

# Project Proposal: Predicting House Prices using Machine Learning

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## 1. Introduction

Predicting house prices is a fundamental problem in the real estate industry with widespread applications. Accurate predictions of house prices enable both buyers and sellers to make informed decisions. In this project, we aim to leverage machine learning techniques to create a predictive model for house prices. By analyzing various features of houses such as location, square footage, number of bedrooms and bathrooms, and other relevant factors, we intend to develop a model that can provide reliable price estimates.

## **2. Problem Definition**

The primary goal of this project is to predict house prices based on a set of input features. Specifically, we seek to:

### **Input:**

A dataset containing information about houses, including features like location, square footage, bedrooms, bathrooms, and price.

### **Output:**

Develop a predictive model that can accurately estimate the prices of houses based on the provided features.

## **3. Design Thinking**

### **3.1 Data Source**

To tackle this problem, we need an appropriate dataset. We will obtain a dataset that includes relevant information about houses. This dataset should contain features like:

- Location
- Square footage
- Number of bedrooms
- Number of bathrooms
- Other relevant attributes (e.g., year built, amenities)

### **3.2 Data Preprocessing**

Before building the predictive model, we need to preprocess the data. This includes:

#### **Data Cleaning:**

Handling missing values, outliers, and errors in the dataset.

### **Feature Scaling:**

Scaling numerical features if necessary to ensure they are on the same scale.

### **Feature Encoding:**

Converting categorical features into numerical representations using techniques like one-hot encoding or label encoding.

## **3.3 Feature Selection**

Not all features in the dataset may be relevant for predicting house prices. We will perform feature selection to identify the most important features. This step can help improve model performance and reduce complexity.

## **3.4 Model Selection**

Choosing the right machine learning algorithm is crucial for accurate predictions. We will consider various regression algorithms, including:

- Linear Regression
- Random Forest Regressor
- Gradient Boosting Regressor

The selection will be based on their suitability for the problem and performance during evaluation.

## **3.5 Model Training**

Once we have selected an appropriate regression algorithm, we will train the model using the preprocessed dataset. This step involves splitting the data into training and testing sets to assess the model's performance.

### 3.6 Evaluation

We will evaluate the model's performance using the following metrics:

**Mean Absolute Error (MAE):**

Measures the average absolute difference between predicted and actual prices.

**Root Mean Squared Error (RMSE):**

Measures the square root of the mean squared differences between predicted and actual prices.

**R-squared ( $R^2$ ):**

Indicates the proportion of variance in the target variable that is predictable from the input features.

The model will be fine-tuned and iteratively improved based on the evaluation results.

### 4. Project Timeline

To effectively manage the project, we will establish a timeline with milestones for each phase, including data acquisition, preprocessing, model development, and evaluation. A detailed project plan will be created to track progress.

### 5. Conclusion

This project aims to create a robust machine learning model that can predict house prices accurately. By following the design thinking process outlined above, we will systematically address the problem, ensuring that our model is well-prepared to make reliable predictions. Accurate house price predictions can benefit various stakeholders in the real estate industry and empower individuals making informed decisions about buying or selling properties.