CH: - 5 Regression Analysis ¶

1. Supervised Learning

Regression(integer or float)

- Linear
- 1.)simple
- 2.)multiple
- polynomial
- 1.)simple
- 2.)multiple

Classification(categorical data)

- KNN
- · Decision Tree

equation

- y = mx+b (linear)
- y = a+bx+cx1+dx2+ex3 (linear)
- $y = a+bx+cx^2+dx^3$ (polynomial)
- $y = a+bx1+cx2^2+dx3^3$ (polynomial)

Simple Linear Regression

scikit library

preprocessing

```
In [2]: import matplotlib.pyplot as plt
import pandas as pd
import numpy as np
df = pd.read_csv("Book1.csv")
df.head()
```

Out[2]:

| | cgpa | package |
|---|------|---------|
| 0 | 6.89 | 3.26 |
| 1 | 5.12 | 1.98 |
| 2 | 7.82 | 3.25 |
| 3 | 7.42 | 3.67 |
| 4 | 6.94 | 3.57 |

```
In [3]: x = df.iloc[:,0:1]
    y = df.iloc[:,-1]
    # x = df.drop("package",axis=1)
    # x = df[['cgpa', 'x1', 'x2']]
    # x = df['cgpa']
    # y = df['package']

In [4]: x.shape

Out[4]: (200, 1)

In [5]: y.shape

Out[5]: (200,)
    • processing

In [6]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    x_train.shape

Out[6]: (160, 1)
```

In [7]: help(train_test_split)

```
Help on function train_test_split in module sklearn.model_selection._split:
train_test_split(*arrays, **options)
    Split arrays or matrices into random train and test subsets
    Quick utility that wraps input validation and
    ``next(ShuffleSplit().split(X, y))`` and application to input data
    into a single call for splitting (and optionally subsampling) data in a
    oneliner.
    Read more in the :ref:`User Guide <cross_validation>`.
    Parameters
    -----
    *arrays : sequence of indexables with same length / shape[0]
        Allowed inputs are lists, numpy arrays, scipy-sparse
        matrices or pandas dataframes.
    test_size : float or int, default=None
        If float, should be between 0.0 and 1.0 and represent the proportion
        of the dataset to include in the test split. If int, represents the
        absolute number of test samples. If None, the value is set to the
        complement of the train size. If ``train_size`` is also None, it wil
1
        be set to 0.25.
    train_size : float or int, default=None
        If float, should be between 0.0 and 1.0 and represent the
        proportion of the dataset to include in the train split. If
        int, represents the absolute number of train samples. If None,
        the value is automatically set to the complement of the test size.
    random_state : int or RandomState instance, default=None
        Controls the shuffling applied to the data before applying the spli
t.
        Pass an int for reproducible output across multiple function calls.
        See :term:`Glossary <random state>`.
    shuffle : bool, default=True
        Whether or not to shuffle the data before splitting. If shuffle=Fals
e
        then stratify must be None.
    stratify : array-like, default=None
        If not None, data is split in a stratified fashion, using this as
        the class labels.
    Returns
    splitting : list, length=2 * len(arrays)
        List containing train-test split of inputs.
        .. versionadded:: 0.16
            If the input is sparse, the output will be a
            ``scipy.sparse.csr_matrix``. Else, output type is the same as th
            input type.
    Examples
```

```
>>> import numpy as np
             >>> from sklearn.model_selection import train_test_split
             >>> X, y = np.arange(10).reshape((5, 2)), range(5)
             >>> X
             array([[0, 1],
                    [2, 3],
                    [4, 5],
                    [6, 7],
                    [8, 9]])
             >>> list(y)
             [0, 1, 2, 3, 4]
             >>> X_train, X_test, y_train, y_test = train_test_split(
                     X, y, test_size=0.33, random_state=42)
             . . .
             >>> X_train
             array([[4, 5],
                    [0, 1],
                    [6, 7]]
             >>> y_train
             [2, 0, 3]
             >>> X test
             array([[2, 3],
                    [8, 9]])
             >>> y_test
             [1, 4]
             >>> train_test_split(y, shuffle=False)
             [[0, 1, 2], [3, 4]]
 In [8]: from sklearn.model_selection import train_test_split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         x_test.shape
 Out[8]: (40, 1)
 In [9]: from sklearn.model selection import train test split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         y_train.shape
Out[9]: (160,)
In [10]: from sklearn.model selection import train test split
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         y_test.shape
Out[10]: (40,)
In [11]: | from sklearn.model_selection import train_test_split
         from sklearn.linear model import LinearRegression
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         lr=LinearRegression()
         model = lr.fit(x_train,y_train) # creating a model or equation
         lr.coef
Out[11]: array([0.56087409])
```

```
In [12]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    lr.intercept_
```

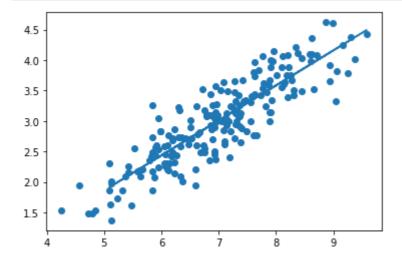
Out[12]: -0.9923344509732703

posting

```
In [13]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    y_pred = lr.predict(x_test)
    y_pred
```

```
Out[13]: array([2.0881033 , 3.27231514, 2.37569761, 2.93960801, 3.41329274, 2.51667521, 2.67457012, 2.96216442, 2.31930657, 2.34750209, 3.90389479, 3.18772858, 2.91141249, 4.12381985, 3.76291719, 3.92081211, 2.42644954, 2.96780353, 3.31742798, 2.34750209, 1.78923079, 2.61253997, 1.45652365, 2.86066055, 3.31742798, 3.50915751, 2.54487073, 2.6463746 , 4.21968462, 2.05426868, 2.80990862, 2.3700585 , 2.97908174, 3.44148826, 3.96592494, 3.0749465 , 4.45652699, 2.34750209, 3.74036078, 2.11065972])
```

```
In [14]: from sklearn.model_selection import train_test_split
    from sklearn.linear_model import LinearRegression
    import matplotlib.pyplot as plt
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
    lr=LinearRegression()
    model = lr.fit(x_train,y_train) # creating a model or equation
    y_pred = lr.predict(x_test)
    plt.scatter(df['cgpa'],df['package'])
    plt.plot(x_test,y_pred)
    plt.show()
```



```
from sklearn.model_selection import train_test_split
In [15]:
         from sklearn.linear_model import LinearRegression
         from sklearn import metrics
         import math
         x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2)
         lr=LinearRegression()
         model = lr.fit(x_train,y_train)
         y_pred = lr.predict(x_test)
         print("MAE",metrics.mean_absolute_error(y_test,y_pred))
         print("MSE", metrics.mean squared error(y test, y pred))
         print("RMSE",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
         print("R2score", metrics.r2_score(y_test,y_pred)) # x depend on y
         MAE 0.28640979648873444
         MSE 0.1297678341291693
         RMSE 0.3602330275379665
         R2score 0.6884288307682702
In [16]: import numpy as np
         test_data = np.array([9.3,7.8])
         print(test_data)
         print(test_data.shape)
         [9.3 7.8]
         (2,)
In [17]: import numpy as np
         test_data = test_data.reshape(-1,1)
         print(test data)
         print(test_data.shape)
         [[9.3]
          [7.8]]
         (2, 1)
In [18]: lr.predict(test_data)
Out[18]: array([4.35342271, 3.48035551])
```

Multiple Linear Regression

```
In [19]: # Target is selling price.
         import pandas as pd
         df = pd.read_csv("car data.csv")
         df.isna().sum()
Out[19]: Car_Name
                           0
                           0
         Year
         Selling_Price
         Present_Price
                           0
         Kms_Driven
         Fuel_Type
         Seller Type
                           0
         Transmission
                           0
         0wner
         dtype: int64
```

```
In [20]: import pandas as pd
    df = pd.read_csv("car data.csv")
    df = df.drop("Car_Name",axis=1)
    df
```

Out[20]:

| | Year | Selling_Price | Present_Price | Kms_Driven | Fuel_Type | Seller_Type | Transmission | Owi |
|-----|------|---------------|---------------|------------|-----------|-------------|--------------|-----|
| 0 | 2014 | 3.35 | 5.59 | 27000 | Petrol | Dealer | Manual | |
| 1 | 2013 | 4.75 | 9.54 | 43000 | Diesel | Dealer | Manual | |
| 2 | 2017 | 7.25 | 9.85 | 6900 | Petrol | Dealer | Manual | |
| 3 | 2011 | 2.85 | 4.15 | 5200 | Petrol | Dealer | Manual | |
| 4 | 2014 | 4.60 | 6.87 | 42450 | Diesel | Dealer | Manual | |
| | | | | | | | | |
| 296 | 2016 | 9.50 | 11.60 | 33988 | Diesel | Dealer | Manual | |
| 297 | 2015 | 4.00 | 5.90 | 60000 | Petrol | Dealer | Manual | |
| 298 | 2009 | 3.35 | 11.00 | 87934 | Petrol | Dealer | Manual | |
| 299 | 2017 | 11.50 | 12.50 | 9000 | Diesel | Dealer | Manual | |
| 300 | 2016 | 5.30 | 5.90 | 5464 | Petrol | Dealer | Manual | |
| | | | | | | | | |

301 rows × 8 columns

In [21]: df["Age"] = 2025-df["Year"]
df

Out[21]:

| | Year | Selling_Price | Present_Price | Kms_Driven | Fuel_Type | Seller_Type | Transmission | Owi | | | |
|-------|----------------------|---------------|---------------|------------|-----------|-------------|--------------|-----|--|--|--|
| 0 | 2014 | 3.35 | 5.59 | 27000 | Petrol | Dealer | Manual | | | | |
| 1 | 2013 | 4.75 | 9.54 | 43000 | Diesel | Dealer | Manual | | | | |
| 2 | 2017 | 7.25 | 9.85 | 6900 | Petrol | Dealer | Manual | | | | |
| 3 | 2011 | 2.85 | 4.15 | 5200 | Petrol | Dealer | Manual | | | | |
| 4 | 2014 | 4.60 | 6.87 | 42450 | Diesel | Dealer | Manual | | | | |
| | | | | | | | | | | | |
| 296 | 2016 | 9.50 | 11.60 | 33988 | Diesel | Dealer | Manual | | | | |
| 297 | 2015 | 4.00 | 5.90 | 60000 | Petrol | Dealer | Manual | | | | |
| 298 | 2009 | 3.35 | 11.00 | 87934 | Petrol | Dealer | Manual | | | | |
| 299 | 2017 | 11.50 | 12.50 | 9000 | Diesel | Dealer | Manual | | | | |
| 300 | 2016 | 5.30 | 5.90 | 5464 | Petrol | Dealer | Manual | | | | |
| 201 r | 201 rows x 0 columns | | | | | | | | | | |

301 rows × 9 columns

In [22]: df = df.drop("Year",axis=1)
df

Out[22]:

| | Selling_Price | Present_Price | Kms_Driven | Fuel_Type | Seller_Type | Transmission | Owner | Ą |
|-----|---------------|---------------|------------|-----------|-------------|--------------|-------|---|
| 0 | 3.35 | 5.59 | 27000 | Petrol | Dealer | Manual | 0 | |
| 1 | 4.75 | 9.54 | 43000 | Diesel | Dealer | Manual | 0 | |
| 2 | 7.25 | 9.85 | 6900 | Petrol | Dealer | Manual | 0 | |
| 3 | 2.85 | 4.15 | 5200 | Petrol | Dealer | Manual | 0 | |
| 4 | 4.60 | 6.87 | 42450 | Diesel | Dealer | Manual | 0 | |
| | | | | | | | | |
| 296 | 9.50 | 11.60 | 33988 | Diesel | Dealer | Manual | 0 | |
| 297 | 4.00 | 5.90 | 60000 | Petrol | Dealer | Manual | 0 | |
| 298 | 3.35 | 11.00 | 87934 | Petrol | Dealer | Manual | 0 | |
| 299 | 11.50 | 12.50 | 9000 | Diesel | Dealer | Manual | 0 | |
| 300 | 5.30 | 5.90 | 5464 | Petrol | Dealer | Manual | 0 | |
| | | | | | | | | |

301 rows × 8 columns

In [23]: pd.get_dummies(df)

Out[23]:

| | Selling_Price | Present_Price | Kms_Driven | Owner | Age | Fuel_Type_CNG | Fuel_Type_Diesel | | | |
|-------|-----------------------|---------------|------------|-------|-----|---------------|------------------|--|--|--|
| 0 | 3.35 | 5.59 | 27000 | 0 | 11 | 0 | 0 | | | |
| 1 | 4.75 | 9.54 | 43000 | 0 | 12 | 0 | 1 | | | |
| 2 | 7.25 | 9.85 | 6900 | 0 | 8 | 0 | 0 | | | |
| 3 | 2.85 | 4.15 | 5200 | 0 | 14 | 0 | 0 | | | |
| 4 | 4.60 | 6.87 | 42450 | 0 | 11 | 0 | 1 | | | |
| | | | | | | | | | | |
| 296 | 9.50 | 11.60 | 33988 | 0 | 9 | 0 | 1 | | | |
| 297 | 4.00 | 5.90 | 60000 | 0 | 10 | 0 | 0 | | | |
| 298 | 3.35 | 11.00 | 87934 | 0 | 16 | 0 | 0 | | | |
| 299 | 11.50 | 12.50 | 9000 | 0 | 8 | 0 | 1 | | | |
| 300 | 5.30 | 5.90 | 5464 | 0 | 9 | 0 | 0 | | | |
| 301 r | 301 rows × 12 columns | | | | | | | | | |

In [24]: df = pd.get_dummies(df,drop_first=True) # only one column remove from each cat
df

Out[24]:

| | Selling_Price | Present_Price | Kms_Driven | Owner | Age | Fuel_Type_Diesel | Fuel_Type_Petrol |
|-----|---------------|---------------|------------|-------|-----|------------------|------------------|
| 0 | 3.35 | 5.59 | 27000 | 0 | 11 | 0 | 1 |
| 1 | 4.75 | 9.54 | 43000 | 0 | 12 | 1 | 0 |
| 2 | 7.25 | 9.85 | 6900 | 0 | 8 | 0 | 1 |
| 3 | 2.85 | 4.15 | 5200 | 0 | 14 | 0 | 1 |
| 4 | 4.60 | 6.87 | 42450 | 0 | 11 | 1 | 0 |
| | | | | | | | |
| 296 | 9.50 | 11.60 | 33988 | 0 | 9 | 1 | 0 |
| 297 | 4.00 | 5.90 | 60000 | 0 | 10 | 0 | 1 |
| 298 | 3.35 | 11.00 | 87934 | 0 | 16 | 0 | 1 |
| 299 | 11.50 | 12.50 | 9000 | 0 | 8 | 1 | 0 |
| 300 | 5.30 | 5.90 | 5464 | 0 | 9 | 0 | 1 |
| | | | | | | | |

301 rows × 9 columns

```
In [25]: # target is selling price
y = df['Selling_Price']
x = df.drop('Selling_Price',axis=1)
x.shape
```

Out[25]: (301, 8)

```
In [26]: y = df['Selling_Price']
x = df.drop('Selling_Price',axis=1)
y.shape
```

Out[26]: (301,)

```
In [27]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_stat
    x_train.shape
```

Out[27]: (240, 8)

```
In [28]: x_test.shape
```

Out[28]: (61, 8)

```
In [29]: y_train.shape
```

Out[29]: (240,)

```
In [30]: y_test.shape
```

Out[30]: (61,)

```
In [31]:
         from sklearn.linear model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.coef
Out[31]: array([ 4.28802981e-01, -5.92316903e-06, -8.49009727e-01, -4.06411225e-01,
                 2.86056285e+00, 7.03103550e-01, -1.07477347e+00, -1.50897975e+00])
In [32]: print(x.columns)
         Index(['Present_Price', 'Kms_Driven', 'Owner', 'Age', 'Fuel_Type_Diesel',
                'Fuel_Type_Petrol', 'Seller_Type_Individual', 'Transmission_Manual'],
               dtype='object')
In [33]: from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.intercept
Out[33]: 6.850757321843104
In [34]: from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         y pred
                             0.71754255, 4.23613904, 5.17855444,
Out[34]: array([10.57889241,
                                                                    9.75533583,
                 4.20615652, 2.6748089, 7.63221623, 0.17032381,
                                                                    5.13283747,
                 6.15727726, 6.44545506, 2.11900126, 7.6644102, 1.91710055,
                 1.71680788,
                             2.02672159, 1.85294787, 9.56717323, 4.23076317,
                             9.36189904, 1.46103587, 9.54572517,
                 1.48724915,
                                                                    0.82954545,
                 8.32266051, 1.53698563, -3.16293717, 4.2127377, 2.10385587,
                 3.42279444, 3.71284702, 5.54538889, 7.6834828, -1.89757368,
                             8.46027656, 5.77704136, 6.10531922,
                 7.06591847,
                                                                    6.52919752,
                16.08564344,
                             2.07410694, 1.05462465, -0.44710144, 7.06940454,
                 6.73548589, 0.98609923, 7.08961114, 14.16420523, 3.00434598,
                 8.2871215 , -0.87862345, 8.8814273 , 1.15099681, 2.11848057,
                             0.69287067, 9.88583992, -0.45444272, -2.41051663,
                -0.82641923,
                10.28627971])
In [35]: from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         print("MAE : ",metrics.mean_absolute_error(y_test,y_pred))
         print("MSE : ",metrics.mean_squared_error(y_test,y_pred))
         print("RMSE : ",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
         print("R2score : ",metrics.r2 score(y test,y pred))
         MAE : 1.2453565634870642
         MSE: 2.750689262139753
         RMSE: 1.658520202511791
```

R2score: 0.8502332355855595

Advertising csv

```
In [38]:
        import pandas as pd
         df = pd.read_csv("advertising.csv")
         df.isna().sum()
Out[38]: TV
                      0
         Radio
         Newspaper
                      0
         Sales
                       0
         dtype: int64
In [39]: y = df["Sales"]
         x = df.drop("Sales",axis=1)
In [40]: from sklearn.model_selection import train_test_split
         x train,x test,y train,y test = train test split(x,y,test size=0.2)
         x train.shape
Out[40]: (160, 3)
In [41]: x_test.shape
Out[41]: (40, 3)
In [42]: |y_train.shape
Out[42]: (160,)
In [43]: |y_test.shape
Out[43]: (40,)
```

```
In [44]:
         from sklearn.linear model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.coef
Out[44]: array([0.05313408, 0.11357858, 0.00333475])
In [45]: | from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         lr.intercept_
Out[45]: 4.645434610871224
In [46]: from sklearn.linear_model import LinearRegression
         lr = LinearRegression()
         model = lr.fit(x_train,y_train)
         y_pred = lr.predict(x_test)
         y_pred
Out[46]: array([ 7.32219292, 9.44780358, 16.08001839, 10.91179974, 9.94024746,
                16.95992676, 10.11639368, 12.36888929, 11.62393635, 12.57834914,
                11.08245536, 14.11836139, 8.3597048, 21.14862847, 12.56036115,
                20.98615338, 10.46624475, 10.35001541, 17.97571054, 9.65169244,
                19.07693595, 9.04246403, 15.35125093, 8.93746723, 16.92844
                14.69215049, 9.60888352, 8.02648269, 11.81179103, 17.05220234,
                13.75124131, 17.83846248, 20.65012944, 16.66301181, 9.64682542,
                 7.78662847, 24.85375342, 9.88188164, 17.58112532, 12.78679528])
In [47]: from sklearn import metrics
         import math
         y_pred = lr.predict(x_test)
         print("MAE : ",metrics.mean_absolute_error(y_test,y_pred))
         print("MSE : ",metrics.mean_squared_error(y_test,y_pred))
         print("RMSE : ",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
         print("R2score : ",metrics.r2_score(y_test,y_pred))
         MAE :
                1.220206245120477
         MSE: 2.7003495910559043
         RMSE: 1.643274046243019
```

Simple Ploynomial regression

```
In [48]: import pandas as pd
    df = pd.read_csv("temp.csv")
    df
```

Out[48]:

| | sno | Temperature | Pressure |
|---|-----|-------------|----------|
| 0 | 1 | 0 | 0.0002 |
| 1 | 2 | 20 | 0.0012 |
| 2 | 3 | 40 | 0.0060 |
| 3 | 4 | 60 | 0.0300 |
| 4 | 5 | 80 | 0.0900 |
| 5 | 6 | 100 | 0.2700 |

```
In [49]: df = df.drop("sno",axis=1)
df
```

Out[49]:

| | Temperature | Pressure |
|---|-------------|----------|
| 0 | 0 | 0.0002 |
| 1 | 20 | 0.0012 |
| 2 | 40 | 0.0060 |
| 3 | 60 | 0.0300 |
| 4 | 80 | 0.0900 |
| 5 | 100 | 0.2700 |

```
In [50]: y = df["Pressure"]
x = df.drop("Pressure",axis=1)
```

```
In [51]: from sklearn.preprocessing import PolynomialFeatures
poly = PolynomialFeatures(degree=2)
x_poly = poly.fit_transform(x)
x_poly
```

```
In [65]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x_poly,y,test_size=0.2,randon
```

Out[69]: array([0.02446304, 0.07210816])

lr.predict(test_data)

Multiple Polynomial Regression

test_data = poly.fit_transform(test_data)

```
In [99]: import pandas as pd
df = pd.read_csv("Real estate.csv")
df
```

Out[99]:

| | No | X1 transaction date | X2 house age | X3 distance to the nearest MRT station | X4 number of convenience stores | X5 latitude | X6 longitude | Y house price of unit area |
|-----|-----|---------------------------|--------------------|--|---------------------------------|----------------|-----------------|-------------------------------------|
| 0 | 1 | 2012.917 | 32.0 | 84.87882 | 10 | 24.98298 | 121.54024 | 37.9 |
| 1 | 2 | 2012.917 | 19.5 | 306.59470 | 9 | 24.98034 | 121.53951 | 42.2 |
| 2 | 3 | 2013.583 | 13.3 | 561.98450 | 5 | 24.98746 | 121.54391 | 47.3 |
| 3 | 4 | 2013.500 | 13.3 | 561.98450 | 5 | 24.98746 | 121.54391 | 54.8 |
| 4 | 5 | 2012.833 | 5.0 | 390.56840 | 5 | 24.97937 | 121.54245 | 43.1 |
| | | | | | | | | |
| 409 | 410 | 2013.000 | 13.7 | 4082.01500 | 0 | 24.94155 | 121.50381 | 15.4 |
| 410 | 411 | 2012.667 | 5.6 | 90.45606 | 9 | 24.97433 | 121.54310 | 50.0 |
| 411 | 412 | 2013.250 | 18.8 | 390.96960 | 7 | 24.97923 | 121.53986 | 40.6 |
| 412 | 413 | 2013.000 | 8.1 | 104.81010 | 5 | 24.96674 | 121.54067 | 52.5 |
| 413 | 414 | 2013.500 | 6.5 | 90.45606 | 9 | 24.97433 | 121.54310 | 63.9 |
| | | | | | | | | |

414 rows × 8 columns

```
In [100]: df = df.drop("No",axis=1)
df
```

Out[100]:

| | X1 transaction date | X2 house age | X3 distance to the nearest MRT station | X4 number of convenience stores | X5 latitude | X6 longitude | Y house price of unit area |
|-----|---------------------------|--------------------|--|---------------------------------------|----------------|-----------------|----------------------------------|
| 0 | 2012.917 | 32.0 | 84.87882 | 10 | 24.98298 | 121.54024 | 37.9 |
| 1 | 2012.917 | 19.5 | 306.59470 | 9 | 24.98034 | 121.53951 | 42.2 |
| 2 | 2013.583 | 13.3 | 561.98450 | 5 | 24.98746 | 121.54391 | 47.3 |
| 3 | 2013.500 | 13.3 | 561.98450 | 5 | 24.98746 | 121.54391 | 54.8 |
| 4 | 2012.833 | 5.0 | 390.56840 | 5 | 24.97937 | 121.54245 | 43.1 |
| | | | | | | | |
| 409 | 2013.000 | 13.7 | 4082.01500 | 0 | 24.94155 | 121.50381 | 15.4 |
| 410 | 2012.667 | 5.6 | 90.45606 | 9 | 24.97433 | 121.54310 | 50.0 |
| 411 | 2013.250 | 18.8 | 390.96960 | 7 | 24.97923 | 121.53986 | 40.6 |
| 412 | 2013.000 | 8.1 | 104.81010 | 5 | 24.96674 | 121.54067 | 52.5 |
| 413 | 2013.500 | 6.5 | 90.45606 | 9 | 24.97433 | 121.54310 | 63.9 |
| | | | | | | | |

414 rows × 7 columns

```
In [101]: y = df["Y house price of unit area"]
x = df.drop("Y house price of unit area",axis=1)
```

```
In [102]: from sklearn.preprocessing import PolynomialFeatures
    poly = PolynomialFeatures(degree=2)
    x_poly = poly.fit_transform(x)
    x_poly
```

```
Out[102]: array([[1.00000000e+00, 2.01291700e+03, 3.20000000e+01, ..., 6.24149290e+02, 3.03643739e+03, 1.47720299e+04], [1.00000000e+00, 2.01291700e+03, 1.95000000e+01, ..., 6.24017387e+02, 3.03609828e+03, 1.47718525e+04], [1.00000000e+00, 2.01358300e+03, 1.33000000e+01, ..., 6.24373157e+02, 3.03707359e+03, 1.47729221e+04], ..., [1.00000000e+00, 2.01325000e+03, 1.88000000e+01, ..., 6.23961931e+02, 3.03597212e+03, 1.47719376e+04], [1.00000000e+00, 2.01300000e+03, 8.10000000e+00, ..., 6.23338106e+02, 3.03447431e+03, 1.47721345e+04], [1.00000000e+00, 2.01350000e+03, 6.50000000e+00, ..., 6.23717159e+02, 3.03545749e+03, 1.47727252e+04]])
```

```
In [107]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test = train_test_split(x_poly,y,test_size=0.2,random)
```

```
from sklearn.linear model import LinearRegression
In [108]:
          lr = LinearRegression()
          model = lr.fit(x_train,y_train)
          model.coef
          # lr.coef
Out[108]: array([-2.30201635e+02, -1.88045458e+04, 9.28403731e+02, 2.61016682e+01,
                  1.05269906e+04, 1.00715255e+05, -4.97732053e+05, 5.31841968e+00,
                  1.41151382e-01, -1.48819301e-03, 3.58924294e-02, -1.25377084e+02,
                  4.33398978e+00, 2.23482207e-02, -8.54009205e-05, 1.05267264e-02,
                 -9.54992785e+00, -8.02419068e+00, -1.48516558e-06, -1.89678872e-03,
                 -2.89866581e-01, -1.30560016e-01, -1.33071674e-02, -1.27599274e+02,
                 -6.09794087e+01, 8.53750188e+03, -2.24954222e+03, 2.24478069e+03])
In [105]:
          # lr.intercept_
          model.intercept
Out[105]: 47867918.24478324
          y_pred = lr.predict(x_test)
          y_pred
Out[106]: array([47.737425 , 39.02338394, 22.23622888, 34.67978097, 24.77320855,
                 33.37398518, 37.99670412, 32.80283378, 42.67046385, 57.77061178,
                 45.22143111, 24.95019238, 40.79493067, 52.53829317, 50.74464512,
                 27.85910624, 37.64700218, 28.56401196, 18.76919939, 18.3504373,
                 43.32510749, 15.97532092, 39.9409853, 38.53202164, 26.94980829,
                 28.80667563, 29.62812249, 53.64500886, 15.9352848, 44.73877761,
                 55.76121162, 43.98363999, 35.69946508, 49.21053347, 29.9582713,
                 19.17678588, 49.94324894, 28.45277803, 32.23210655, 15.33233763,
                 22.01546432, 30.02814393, 52.37805844, 39.46631756, 38.20469537,
                 50.88411851, 45.81583012, 17.46285062, 45.22183378, 34.75537521,
                 31.80412726, 42.46314272, 50.5586628 , 40.45696788, 48.82210195,
                 45.57401511, 47.02995715, 8.1405787, 44.21885642, 42.22999367,
                 18.06767014, 55.08323424, 19.82581403, 47.24241443, 36.43434843,
                 47.75834337, 59.97113802, 26.3289122 , 16.66986829, 18.48499367,
                 25.28848647, 30.33666265, 21.15872747, 19.01363268, 19.32715356,
                 21.47406061, 42.43209426, 49.52050402, 47.35496488, 30.59281138,
                 19.17439804, 36.55856462, 52.7026816 ])
 In [96]:
          from sklearn import metrics
          import math
          y pred = lr.predict(x test)
          print("MAE : ",metrics.mean_absolute_error(y_test,y_pred))
          print("MSE : "
                        ',metrics.mean squared error(y test,y pred))
          print("RMSE : ",math.sqrt(metrics.mean_squared_error(y_test,y_pred)))
          print("R2score : ",metrics.r2_score(y_test,y_pred))
                 5.226824101554342
          MAE :
          MSE: 96.73983431451933
          RMSE: 9.835641022044234
          R2score: 0.5636304149032404
```

```
In [110]: import numpy as np
    test_data = np.array([54,67,69,34,14,22])
    test_data = test_data.reshape(1,6)
    test_data = poly.fit_transform(test_data)
    lr.predict(test_data)
```

Out[110]: array([39599971.72038367])