly\_rainfall.csv

Importin some library function.

```
1 import numpy as np # Numpy Header
2 import pandas as pd #pandas Header
3 import random
```

read the dataset as pandas object.

```
df = pd.read_csv('data_monthly_rainfall.csv') #read dataset
```

checking the data.

```
1 print(df.tail())
```

	Year	Station	Month	Rainfall	StationIndex
16750	2016	Teknaf	8	920	33
16751	2016	Teknaf	9	512	33
16752	2016	Teknaf	10	208	33
16753	2016	Teknaf	11	53	33
16754	2016	Teknaf	12	0	33

Here 'm' is the length of dataset.

Initializing the features Xs and label data Y.

```
x1=df["Year"]
x2=df["StationIndex"]
x3=df["Month"]
y= df["Rainfall"]
print(x1.tail())
print(x2.head())
print(x3.head())
print(y.head())
len(x1)
```

Preprocessing of dataset/ Normalization of dataset.

Normalizing the dataset between -1 and 1

Code Block 9.

Setting the value of learning rate (alpha)

Here, as we know that the dataset is huge. To make it time efficient we are using the value of alpha 0.2.

But the perfect way to do this: take the lowest value of all features and divide it by 100 then take the value. But by this the optimization process will take more than day in my computer.

So, I had to make it 0.2.

```
#learning rate
#alpha = min(min(x1),min(x2),min(x3))/100
alpha=0.2
print(alpha)
```

Now the Code block 10 is the optimization process.  
where we are initializing the values of coefficients (theta)

Then following the step:

1. Hypothesis function (to calculate hypothetical values)
2. Calculating cost
3. Gradient decent (changing the value of coefficients)

```
J = 100
n = 0 # iteration number
theta = [1,1.5, 8,9]
```

```
for i in range(1000):

    print("Iteration number: ",n+1)
    n = n+1
    # hypothesis function
    h = []
    print("Hypothesis function value is: h0(x)=theta_0+theta_1 * x")
    for i2 in range(m):
        temp = theta[0] + theta[1]*x1[i2] + theta[2]*x2[i2]+theta[3]*x3[i2]
        h.append(temp)

    # cost function
    error_sum = 0
    print("Cost function is: j(theta)=1/(2*m) * i=1_samtionSign_m (h_theta_(x)-y)**2")

    for i3 in range(m):
        error_sum = error_sum + (h[i3] - y[i3])**2

    J = (1/(2*m))*error_sum
    if J == float("inf") :
        theta = [random.randint(-100,100),random.randint(-100,100), random.randint(-100,100),random.randint(-100,100)]
        continue
    print("\nCost function is:",J)
```

```

    # gradient descent
print("\ngradient decent:")
temp0 = 0
for i4 in range(m):
    temp0 = temp0 + (h[i4] - y[i4])

theta[0] = theta[0] - (alpha/m)*temp0

temp1 = 0
for i5 in range(m):
    temp1 = temp1 + (h[i5] - y[i5])*x1[i5]

theta[1] = theta[1] - (alpha/m)*temp1

temp1 = 0
for i5 in range(m):
    temp1 = temp1 + (h[i5] - y[i5])*x2[i5]

theta[2] = theta[2] - (alpha/m)*temp1

temp1 = 0
for i5 in range(m):
    temp1 = temp1 + (h[i5] - y[i5])*x3[i5]

theta[3] = theta[3] - (alpha/m)*temp1

print("New parameter value is: ",theta)

print("result coefficient is",thetaResult)

```

Now after optimization we can predict the result of given inputs:

```
y=int (input("year :"))
m=int (input("month :"))
d=int (input("district :"))

#prediction
rainfall=theta[0]+theta[1]*y +theta[2]* d +theta[3] *m
print(rainfall)
```

The model works fine.