linearregression-example

April 18, 2023

0.1 Predict Banglore House Price Using Liearn Regression - Supervised Learning Algo.

```
[1]: import pandas as pd
[2]: df = pd.read_csv("/content/drive/MyDrive/MyDataSet/bangalore house price_
      ⇔prediction OHE-data.csv")
     df.head()
[2]:
        bath balcony price total_sqft_int
                                                    price_per_sqft
                                               bhk
     0
         3.0
                  2.0
                       150.0
                                       1672.0
                                                  3
                                                        8971.291866
     1
         3.0
                  3.0
                       149.0
                                       1750.0
                                                        8514.285714
                                                  3
     2
         3.0
                  2.0 150.0
                                       1750.0
                                                  3
                                                        8571.428571
     3
         2.0
                  2.0
                        40.0
                                       1250.0
                                                  2
                                                        3200.000000
         2.0
                  2.0
     4
                        83.0
                                       1200.0
                                                        6916.666667
        area_typeSuper built-up Area area_typeBuilt-up Area
     0
                                     1
                                                               0
     1
                                     0
                                                               1
     2
                                     1
                                                               0
     3
                                     1
                                                               0
                                     0
     4
                                                               0
        area_typePlot Area
                             availability_Ready To Move
     0
                           0
                           0
     1
                                                        1
     2
                           0
                                                        1
                           0
     3
                                                        1
     4
        location_Kalena Agrahara
                                  location_Horamavu Agara location_Vidyaranyapura
     0
     1
                                0
                                                          0
                                                                                    0
                                0
                                                          0
     2
                                                                                    0
     3
                                0
                                                          0
                                                                                    0
     4
                                0
                                                          0
                                                                                    0
```

location_BTM 2nd Stage location_Hebbal Kempapura location_Hosur Road \

```
1
                          0
                                                    0
                                                                       0
    2
                          0
                                                    0
                                                                       0
    3
                                                    0
                                                                       0
                          0
    4
                           0
                                                    0
                                                                       0
                                 location_Domlur location_Mahadevpura
       location_Horamavu Banaswadi
    0
    1
                               0
                                               0
                                                                    0
    2
                               0
                                               0
                                                                    0
    3
                               0
                                               0
                                                                    0
    4
                               0
                                               0
                                                                    0
       location_Tumkur Road
    0
                         0
                         0
    1
    2
                         0
    3
                         0
    4
                         0
    [5 rows x 108 columns]
[3]: # "price" is our dependent Variable
    X = df.drop("price",axis=1) # X = Matrix
    y = df["price"]
                             # y = Vector
    print(X.shape)
    print(y.shape)
    (7120, 107)
    (7120,)
[4]: from sklearn.model_selection import train_test_split
    →random_state=51)
    print(X_train.shape)
    print(y_train.shape)
    print(X_test.shape)
    print(y_test.shape)
    (5696, 107)
    (5696,)
    (1424, 107)
    (1424,)
```

1 Feature Scaling

```
[5]: from sklearn.preprocessing import StandardScaler
    sc = StandardScaler()
    sc.fit(X train)
    X_train = sc.transform(X_train)
    X test = sc.transform(X test)
    #Linear Regression - ML- Model Training
[6]: from sklearn.linear model import LinearRegression
    lr = LinearRegression()
    lr.fit(X_train,y_train)
[6]: LinearRegression()
[7]:
    lr.coef_
[7]: array([-5.70206143e+00, -1.25679916e+00, 8.27341833e+01, -1.44906911e+01,
                             1.88468905e-01, -1.72593897e+00, -4.51058311e+00,
            5.75662723e+01,
            -2.22589244e+00, -4.28978455e+00, -2.44590976e+00, 5.40246226e-01,
            -1.03633400e+00, 1.43064873e+00, -6.25029424e-02, -1.51548783e+00,
           -2.14422789e-01, 2.16244155e+00, -1.48710228e+00, 1.95250816e+00,
           -3.10761125e+00, -1.28138668e+00, -1.01367155e+00, 1.37968545e-02,
            1.10383858e+00, 1.26497611e+00, -3.52405517e+00, -1.21398741e+00,
            -5.04622019e-01, 1.46299181e+00, -5.50064233e-01, -8.46468162e-02,
            6.84882188e-01, -1.39849820e+00, -1.94761710e-02, -1.57716300e+00,
            4.20886278e-01, 8.03443207e-01, 2.99182164e+00, 3.86430413e-03,
            1.05037261e-01, 2.89115612e-01, -3.16916626e-01, 1.05625868e+00,
            -1.39649279e+00, -3.10533604e+00, 1.01764011e-01, -7.49672917e-02,
            -8.03271555e-01, -1.27061856e+00, -8.54046164e-01, 2.64566484e-01,
            9.10688839e-01, -8.23059458e-01, -9.07215234e-01, 1.22059216e+00,
            2.11418894e+00, -5.38187400e-01, -1.32164338e+00, -8.28349340e-01,
            1.28167980e+00, -1.92911295e-01, 6.65824485e-02,
                                                               3.65563139e-02,
            -1.85069853e+00, 1.49068024e+00, -9.57964753e-01, -9.36110163e-01,
            -7.45634897e-01, 7.22643165e-02, -6.79260144e-01, -1.70853833e-01,
            -1.72288643e+00, -1.15833746e+00, 5.78931788e-01, 1.37836966e+00,
            -1.14424496e+00, 3.96188294e-01, -6.08013157e-01, -2.20959218e+00,
            3.45270810e-01, 1.01747431e-03, 1.06563895e-01, 3.04728530e+00,
            2.09496392e+00, -8.13481923e-01, -4.18437282e-01, 2.30993396e+00,
            3.31858800e-02, 8.07865914e-02, 5.37064987e-02, 1.55347699e+00,
            8.13889657e-01, -1.14636462e+00, 3.41805788e-01, -8.28022037e-01,
            1.68897360e+00, 2.97657524e-01, 9.59437517e-01, 4.57297702e-01,
            -2.22729515e-01, -1.48290835e+00, -6.26342867e-01,
                                                               5.86538254e-01,
            -1.78547310e+00, 2.19020231e-01, -3.45032599e-01])
```

```
[8]: lr.intercept_
```

[8]: 95.0802729985955

2 Predict the Value of Home and Test

```
[9]: X test[0,:]
 [9]: array([ 0.71301986, 0.0112734 , 0.30202307, 0.65677518, -0.48064341,
             -1.7385623, 2.11587407, -0.25430867, 0.51007548, -0.18373025,
             -0.16389438, -0.1473229 , -0.13023539, -0.12812824, -0.12598816,
             -0.12454231, -0.12953656, -0.12381344, -0.12010681, -0.11551113,
             -0.10992018, -0.10909925, -0.10660036, -0.11234866, -0.09315135,
             -0.08618799, -0.08923672, -0.09023078, -0.08721571, -0.09023078,
             -0.08721571, -0.08195215, -0.08195215, -0.07633675, -0.0751646,
             -0.08085949, -0.0739743 , -0.07975227, -0.07153563, -0.0751646 ,
             -0.0677166 , -0.08085949, -0.07153563, -0.07862985, -0.0751646 ,
             -0.07862985, -0.06504853, -0.0751646 , -0.06901264, -0.0751646 ,
             -0.06901264, -0.07028523, -0.07276497, -0.07028523, -0.06367332,
             -0.06226825, -0.06226825, -0.06639573, -0.06504853, -0.05935999,
             -0.06083125, -0.06639573, -0.06639573, -0.06226825, -0.06367332,
             -0.05935999, -0.06639573, -0.06367332, -0.06226825, -0.06226825,
             -0.05935999, -0.05935999, -0.05935999, -0.05630391, -0.05935999,
             -0.05785186, -0.05935999, -0.05935999, -0.06083125, -0.06083125,
             -0.05471275, -0.06083125, -0.06226825, -0.05935999, -0.05935999,
             -0.06226825, -0.06226825, -0.05785186, -0.06504853, -0.06226825,
             -0.06083125, -0.05935999, -0.05307449, -0.05630391, -0.06226825,
             -0.05471275, -0.05935999, -0.05471275, -0.05471275, -0.05138463,
             -0.05307449, -0.05307449, -0.05471275, -0.05471275, -0.05630391,
             -0.05630391, -0.05138463
[10]: lr.predict([X_test[0,:]])
[10]: array([76.90661876])
[11]: lr.predict(X_test)
[11]: array([76.90661876, 15.25005377, 113.6828165, ..., 21.30296864,
              71.43462962, 230.0414626 ])
[12]:
     y_test
[12]: 2435
               80.00
      3113
               40.00
      426
              120.00
      1124
               79.00
```

```
1161
               45.00
      2078
               28.34
               84.00
      6855
      4381
               32.00
      3862
               63.00
      43
              180.00
      Name: price, Length: 1424, dtype: float64
     #Check Accuracy for Train Predicted Data and Test Data
[13]: lr.score(X_test,y_test)
[13]: 0.7903837092682251
     #Implementing Ridge & Lasso Regression
[14]: from sklearn.linear_model import Ridge, Lasso
[15]: rd = Ridge()
      rd.fit(X_train,y_train)
[15]: Ridge()
[16]: rd.score(X_test, y_test)
[16]: 0.7905686374336628
[17]: ls = Lasso()
      ls.fit(X_train,y_train)
      ls.score(X_test, y_test)
[17]: 0.8036373003525774
[18]: rd2 = Ridge(alpha=2)
      rd2.fit(X_train,y_train)
      rd2.score(X_test, y_test)
[18]: 0.7907530260397942
[19]: ls2= Lasso(alpha=2)
      ls2.fit(X_train,y_train)
      ls2.score(X_test, y_test)
[19]: 0.8160181533703601
[20]: ls3= Lasso(alpha=3)
      ls3.fit(X_train,y_train)
```

ls3.score(X_test, y_test)

[20]: 0.8263450613017426