

# Real Estate Price Prediction Using Machine Learning Algorithm

Shreyash Rajendra Pawar

Pratik Sanjay Alai

Nikhil Madhav Taral

Yasim Mehboob Shaikh

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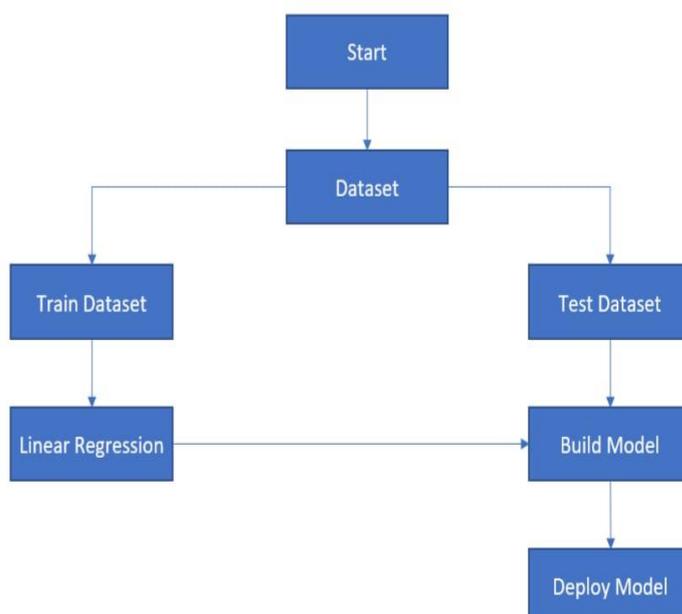
**Abstract** - Accurately forecasting prices carries immense significance across multiple industries, encompassing finance, real estate, and banking. Through the amalgamation of historical and present data, machine learning algorithms can be effectively employed to anticipate price trends. Notably, machine learning has demonstrated remarkable potential in the realm of predicting prices for diverse real estate assets, including houses and properties. By harnessing the capabilities of data and sophisticated algorithms, these models contribute to empowering individuals and enterprises to make well-informed choices while engaging in real estate transactions.

Given the significant fluctuations experienced in the real estate market, accurate price prediction has become essential for making informed decisions, whether it involves purchasing, selling, or investing in properties. Machine learning algorithms, such as linear regression, have emerged as powerful tools for real estate price prediction, capitalizing on technological advancements and the growing availability of data. Linear regression, a statistical approach that models the relationship between a dependent variable and one or more independent variables, has gained popularity as a suitable choice for regression tasks, including real estate price prediction.

## 1.INTRODUCTION

In the past, the task of finding suitable real estate properties within budgetary limitations posed considerable challenges for many individuals. This was mainly due to the time-consuming and arduous process of physically visiting each location and evaluating prices. However, the advent of machine learning models for real estate price prediction has alleviated this problem. These models leverage data for training purposes and empower buyers to access property prices conveniently through their mobile devices, from the comfort of their homes. As a result, individuals can now more easily identify and acquire properties that align with their budgetary constraints.

## System Architecture



Methodology

Linear regression is a statistical method used to establish a relationship between a dependent variable (Y) and one or more independent variables (X) by fitting a linear equation to the data. The equation of a simple linear regression model can be represented as:

$$Y = \beta_0 + \beta_1 X + \epsilon$$

where:

Y is the dependent variable (the variable you want to predict)

X is the independent variable (the variable you use to predict Y)

$\beta_0$  is the intercept (the value of Y when X = 0)

$\beta_1$  is the slope (the change in Y for every one-unit change in X)

$\epsilon$  is the error term (the difference between the predicted and actual Y value)

The goal of linear regression is to estimate the values of  $\beta_0$  and  $\beta_1$  such that the line fits the data as closely as possible. This is typically done using a technique called least squares regression, which minimizes the sum of the squared errors.

To estimate the values of  $\beta_0$  and  $\beta_1$ , you can use the following formulas:

$$\beta_1 = \frac{\sum(X_i - \bar{X})(Y_i - \bar{Y})}{\sum(X_i - \bar{X})^2}$$

$$\beta_0 = \bar{Y} - \beta_1 \bar{X}$$

where:

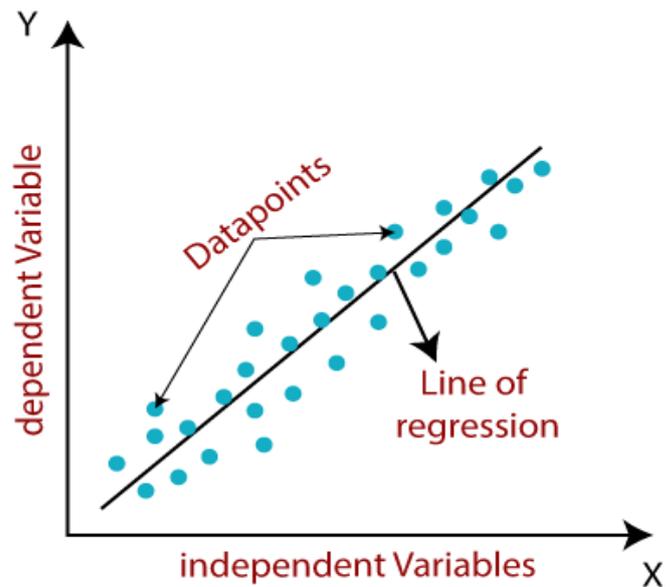
$X_i$  is the  $i$ th value of X

$\bar{X}$  is the mean of X

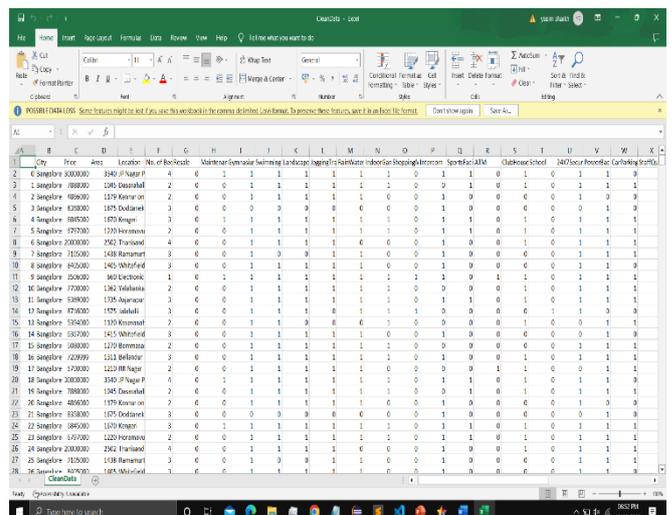
$Y_i$  is the  $i$ th value of Y

$\bar{Y}$  is the mean of Y

Linear Regression



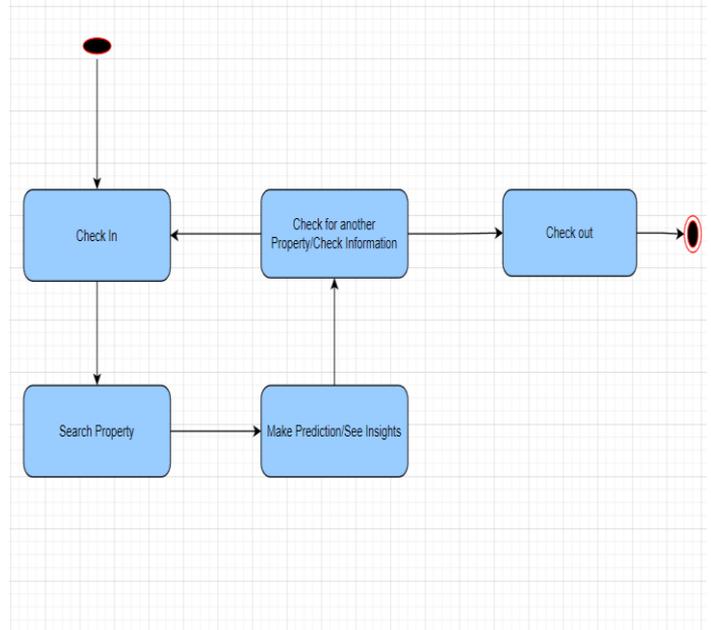
Dataset Screenshots



City	Area	Location	No. of Bookshops	Management	Services	Swimming	Landscaping	Apparel	Children	Indoor	Shopping	Interiors	Sports	Facilities	Clubhouse	Child	24/7	Year	Revenue	Car Parking	Portals
1	Singapore	3000000	3340	P Nagar P	4	0	1	1	1	1	1	1	0	3	1	0	1	1	1	1	1
2	1	Singapore	1888000	1245	Donaiahall	2	0	0	1	1	1	1	1	0	0	1	0	1	1	1	1
3	2	Singapore	1696000	1378	Kovan	2	0	0	1	1	1	1	0	0	1	0	0	1	1	1	1
4	3	Singapore	1828000	1678	Donaiahall	3	0	0	0	0	0	0	0	0	1	0	0	1	1	1	1
5	4	Singapore	1685000	1670	Kovan	3	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1
6	5	Singapore	1797000	1300	Hortensia	2	0	0	1	1	1	1	1	0	1	1	0	1	1	1	1
7	6	Singapore	2000000	2862	Parsonage	4	0	0	1	1	1	1	1	0	0	1	0	0	1	1	1
8	7	Singapore	1155000	1438	Ramanart	3	0	0	1	0	0	1	1	0	0	1	0	0	1	1	1
9	8	Singapore	1455000	1465	Whitefield	3	0	0	1	1	1	1	1	0	0	1	0	0	1	1	1
10	9	Singapore	3000000	340	Electronic	1	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1
11	10	Singapore	1720000	1262	Whitefield	2	0	0	1	1	1	1	1	0	0	0	0	1	1	1	1
12	11	Singapore	1394000	1735	Ajman	3	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
13	12	Singapore	1724000	1578	Whitefield	3	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
14	13	Singapore	1554000	1100	Kovan	2	0	0	1	1	1	1	1	0	0	0	0	1	1	1	1
15	14	Singapore	1553000	1455	Whitefield	3	0	0	1	1	1	1	1	0	0	1	0	0	1	1	1
16	15	Singapore	1888000	1270	Beremasa	2	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
17	16	Singapore	1209999	1111	Bedford	3	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
18	17	Singapore	1428000	1221	Whitefield	2	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
19	18	Singapore	2000000	3340	P Nagar P	4	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1
20	19	Singapore	1888000	1545	Donaiahall	2	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
21	20	Singapore	1400000	1170	Kovan	2	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
22	21	Singapore	1858000	1675	Donaiahall	3	0	0	0	0	0	0	0	0	1	0	0	0	1	1	1
23	22	Singapore	1845000	1470	Kovan	3	0	1	1	1	1	1	1	1	1	1	1	1	1	1	1
24	23	Singapore	1571000	1221	Hortensia	2	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
25	24	Singapore	2000000	2862	Parsonage	4	0	0	1	1	1	1	1	1	1	1	1	1	1	1	1
26	25	Singapore	1155000	1438	Ramanart	3	0	0	1	0	0	1	1	0	0	1	0	0	1	1	1
27	26	Singapore	1407000	1465	Whitefield	3	0	0	1	1	1	1	1	0	0	1	0	0	1	1	1

UML Diagrams

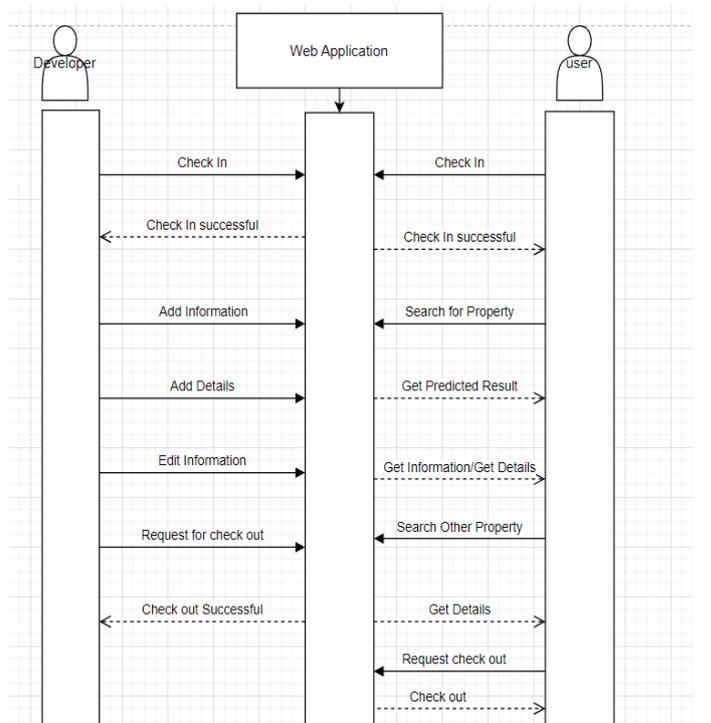
State Diagram



The main features of our dataset are city, price, Area, Location and bhk.

id	City	Price	Area	Location	bhk
0	Bangalore	6700000.0	1180.0	Bangalore 5th Phase	2.0
1	Bangalore	5800000.0	1185.0	Bangalore 5th Phase	2.0
2	Bangalore	8500000.0	1430.0	Bangalore 5th Phase	3.0
3	Bangalore	8400000.0	1180.0	Bangalore 5th Phase	2.0
4	Bangalore	4876000.0	1185.0	Bangalore 5th Phase	2.0

Sequence Diagram



### Steps

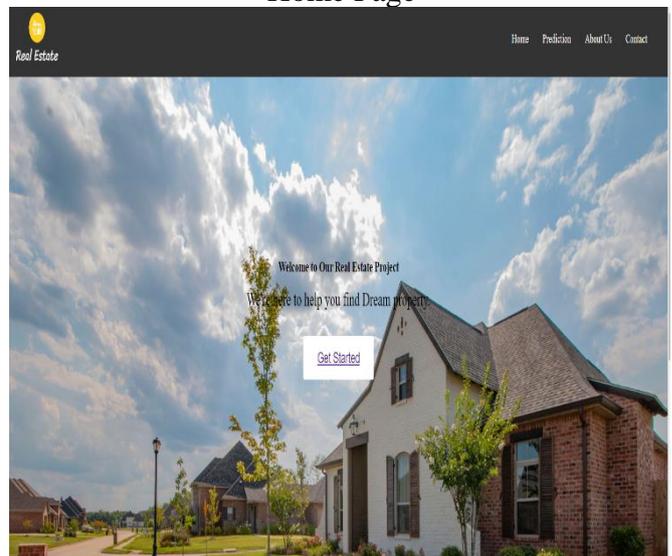
1. **Data Collection:** Gather data on the real estate market, including properties, their features, and their respective prices. This data can be collected from public sources such as real estate websites, government property listings, and real estate agents.
2. **Data Preparation:** Once data is collected, it needs to be preprocessed to remove any inconsistencies or outliers. This can be done by cleaning the data, handling missing values, and encoding categorical variables.
3. **Feature Selection:** Identify relevant features that can help predict the price of a property. These may include the location, size, number of bedrooms/bathrooms, amenities, and other factors that may affect the price.
4. **Split Data:** Split the dataset into training and testing sets. The training set is used to train the model, and the testing set is used to evaluate the model's performance.
5. **Model Training:** Apply the linear regression algorithm to the training set to train the model. The algorithm will attempt to learn the relationship between the features and the price of the property.
6. **Model Evaluation:** Evaluate the model's performance on the testing set. This can be done by computing performance metrics such as root mean squared error (RMSE), mean absolute error (MAE), and coefficient of determination (R-squared).
7. **Model Improvement:** If the model's performance is not satisfactory, try improving it by modifying the algorithm or adding more features. This process can be iterative, and multiple models can be trained

and evaluated until satisfactory results are obtained.

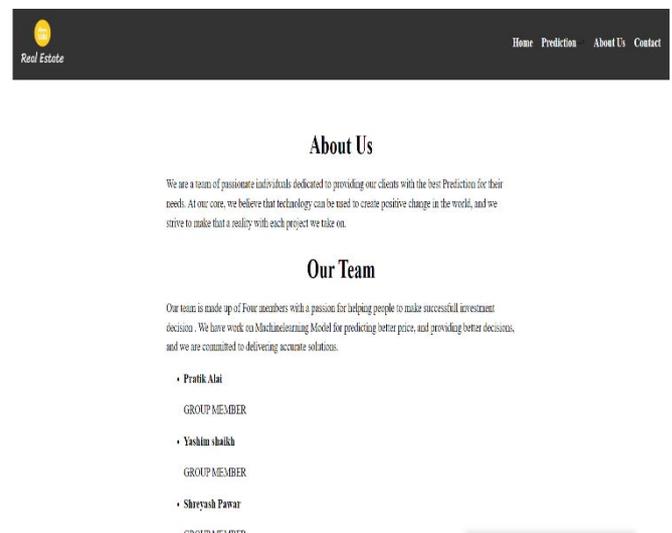
8. **Model Deployment:** Once the model is trained and validated, it can be deployed for use in predicting real estate prices. This can be done by providing the model with input data (property features), and it will output a predicted price.

### Implementation

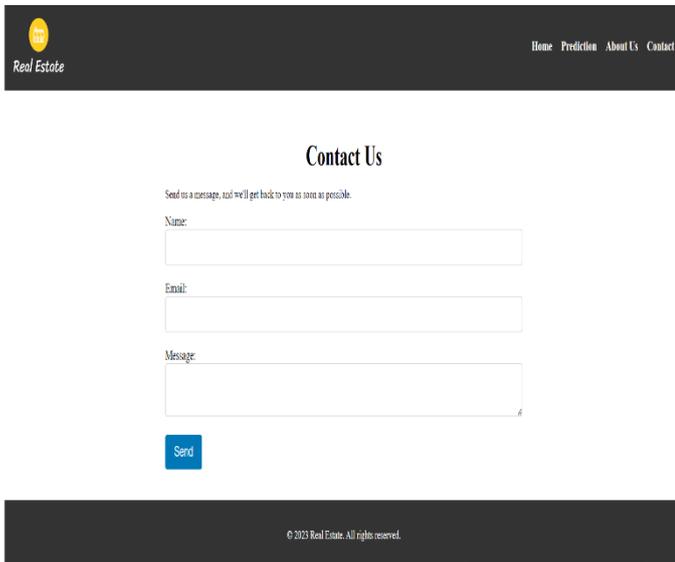
#### Home Page



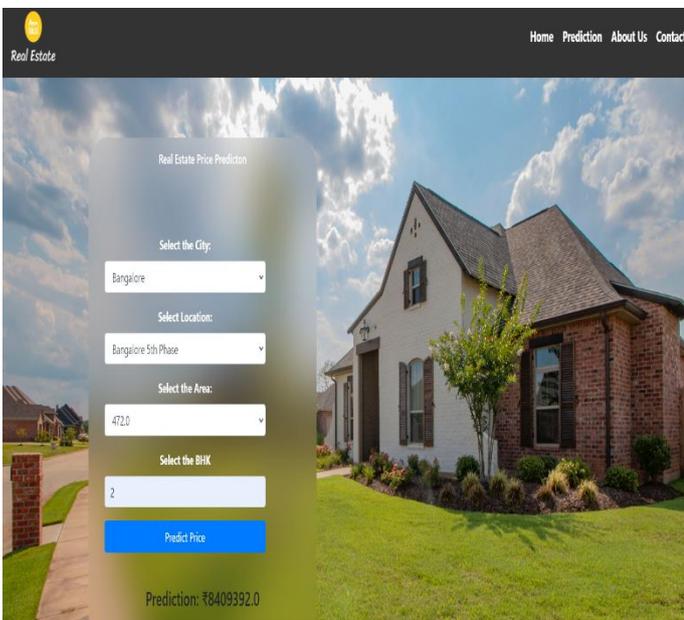
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### Prediction Page



### 3. CONCLUSION

In this study, our main focus has been on the development of a robust real estate price prediction system for multiple cities. The system we have designed enables users to conveniently search for property prices and houses in various cities, all from the comfort of their own homes.

### ACKNOWLEDGEMENT

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### REFERENCES

- [1] 1Ayush Varma, Abhijit Sarma, Sagar Doshi, Rohini Nair - "Housing Price Prediction Using Machine Learning and Neural Networks" 2018, IEEE.
- [2] G.Naga Satish, Ch.V.Raghavendran, M.D.Sugana Rao, Ch.Srinivasulu "House Price Prediction Using Machine Learning". IJITEE, 2019.
- [3] CH. Raga Madhuri, G. Anuradha, M. Vani Pujitha -" House Price Prediction Using Regression Techniques: A Comparative Study" 2019 in (ICSSS),IEEE
- [4] Sifei Lu, Zengxiang Li, Zheng Qin , Xulei Yang ,
- [5] Bharatiya, Dinesh, et al. "Stock market prediction using linear regression." *Electronics, Communication, and Aerospace Technology (ICECA), 2017 International conference of. Vol. 2. IEEE, 2017*