## pratical-4

April 17, 2024

## 1 Data Analytics I

Create a Linear Regression Model using Python/R to predict home prices using Boston Housing Dataset (https://www.kaggle.com/c/boston-housing). The Boston Housing dataset contains information about various houses in Boston through different parameters. There are 506 samples and 14 feature variables in this dataset. The objective is to predict the value of prices of the house using the given features

```
[]: # # Import the required libraries
     # %matplotlib inline
     # import numpy as np
     # import pandas as pd
     # import matplotlib.pyplot as plt
     # plt.rcParams['figure.figsize'] = (20.0, 10.0)
     # from sklearn.linear_model import LinearRegression
     # from sklearn.metrics import mean_squared_error, r2_score
     # # Reading data
     # data = pd.read_csv('BostonHousing.csv')
     # print(data.shape)
     # data.head()
     # # Collecting X and Y
     # X = data['dis'].values
     # Y = data['medv'].values
     # # Calculating Coefficient
     # # Mean X and Y
     \# mean x = np.mean(X)
     \# mean_y = np.mean(Y)
     # # Total number of values
     \# n = len(X)
```

```
# # using the formula to calculate b1 and b2
# numer = 0
\# denom = 0
# for i in range(n):
    numer += (X[i] - mean_x) * (Y[i] - mean_y)
      denom += (X[i] - mean_x) ** 2
# b1 = numer/denom
\# b0 = mean_y - (b1 * mean_x)
# # m(b1) and c(b0)
# # Pinting coefficients
# print("Coefficients")
# print(f"m = {b1}")
# print(f"c = {b0}")
# # 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
# # Ploting Line
# plt.plot(x, y, color='green', label='Regression Line')
# # Ploting Scatter Points
# 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()
# # 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)
# data=pd.read_csv('BostonHousing.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 obect for class
    # linear_regressor.fit(X, Y) # perform Linear regression
    # y_pred=linear_regressor.predict(X) # make prediction
     # plt.scatter(X,Y)
    # plt.plot(X, y_pred, color = 'red')
    # # The coefficients
     # print(f"Coefficients:\n{linear_regressor.coef_}")
    # print("Coefficient of determination: %.2f" % r2_score(Y, y_pred))
[1]: %matplotlib inline
    import numpy as np
    import pandas as pd
    import matplotlib.pyplot as plt
    plt.rcParams['figure.figsize'] = (20.0, 10.0)
    from sklearn.linear_model import LinearRegression
    from sklearn.metrics import mean_squared_error, r2_score
[2]: # Reading data
    data = pd.read_csv('BostonHousing.csv')
    print(data.shape)
    data.head()
    (506, 14)
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    1 0.02731
                      7.07
                                                                   242
                 0.0
                                0 0.469
                                         6.421 78.9 4.9671
                                                                2
                                                                           17.8
    2 0.02729
                 0.0
                      7.07
                                0 0.469
                                         7.185 61.1 4.9671
                                                                2 242
                                                                           17.8
    3 0.03237
                 0.0
                                         6.998 45.8 6.0622
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                       2.18
                                0 0.458
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    4 0.06905
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                                0 0.458 7.147 54.2 6.0622
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            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
                5.33 36.2
[3]: \# Collecting X and Y
    X = data['dis'].values
    Y = data['medv'].values
```

```
[4]: # Calculating Coefficient
     # Mean X and Y
     mean_x = np.mean(X)
     mean_y = np.mean(Y)
     # Total number of values
     n = len(X)
[5]: n
[5]: 506
[6]: # using the formula to calculate b1 and b2
     numer = 0
     denom = 0
     for i in range(n):
         numer += (X[i] - mean_x) * (Y[i] - mean_y)
         denom += (X[i] - mean_x) ** 2
[7]: b1 = numer/denom
     b0 = mean_y - (b1 * mean_x)
[8]: \# m(b1) and c(b0)
     # Pinting coefficients
     print("Coefficients")
     print(f"m = {b1}")
     print(f"c = {b0}")
    Coefficients
    m = 1.0916130158411097
    c = 18.390088330493384
[9]: # Plotting Values and Regression Line
     \max_{x} = \min_{x} (X)
    min_x = np.min(X)
     \# Calculating Line values x and y
     x = np.linspace(min_x, max_x, 1000)
     y = b0 + b1 * x
     # Ploting Line
     plt.plot(x, y, color='green', label='Regression Line')
     # Ploting Scatter Points
     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()
```

```
20 - Regression Line Scatter Plot

10 - Regression
```

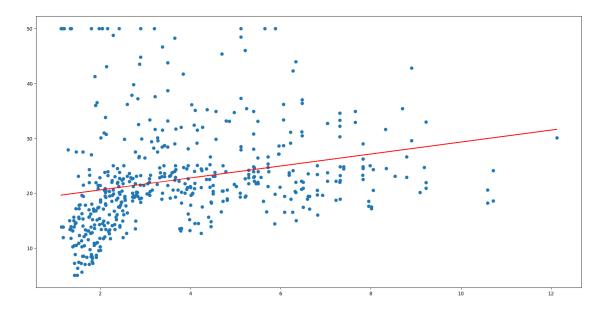
```
[10]: # Calculating R2 Score
ss_tot = 0
ss_res = 0
for i in range(n):
        y_pred = b0 + b1 * X[i]
        ss_tot += (Y[i] - mean_y) ** 2
        ss_res += (Y[i] - y_pred) ** 2
r2 = 1 - (ss_res/ss_tot)
print("R2 Score")
print(r2)
```

R2 Score 0.06246437212178291

```
[11]: data=pd.read_csv('BostonHousing.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 obect for class
linear_regressor.fit(X, Y) # perform Linear regression
y_pred=linear_regressor.predict(X) # make prediction
```

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

## [12]: [<matplotlib.lines.Line2D at 0x2e079998890>]



```
[13]: # The coefficients
print(f"Coefficients:\n{linear_regressor.coef_}")
```

Coefficients:

[[1.09161302]]

[14]: print("Coefficient of determination: %.2f" % r2\_score(Y, y\_pred))

Coefficient of determination: 0.06

[]: