

# Alyakbar Ahmed Sheikh RegNo. BCS-03-0003/2022

## Assignment 1 Question

1. USING THE HOUSING PRICING DATA ATTACHED, CREATE A LINEAR REGRESSION TO PREDICT THE HOUSE PRICES ANALYZE THE PERFORMANCE OF THE MODEL USING MSE, MAE MAE AND RMSE Real estate.csv importing the libraries needed

loading the data

```
In [ ]: import numpy as np
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import LinearRegression
        from sklearn.metrics import mean_squared_error, mean_absolute_error
        import pandas as pd
        df = pd.read_csv('Real estate.csv')
        print(df)
```

|     | No  | X1 transaction date | X2 house age \ |
|-----|-----|---------------------|----------------|
| 0   | 1   | 2012.917            | 32.0           |
| 1   | 2   | 2012.917            | 19.5           |
| 2   | 3   | 2013.583            | 13.3           |
| 3   | 4   | 2013.500            | 13.3           |
| 4   | 5   | 2012.833            | 5.0            |
| ..  | ... | ...                 | ...            |
| 409 | 410 | 2013.000            | 13.7           |
| 410 | 411 | 2012.667            | 5.6            |
| 411 | 412 | 2013.250            | 18.8           |
| 412 | 413 | 2013.000            | 8.1            |
| 413 | 414 | 2013.500            | 6.5            |

|     | X3 distance to the nearest MRT station | X4 number of convenience stores \ |
|-----|--|-----------------------------------|
| 0   | 84.87882                               | 10                                |
| 1   | 306.59470                              | 9                                 |
| 2   | 561.98450                              | 5                                 |
| 3   | 561.98450                              | 5                                 |
| 4   | 390.56840                              | 5                                 |
| ..  | ...                                    | ...                               |
| 409 | 4082.01500                             | 0                                 |
| 410 | 90.45606                               | 9                                 |
| 411 | 390.96960                              | 7                                 |
| 412 | 104.81010                              | 5                                 |
| 413 | 90.45606                               | 9                                 |

|     | X5 latitude | X6 longitude | Y house price of unit area |
|-----|-------------|--------------|----------------------------|
| 0   | 24.98298    | 121.54024    | 37.9                       |
| 1   | 24.98034    | 121.53951    | 42.2                       |
| 2   | 24.98746    | 121.54391    | 47.3                       |
| 3   | 24.98746    | 121.54391    | 54.8                       |
| 4   | 24.97937    | 121.54245    | 43.1                       |
| ..  | ...         | ...          | ...                        |
| 409 | 24.94155    | 121.50381    | 15.4                       |
| 410 | 24.97433    | 121.54310    | 50.0                       |
| 411 | 24.97923    | 121.53986    | 40.6                       |
| 412 | 24.96674    | 121.54067    | 52.5                       |
| 413 | 24.97433    | 121.54310    | 63.9                       |

[414 rows x 8 columns]

Exploring the data or look inside it

```
In [ ]: print(df.head())
        print(df.info())
```

```

      No  X1 transaction date  X2 house age  \
0    1          2012.917          32.0
1    2          2012.917          19.5
2    3          2013.583          13.3
3    4          2013.500          13.3
4    5          2012.833           5.0

      X3 distance to the nearest MRT station  X4 number of convenience stores  \
0                                84.87882                                10
1                                306.59470                                9
2                                561.98450                                5
3                                561.98450                                5
4                                390.56840                                5

      X5 latitude  X6 longitude  Y house price of unit area
0      24.98298      121.54024              37.9
1      24.98034      121.53951              42.2
2      24.98746      121.54391              47.3
3      24.98746      121.54391              54.8
4      24.97937      121.54245              43.1
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 414 entries, 0 to 413
Data columns (total 8 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   No                                    414 non-null   int64
1   X1 transaction date                  414 non-null   float64
2   X2 house age                        414 non-null   float64
3   X3 distance to the nearest MRT station 414 non-null   float64
4   X4 number of convenience stores        414 non-null   int64
5   X5 latitude                          414 non-null   float64
6   X6 longitude                          414 non-null   float64
7   Y house price of unit area            414 non-null   float64
dtypes: float64(6), int64(2)
memory usage: 26.0 KB
None

```

seperating the data into variables

```
In [ ]: X = df.drop(['No', 'Y house price of unit area'], axis=1)
        y = df['Y house price of unit area']
```

Training and testing the sets

```
In [ ]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_sta
```

creating linear regression

```
In [ ]: model = LinearRegression()
        model.fit(X_train, y_train)
```

Out [ ]:

LinearRegression ⓘ ?

LinearRegression()

Making predictions

In [ ]:

```
predictions = model.predict(X_test)
```

Calculating the Mean Squared Error (MSE), Mean Absolute Error (MAE) and Root Mean Squared Error (RMSE)

In [ ]:

```
mse = mean_squared_error(y_test, predictions)
mae = mean_absolute_error(y_test, predictions)
rmse = np.sqrt(mse)

print("Mean Squared Error (MSE):", mse)
print("Mean Absolute Error (MAE):", mae)
print("Root Mean Squared Error (RMSE):", rmse)
```

Mean Squared Error (MSE): 53.50561912450505  
Mean Absolute Error (MAE): 5.305355690074272  
Root Mean Squared Error (RMSE): 7.314753524521866