# **Business Problem - Predict the Price of Bangalore House**

Using Linear Regression - Supervised Machine Learning Algorithm

#### **Load Libraries**

```
In [1]: import pandas as pd
```

#### **Load Data**

```
In [2]: # path = r"https://drive.google.com/uc?export=download&id=1xxDtrZKfuWQfl-6KA9XEd_eatitNPnkB"
# df = pd.read_csv(path)
```

```
In [3]:
    df = pd.read_csv("bangalore house price prediction OHE-data.csv")
```

```
In [4]: df.head()
```

Out[4]:

	bath	balcony	price	total_sqft_int	bhk	price_per_sqft	area_typeSuper built-up Area	area_typeBuilt- up Area	area_typePlot Area	avail
0	3.0	2.0	150.0	1672.0	3	8971.291866	1	(	)	0
1	3.0	3.0	149.0	1750.0	3	8514.285714	0	,	1	0
2	3.0	2.0	150.0	1750.0	3	8571.428571	1	(	)	0
3	2.0	2.0	40.0	1250.0	2	3200.000000	1	(	)	0
4	2.0	2.0	83.0	1200.0	2	6916.666667	0	(	)	1

5 rows × 108 columns

∢.

## **Split Data**

```
In [5]: X = df.drop('price', axis=1)
y = df['price']

print('Shape of X = ', X.shape)
print('Shape of y = ', y.shape)

Shape of X = (7120, 107)
Shape of y = (7120,)
```

```
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, random_state=51)

print('Shape of X_train = ', X_train.shape)
print('Shape of y_train = ', y_train.shape)
print('Shape of X_test = ', X_test.shape)
print('Shape of y_test = ', y_test.shape)
Shape of X_train = (5696, 107)
Shape of y_train = (5696,)
Shape of X_test = (1424, 107)
Shape of y_test = (1424,)
```

#### **Feature Scaling**

```
In [7]: 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**

```
In [8]: from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(X_train, y_train)
```

Out[8]: LinearRegression()

```
In [9]: lr.coef
Out[9]: array([-5.70206143e+00, -1.25679916e+00, 8.27341833e+01, -1.44906911e+01,
                 5.75662723e+01, 1.88468905e-01, -1.72593897e+00, -4.51058311e+00,
                -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])
In [10]: lr.intercept
```

Out[10]: 95.0802729985955

### Predict the value of Home and Test

```
In [11]: X test[0, :]
Out[11]: 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])
In [12]: |lr.predict([X test[0, :]])
Out[12]: array([76.90661876])
In [13]: |lr.predict(X_test)
Out[13]: array([ 76.90661876, 15.25005377, 113.6828165 , ..., 21.30296864,
                 71.43462962, 230.0414626 ])
```

```
In [14]: y_test
Out[14]: 2435
                  80.00
                  40.00
         3113
         426
                 120.00
         1124
                  79.00
                  45.00
         1161
                   . . .
         2078
                  28.34
         6855
                  84.00
         4381
                  32.00
         3862
                  63.00
         43
                 180.00
         Name: price, Length: 1424, dtype: float64
In [15]: lr.score(X_test, y_test)
Out[15]: 0.7903837092682249
```

## model Evaluation

## **Root mean square Error**

```
In [20]: y_test
Out[20]: 2435
                  80.00
         3113
                  40.00
         426
                 120.00
         1124
                  79.00
         1161
                  45.00
                   . . .
                  28.34
         2078
                  84.00
         6855
                  32.00
         4381
                  63.00
         3862
         43
                 180.00
         Name: price, Length: 1424, dtype: float64
In [31]: from sklearn.metrics import mean squared error
         import numpy as np
In [33]: | mse=mean_squared_error(y_test,y_predict)
         rmse=np.sqrt(mse)
         print("MSE=",mse)
         print("RMSE=",rmse)
         MSE= 4211.806905823336
         RMSE= 64.89843531105612
In [ ]:
In [ ]:
```