The plan we have outlined is a structured and comprehensive guide for implementing a machine learning model to predict house prices. Let's break down each step-in detail:

# **Step 1: Data Acquisition**

• **Objective**: Our primary objective was the acquisition of a dataset comprising pertinent information concerning residential properties.

### • Tasks:

 We found data sources like real estate websites, APIs, or public datasets. We made sure to handle data carefully and follow privacy rules.

### **Step 2: Data Preprocessing**

• **Objective**: Our objective here was the meticulous preparation of the dataset for subsequent model training.

#### Tasks:

· Data Cleaning:

We fixed missing values, errors, and odd data using different methods.

Feature Scaling:

We made sure all numbers were in a similar range.

Feature Encoding:

We turned non-number data into numbers, picking between two methods based on the data and model.

• Split Data:

We divided the data into parts to test our model.

### **Step 3: Feature Selection**

• **Objective**: Our objective in this step was the discernment of the most pivotal features influencing house price predictions.

### • Tasks:

- We chose specific features to help our model avoid making predictions that are too specific to the training data.
- We kept a record of what we picked and why, to make it clear and useful for later

### **Step 4: Model Selection**

• **Objective**: The principal aim was to discern and adopt the most suitable regression algorithm for precise predictions.

#### Tasks:

- We tried different tools and found the best one.
- We used measures like Mean Absolute Error, Root Mean Squared Error, and R-squared to check how well it worked.
- We made sure our model was good at predicting for different situations.

# **Step 5: Model Training**

• **Objective**: The goal here was the rigorous training of the selected model using the meticulously pre-processed dataset.

### Tasks:

- We tried different tools and found the best one.
- We used measures like Mean Absolute Error, Root Mean Squared Error, and R-squared to check how well it worked.
- We made sure our model was good at predicting for different situations.

# Step 6: Evaluation

• **Objective**: This step entailed the comprehensive assessment of the model's performance, accompanied by a commitment to iterative enhancements.

### Tasks:

- We kept improving the model using what we learned from checking it.
- We didn't stop making it better, so it stays useful.

# **Step 7: Project Timeline**

• **Objective**: The overarching aim was to meticulously formulate a comprehensive project plan, effectively managing progress throughout the project's lifecycle.

### • Tasks:

- We set clear goals and deadlines to keep the project on track.
- We assigned tasks to the team and used tools to keep track of progress.

# **Step 8: Documentation and Reporting**

• **Objective**: Our paramount objective in this stage was the assiduous documentation of all processes, findings, and outcomes.

# • Tasks:

- We kept detailed records to help others understand and use our work.
- We shared regular progress reports with everyone so they could follow along.