

DSBDA Practical No.04

May 19, 2023

```
[2]: # Importing Necessary Libraries
%matplotlib inline
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
plt.rcParams['figure.figsize'] = (20.0, 10.0)
```

```
[4]: # Reading Data
data = pd.read_csv('/Users/shreyaspeherkar/Desktop/Dataset/HousingData.csv')
print(data.shape)
data.head()
```

(506, 14)

```
[4]:      CRIM      ZN  INDUS  CHAS    NOX     RM   AGE     DIS  RAD  TAX  PTRATIO  \
0  0.00632  18.0    2.31    0.0  0.538  6.575  65.2  4.0900    1  296    15.3
1  0.02731   0.0    7.07    0.0  0.469  6.421  78.9  4.9671    2  242    17.8
2  0.02729   0.0    7.07    0.0  0.469  7.185  61.1  4.9671    2  242    17.8
3  0.03237   0.0    2.18    0.0  0.458  6.998  45.8  6.0622    3  222    18.7
4  0.06905   0.0    2.18    0.0  0.458  7.147  54.2  6.0622    3  222    18.7

      B  LSTAT  MEDV
0  396.90   4.98  24.0
1  396.90   9.14  21.6
2  392.83   4.03  34.7
3  394.63   2.94  33.4
4  396.90   NaN  36.2
```

```
[5]: # Collecting X and Y
X = data['DIS'].values
Y = data['MEDV'].values
```

```
[7]: Y
#Y=mX+b m= difference in y coordinate/difference in x coordinate b= y-intercept
```

```
[7]: array([24. , 21.6, 34.7, 33.4, 36.2, 28.7, 22.9, 27.1, 16.5, 18.9, 15. ,
        18.9, 21.7, 20.4, 18.2, 19.9, 23.1, 17.5, 20.2, 18.2, 13.6, 19.6,
        15.2, 14.5, 15.6, 13.9, 16.6, 14.8, 18.4, 21. , 12.7, 14.5, 13.2,
```

13.1, 13.5, 18.9, 20. , 21. , 24.7, 30.8, 34.9, 26.6, 25.3, 24.7,
 21.2, 19.3, 20. , 16.6, 14.4, 19.4, 19.7, 20.5, 25. , 23.4, 18.9,
 35.4, 24.7, 31.6, 23.3, 19.6, 18.7, 16. , 22.2, 25. , 33. , 23.5,
 19.4, 22. , 17.4, 20.9, 24.2, 21.7, 22.8, 23.4, 24.1, 21.4, 20. ,
 20.8, 21.2, 20.3, 28. , 23.9, 24.8, 22.9, 23.9, 26.6, 22.5, 22.2,
 23.6, 28.7, 22.6, 22. , 22.9, 25. , 20.6, 28.4, 21.4, 38.7, 43.8,
 33.2, 27.5, 26.5, 18.6, 19.3, 20.1, 19.5, 19.5, 20.4, 19.8, 19.4,
 21.7, 22.8, 18.8, 18.7, 18.5, 18.3, 21.2, 19.2, 20.4, 19.3, 22. ,
 20.3, 20.5, 17.3, 18.8, 21.4, 15.7, 16.2, 18. , 14.3, 19.2, 19.6,
 23. , 18.4, 15.6, 18.1, 17.4, 17.1, 13.3, 17.8, 14. , 14.4, 13.4,
 15.6, 11.8, 13.8, 15.6, 14.6, 17.8, 15.4, 21.5, 19.6, 15.3, 19.4,
 17. , 15.6, 13.1, 41.3, 24.3, 23.3, 27. , 50. , 50. , 50. , 22.7,
 25. , 50. , 23.8, 23.8, 22.3, 17.4, 19.1, 23.1, 23.6, 22.6, 29.4,
 23.2, 24.6, 29.9, 37.2, 39.8, 36.2, 37.9, 32.5, 26.4, 29.6, 50. ,
 32. , 29.8, 34.9, 37. , 30.5, 36.4, 31.1, 29.1, 50. , 33.3, 30.3,
 34.6, 34.9, 32.9, 24.1, 42.3, 48.5, 50. , 22.6, 24.4, 22.5, 24.4,
 20. , 21.7, 19.3, 22.4, 28.1, 23.7, 25. , 23.3, 28.7, 21.5, 23. ,
 26.7, 21.7, 27.5, 30.1, 44.8, 50. , 37.6, 31.6, 46.7, 31.5, 24.3,
 31.7, 41.7, 48.3, 29. , 24. , 25.1, 31.5, 23.7, 23.3, 22. , 20.1,
 22.2, 23.7, 17.6, 18.5, 24.3, 20.5, 24.5, 26.2, 24.4, 24.8, 29.6,
 42.8, 21.9, 20.9, 44. , 50. , 36. , 30.1, 33.8, 43.1, 48.8, 31. ,
 36.5, 22.8, 30.7, 50. , 43.5, 20.7, 21.1, 25.2, 24.4, 35.2, 32.4,
 32. , 33.2, 33.1, 29.1, 35.1, 45.4, 35.4, 46. , 50. , 32.2, 22. ,
 20.1, 23.2, 22.3, 24.8, 28.5, 37.3, 27.9, 23.9, 21.7, 28.6, 27.1,
 20.3, 22.5, 29. , 24.8, 22. , 26.4, 33.1, 36.1, 28.4, 33.4, 28.2,
 22.8, 20.3, 16.1, 22.1, 19.4, 21.6, 23.8, 16.2, 17.8, 19.8, 23.1,
 21. , 23.8, 23.1, 20.4, 18.5, 25. , 24.6, 23. , 22.2, 19.3, 22.6,
 19.8, 17.1, 19.4, 22.2, 20.7, 21.1, 19.5, 18.5, 20.6, 19. , 18.7,
 32.7, 16.5, 23.9, 31.2, 17.5, 17.2, 23.1, 24.5, 26.6, 22.9, 24.1,
 18.6, 30.1, 18.2, 20.6, 17.8, 21.7, 22.7, 22.6, 25. , 19.9, 20.8,
 16.8, 21.9, 27.5, 21.9, 23.1, 50. , 50. , 50. , 50. , 50. , 13.8,
 13.8, 15. , 13.9, 13.3, 13.1, 10.2, 10.4, 10.9, 11.3, 12.3, 8.8,
 7.2, 10.5, 7.4, 10.2, 11.5, 15.1, 23.2, 9.7, 13.8, 12.7, 13.1,
 12.5, 8.5, 5. , 6.3, 5.6, 7.2, 12.1, 8.3, 8.5, 5. , 11.9,
 27.9, 17.2, 27.5, 15. , 17.2, 17.9, 16.3, 7. , 7.2, 7.5, 10.4,
 8.8, 8.4, 16.7, 14.2, 20.8, 13.4, 11.7, 8.3, 10.2, 10.9, 11. ,
 9.5, 14.5, 14.1, 16.1, 14.3, 11.7, 13.4, 9.6, 8.7, 8.4, 12.8,
 10.5, 17.1, 18.4, 15.4, 10.8, 11.8, 14.9, 12.6, 14.1, 13. , 13.4,
 15.2, 16.1, 17.8, 14.9, 14.1, 12.7, 13.5, 14.9, 20. , 16.4, 17.7,
 19.5, 20.2, 21.4, 19.9, 19. , 19.1, 19.1, 20.1, 19.9, 19.6, 23.2,
 29.8, 13.8, 13.3, 16.7, 12. , 14.6, 21.4, 23. , 23.7, 25. , 21.8,
 20.6, 21.2, 19.1, 20.6, 15.2, 7. , 8.1, 13.6, 20.1, 21.8, 24.5,
 23.1, 19.7, 18.3, 21.2, 17.5, 16.8, 22.4, 20.6, 23.9, 22. , 11.9])

```
[8]: # Calculating coefficient
      # Mean X and Y
      mean_x = np.mean(X)
```

```
mean_y = np.mean(Y)
# Total number of values
n = len(X)
```

[9]: n

[9]: 506

```
[10]: # Using the formula to calculate b1 and b2
number = 0
denom = 0
for i in range(n):
    number += (X[i] - mean_x) * (Y[i] - mean_y)
    denom += (X[i] - mean_x) ** 2
    b1 = number / denom
    b0 = mean_y - (b1 * mean_x)
# m(b1) and c(b0)
# Printing coefficients
print("Coefficients")
print("m=", b1)
print("c=", b0)
```

Coefficients

m= 1.0916130158411097

c= 18.390088330493384

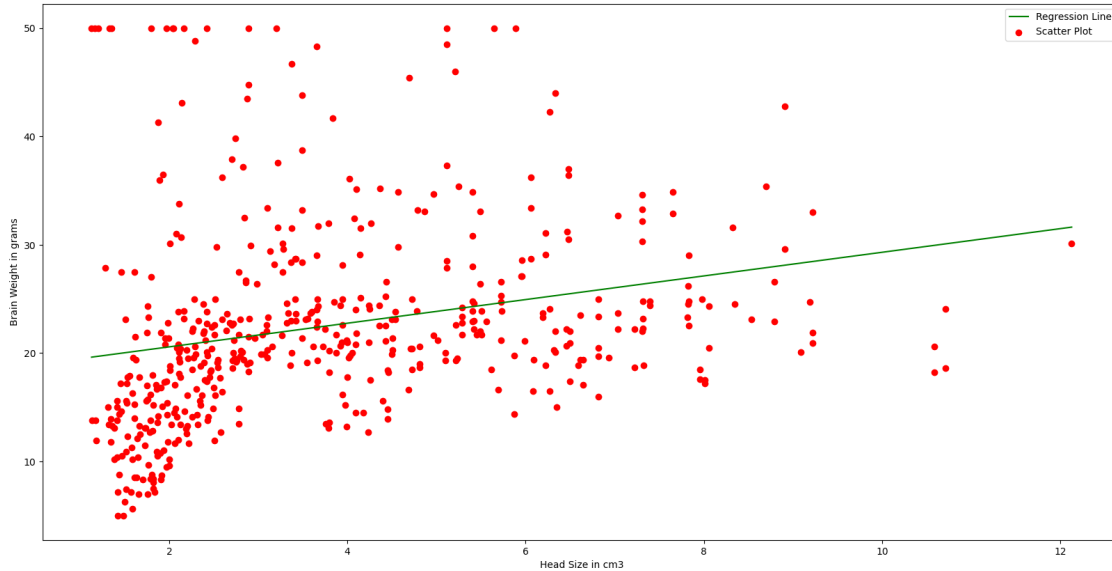
```
[11]: # Plotting Values and Regression Line
max_x = np.max(X)
min_x = np.min(X)

# Calculating line values x and y

x = np.linspace(min_x, max_x, 1000)
y = b0 + b1 * x

# Plotting Line
#plt.plot(x, y, color='#58b970', label='Regression Line')
plt.plot(x, y, color='green', label='Regression Line')
# Plotting Scatter Points
#plt.scatter(X, Y, c='#ef5423', label='Scatter Plot')
plt.scatter(X, Y, c='red', label='Scatter Plot')

plt.xlabel('Head Size in cm3')
plt.ylabel('Brain Weight in grams')
plt.legend()
plt.show()
```



```
[12]: # Calculating R2 Score
ss_tot = 0
ss_res = 0
for i in range(n):
    y_pred = b0 + b1 * X[i]
    ss_tot += (Y[i] - mean_y) ** 2
    ss_res += (Y[i] - y_pred) ** 2
r2 = 1 - (ss_res/ss_tot)
print("R2 Score")
print(r2)
```

R2 Score
0.06246437212178291

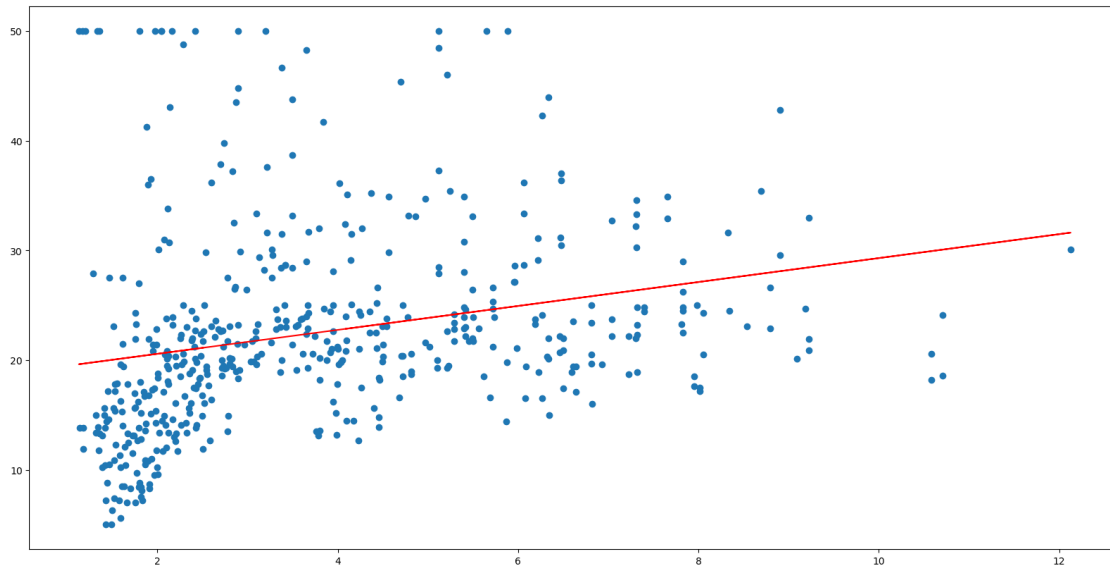
```
[13]: #using scikit-learn
```

```
[14]: # Importing Necessary Libraries
import numpy as np
import matplotlib.pyplot as plt
import pandas as pd
from sklearn.linear_model import LinearRegression
```

```
[16]: data=pd.read_csv('/Users/shreyaspeherkar/Desktop/Dataset/HousingData.csv')
X = data.iloc[:,7].values.reshape(-1,1) #converts it into numpy array
Y = data.iloc[:,13].values.reshape(-1,1)
linear_regressor=LinearRegression() # create object for class
linear_regressor.fit(X,Y) # perform linear regression
y_pred=linear_regressor.predict(X) # make prediction
```

```
[17]: plt.scatter(X,Y)
plt.plot(X,y_pred, color='red')
```

```
[17]: [<matplotlib.lines.Line2D at 0x16c728490>]
```



```
[18]: # The coefficients
print("Coefficients: \n", linear_regressor.coef_)
```

```
Coefficients:
[[1.09161302]]
```

```
[19]: from sklearn.metrics import mean_squared_error, r2_score
print("Coefficient of determination: %.2f" % r2_score(Y, y_pred))
```

```
Coefficient of determination: 0.06
```