

Phase-3 Submission Report

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Institution: C. Abdul Hakeem College of Engineering and Technology

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GitHub Repository Link:

<https://github.com/fathima32/house-price-prediction-3.git>

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1. Problem Statement

Accurate prediction of house prices is a crucial challenge in the real estate industry due to the influence of numerous factors such as location, size,

amenities, and current market dynamics. Traditional models often fail to handle the non-linearity and complex interactions present in housing data, leading to suboptimal pricing insights. This project addresses the issue by applying advanced supervised regression techniques to build a robust predictive model. The aim is to support buyers, sellers, and investors with data-driven insights, thereby enhancing real estate decision-making and pricing strategies.

2. Abstract

This project focuses on predicting housing prices using smart regression models by leveraging the Ames Housing Dataset. The objective is to overcome the limitations of traditional pricing methods that often miss complex relationships in data. The dataset underwent preprocessing, exploratory data analysis, and feature engineering to improve model quality. Various models like Linear Regression, Random Forest, and XGBoost were implemented and evaluated using RMSE, MAE, and R^2 -score. Among these, XGBoost provided the most accurate predictions. The outcome is a predictive system capable of estimating house

prices, assisting stakeholders in making informed real estate decisions.

3. System Requirements

Hardware: Minimum 4GB RAM, Intel i3 Processor
or above Software:

Python 3.10+

Jupyter Notebook / Google Colab

Libraries: pandas, numpy, seaborn, matplotlib, scikit-learn, xgboost, plotly

4.Objectives

Analyze influential features like area, number of rooms, amenities, and location

Preprocess and clean the dataset for high-quality input Engineer relevant features to capture hidden patterns

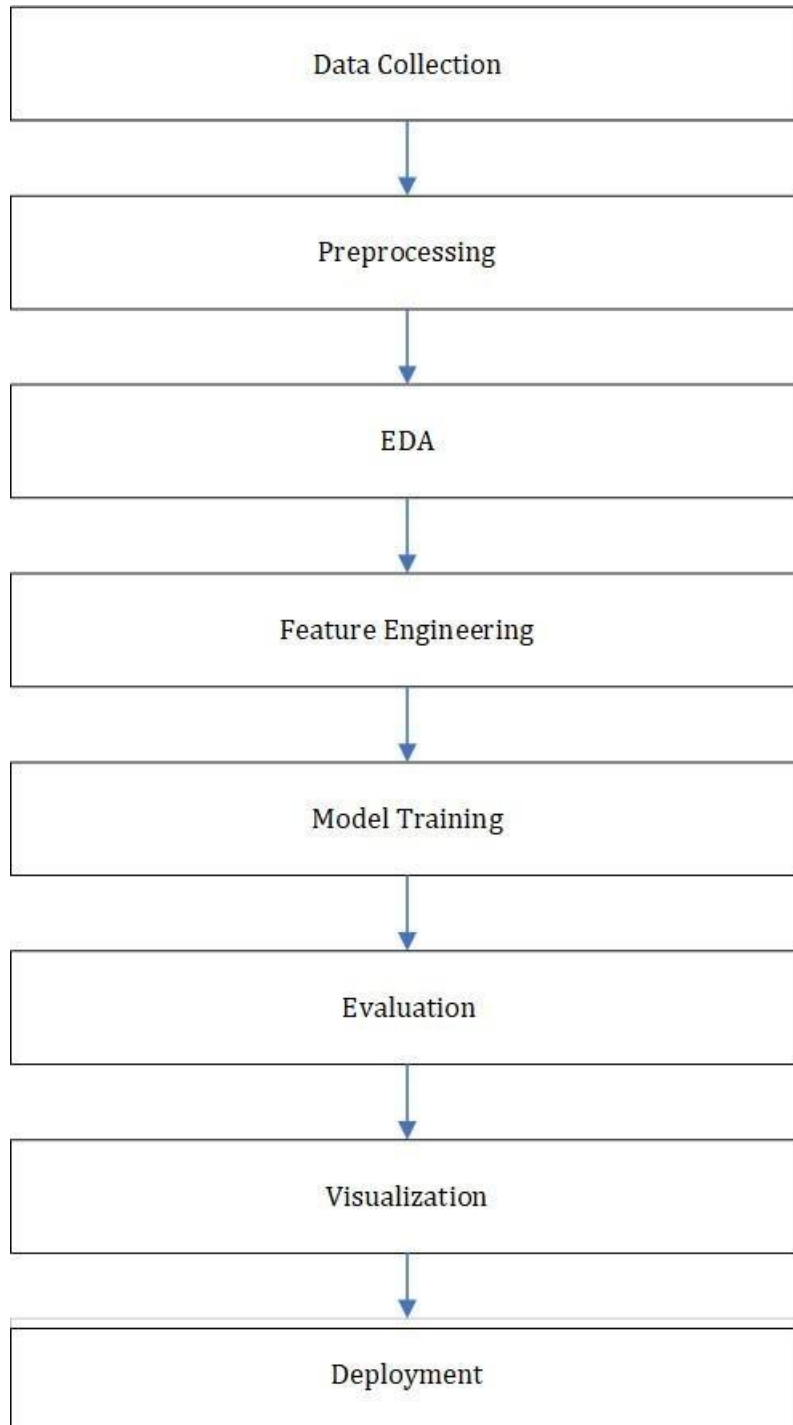
Develop and compare models: Linear Regression, Random Forest, XGBoost

Evaluate models using metrics like RMSE, MAE, and R^2

Identify the best model and present key insights using visualizations

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5.Flowchart of Project Workflow



6.Dataset Description

Dataset Name: Ames Housing Dataset

Source: Kaggle (<https://www.kaggle.com/datasets>)

Type: Public, Structured

Size: ~2,930 records with ~80 features

Target Variable: SalePrice

| | A | B | C | D | E | F |
|---|------|-------------|-------------|---|---------------|----------|
| 1 | S.No | property_id | location_id | page_url | property_type | price |
| 2 | 0 | 237062 | 3325 | https://www.zameen.com/Property/g_10_g_10_2_ground_floor_corner_apartment_with_green_lawn_for_sale-237062-3325-1.html | Flat | 10000000 |
| 3 | 1 | 346905 | 3236 | https://www.zameen.com/Property/e_11_2_services_society_flat_available_for_sale-346905-3236-1.html | Flat | 6900000 |
| 4 | 2 | 386513 | 764 | https://www.zameen.com/Property/islamabad_g_15_house_is_available_for_sale-386513-764-1.html | House | 16500000 |
| 5 | 3 | 656161 | 340 | https://www.zameen.com/Property/islamabad_bani_gala_a_rare_minimalist_concept_in_a_quiet_location-656161-340-1.html | House | 43500000 |
| 6 | 4 | 841645 | 3226 | https://www.zameen.com/Property/dha_valley_dha_homes_islamabad_dha_valley_8_marla_home_for_sale-841645-3226-1.html | House | 7000000 |
| 7 | 5 | 850762 | 3390 | https://www.zameen.com/Property/ghauri_town_ghauri_town_phase_1_house_is_available_for_sale_in_ghauri_town_phase_1-850762-3390-1.html | House | 34500000 |

7.Data Preprocessing

Handled missing values using mean/mode imputation

Removed duplicates and standardized column formats

Treated outliers using IQR method

Encoded categorical variables using One-Hot Encoding

Scaled features using Min-Max and Standard Scalers

RMSE: 25494370.485742256

R2 Score: -0.0608931929993044

| | property_type | price | location | city | province_name | \ |
|----|---------------|-----------|--------------------|-----------|-------------------|---|
| 0 | Flat | 10000000 | G-10 | Islamabad | Islamabad Capital | |
| 1 | Flat | 6900000 | E-11 | Islamabad | Islamabad Capital | |
| 2 | House | 16500000 | G-15 | Islamabad | Islamabad Capital | |
| 3 | House | 43500000 | Bani Gala | Islamabad | Islamabad Capital | |
| 4 | House | 7000000 | DHA Defence | Islamabad | Islamabad Capital | |
| 5 | House | 34500000 | Ghauri Town | Islamabad | Islamabad Capital | |
| 6 | House | 27000000 | Korang Town | Islamabad | Islamabad Capital | |
| 7 | Flat | 7800000 | E-11 | Islamabad | Islamabad Capital | |
| 8 | House | 50000000 | DHA Defence | Islamabad | Islamabad Capital | |
| 9 | Penthouse | 40000000 | F-11 | Islamabad | Islamabad Capital | |
| 10 | Flat | 35000000 | Diplomatic Enclave | Islamabad | Islamabad Capital | |
| 11 | Flat | 48000000 | Diplomatic Enclave | Islamabad | Islamabad Capital | |
| 12 | House | 400000000 | F-6 | Islamabad | Islamabad Capital | |
| 13 | Flat | 13500000 | DHA Defence | Islamabad | Islamabad Capital | |
| 14 | Flat | 3600000 | E-11 | Islamabad | Islamabad Capital | |
| 15 | Flat | 5000000 | E-11 | Islamabad | Islamabad Capital | |
| 16 | House | 19000000 | DHA Defence | Islamabad | Islamabad Capital | |
| 17 | House | 80000000 | DHA Defence | Islamabad | Islamabad Capital | |
| 18 | House | 26900000 | B-17 | Islamabad | Islamabad Capital | |
| 19 | Flat | 1750000 | PWD Housing Scheme | Islamabad | Islamabad Capital | |
| 20 | House | 55000000 | G-11 | Islamabad | Islamabad Capital | |
| 21 | House | 4500000 | Bhara kahu | Islamabad | Islamabad Capital | |
| 22 | Farm House | 88500000 | Bani Gala | Islamabad | Islamabad Capital | |
| 23 | Flat | 47000000 | Diplomatic Enclave | Islamabad | Islamabad Capital | |
| 24 | House | 4500000 | Garden Town | Islamabad | Islamabad Capital | |
| 25 | House | 6800000 | Koral Town | Islamabad | Islamabad Capital | |
| 26 | House | 20000000 | Soan Garden | Islamabad | Islamabad Capital | |
| 27 | Flat | 19400000 | Blue Area | Islamabad | Islamabad Capital | |
| 28 | House | 100000000 | F-6 | Islamabad | Islamabad Capital | |
| 29 | Flat | 8000000 | G-11 | Islamabad | Islamabad Capital | |
| 30 | Flat | 6300000 | E-11 | Islamabad | Islamabad Capital | |

8.Exploratory Data Analysis (EDA)

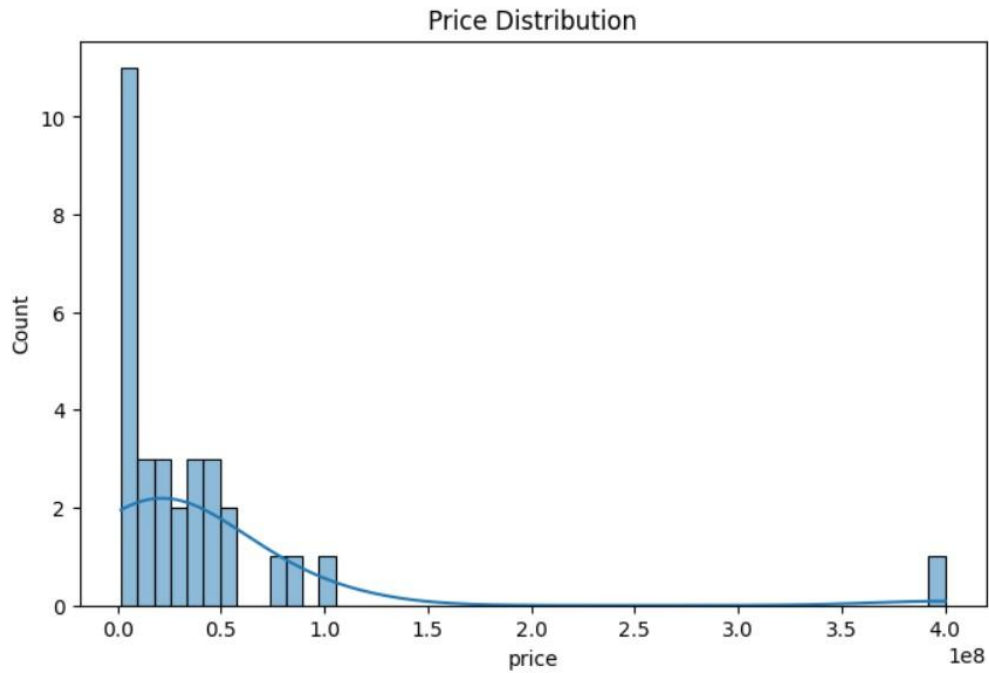
Univariate Analysis: Histograms and boxplots showed skewed distributions in price and area

Bivariate Analysis: Strong correlation between GrLivArea and SalePrice

Multivariate Analysis: Heatmaps showed multicollinearity; scatter plots revealed non-linear trends

Insights: OverallQual, GrLivArea, and Neighborhood are top influencing factors





9.Feature Engineering

Created: HouseAge = YearSold - YearBuilt,

PricePerSqFt =

SalePrice / TotalSqFeet

Encoded categorical features

Applied log transformation to reduce skewness

Removed low-importance features with high null values

10.Model Building

Models Used:

Linear Regression

Ridge & Lasso Regression

Decision Tree Regressor

Random Forest Regressor

XGBoost Regressor

Evaluation Metrics: RMSE, MAE, R²-score

Cross-Validation: 10-Fold CV

Best Model: XGBoost due to handling of non-linearity and feature interactions

RMSE: 25494370.485742256
R2 Score: -0.0608931929993044

| | property_type | price | location | city | province_name | \ |
|----|---------------|-----------|--------------------|-----------|-------------------|---|
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| 9 | Penthouse | 40000000 | F-11 | Islamabad | Islamabad Capital | |
| 10 | Flat | 35000000 | Diplomatic Enclave | Islamabad | Islamabad Capital | |
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| 12 | House | 40000000 | F-6 | Islamabad | Islamabad Capital | |
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| 14 | Flat | 3600000 | E-11 | Islamabad | Islamabad Capital | |
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| 29 | Flat | 8000000 | G-11 | Islamabad | Islamabad Capital | |
| 30 | Flat | 6300000 | E-11 | Islamabad | Islamabad Capital | |

| | latitude | longitude | baths | purpose | bedrooms | Total_Area |
|----|-----------|-----------|-------|----------|----------|------------|
| 0 | 33.679890 | 73.012640 | 2 | For Sale | 2 | 1089.004 |
| 1 | 33.700993 | 72.971492 | 3 | For Sale | 3 | 15246.056 |
| 2 | 33.631486 | 72.926559 | 6 | For Sale | 5 | 2178.008 |
| 3 | 33.707573 | 73.151199 | 4 | For Sale | 4 | 10890.000 |
| 4 | 33.492591 | 73.301339 | 3 | For Sale | 3 | 2178.008 |
| 5 | 33.623947 | 73.126588 | 8 | For Sale | 8 | 87120.000 |
| 6 | 33.579034 | 73.139591 | 8 | For Sale | 8 | 5445.000 |
| 7 | 33.698244 | 72.984238 | 2 | For Sale | 2 | 16879.562 |
| 8 | 33.540894 | 73.095732 | 7 | For Sale | 7 | 5445.000 |
| 9 | 33.679211 | 72.988787 | 5 | For Sale | 5 | 5445.000 |
| 10 | 33.728873 | 73.119628 | 3 | For Sale | 3 | 19329.821 |
| 11 | 33.728873 | 73.119628 | 2 | For Sale | 2 | 21235.578 |
| 12 | 33.731532 | 73.065696 | 0 | For Sale | 0 | 245025.000 |
| 13 | 33.538087 | 73.164536 | 5 | For Sale | 3 | 2722.510 |
| 14 | 33.698137 | 72.978215 | 1 | For Sale | 1 | 8439.781 |
| 15 | 33.698137 | 72.978215 | 2 | For Sale | 2 | 1089.004 |
| 16 | 33.508481 | 73.091826 | 3 | For Sale | 3 | 2722.510 |
| 17 | 33.541728 | 73.094103 | 7 | For Sale | 7 | 10890.000 |
| 18 | 33.694495 | 72.826653 | 6 | For Sale | 6 | 5445.000 |
| 19 | 33.570792 | 73.145256 | 0 | For Sale | 0 | 4083.765 |
| 20 | 33.671640 | 72.991655 | 7 | For Sale | 6 | 3811.514 |
| 21 | 33.737402 | 73.179159 | 3 | For Sale | 3 | 1361.255 |
| 22 | 33.713488 | 73.162680 | 3 | For Sale | 3 | 32670.000 |
| 23 | 33.728873 | 73.119628 | 2 | For Sale | 3 | 22869.084 |
| 24 | 33.636132 | 73.113921 | 4 | For Sale | 4 | 12795.797 |
| 25 | 33.602038 | 73.141966 | 4 | For Sale | 4 | 1089.004 |
| 26 | 33.569648 | 73.151522 | 5 | For Sale | 6 | 3267.012 |
| 27 | 33.713845 | 73.060970 | 1 | For Sale | 1 | 11706.793 |
| 28 | 33.724020 | 73.074524 | 5 | For Sale | 5 | 48460.678 |
| 29 | 33.675604 | 73.000367 | 2 | For Sale | 2 | 18240.817 |
| 30 | 33.698137 | 72.978215 | 3 | For Sale | 3 | 14429.303 |

11. Model Evaluation

Metrics:

RMSE: Lowest for XGBoost

MAE: Moderate error margin

R²-score: ~0.91 for XGBoost

Visuals:

Residual plots

Model comparison bar chart

SHAP values (optional)

RMSE: 25494370.485742256

R2 Score: -0.0608931929993044

12.Source Code

```
# 1. Import Libraries import pandas as pd
import numpy as np import
matplotlib.pyplot as plt import seaborn as
sns from sklearn.model_selection import
train_test_split from
sklearn.preprocessing import
StandardScaler, OneHotEncoder
from sklearn.compose import ColumnTransformer
from sklearn.pipeline import Pipeline from
sklearn.impute import SimpleImputer from
sklearn.ensemble import
RandomForestRegressor from sklearn.metrics
import mean_squared_error, r2_score
```

2. Load Dataset

```
df = pd.read_excel("Forecasting house datasets.xlsx",
sheet_name="Sheet1")
```

3. Data Cleaning

Drop

*unnecessary
columns*

```
df.drop(columns=['S.No', 'property_id', 'location_id',  
'page_url', 'agency', 'agent'], inplace=True)
```

*# Drop rows with missing target
variable*

```
df = df.dropna(subset=['price'])
```

Fill missing values

```
num_cols = df.select_dtypes(include=['float64',  
'int64']).columns
```

```
cat_cols = df.select_dtypes(include=['object']).columns
```

```
for col in num_cols:
```

```
    df[col].fillna(df[col].median(), inplace=True)
```

```
for col in cat_cols:
```

```
    df[col].fillna(df[col].mode()[0], inplace=True)
```

*# 4. EDA (Exploratory Data
Analysis)*

Plot correlations


```
plt.figure(figsize=(10, 6))
sns.heatmap(df.corr(numeric_only=True), annot=True,
            cmap='coolwarm')
plt.title('Correlation Matrix')
plt.show()
```

Plot price distribution

```
plt.figure(figsize=(8, 5))
sns.histplot(df['price'], bins=50, kde=True)
plt.title('Price Distribution')
plt.show()
```

5. Feature

Engineering X =

```
df.drop('price',
        axis=1)
y = df['price']
```

Separate features by type

```
numerical_features = X.select_dtypes(include=['int64',
```

```
'float64'])).columns.tolist()  
categorical_features =  
X.select_dtypes(include=['object']).columns.tolist()
```

6. Preprocessing Pipeline

```
numeric_transformer = Pipeline([  
    ('imputer', SimpleImputer(strategy='median')),  
    ('scaler', StandardScaler())  
)  
  
categorical_transformer = Pipeline([  
    ('imputer', SimpleImputer(strategy='most_frequent')),  
    ('onehot', OneHotEncoder(handle_unknown='ignore'))  
)  
  
preprocessor = ColumnTransformer([  
    ('num', numeric_transformer, numerical_features),  
    ('cat', categorical_transformer, categorical_features)  
)
```

7.

Modeling

model =

Pipeline([

 ('preprocessor', preprocessor),

 ('regressor', RandomForestRegressor(n_estimators=100,

random_state=42))

])

Split the data

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.2, random_state=42)

Train the model

model.fit(X_train, y_train)

Predict and

Evaluate y_pred =

model.predict(X_test

)

print("RMSE:", np.sqrt(mean_squared_error(y_test, y_pred)))

print("R2 Score:", r2_score(y_test,

y_pred)) print(df)

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-
-

13.Future Scope

Implement real-time price prediction using a Streamlit web app

Integrate more external datasets for enhanced accuracy

Use deep learning models (e.g., neural networks) for comparison

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14.Team Members and Roles

1.Jamal Be Fathima [510623104033]

Role: Team Lead & Model Building

Task: Led the project and implemented all regression models

2.Alfiya Amreen. T [510623104007]

Role: Data Collection & Preprocessing

Task: Handled dataset sourcing and cleaning

3.Farah Thasleem. S [510623104022]

Role: EDA & Feature Engineering

Task: Conducted EDA and created new features

4.Jansi Rani. K. S [510623104034]

Role: Model Evaluation & Report Preparation

Task: Evaluated models and compiled
documentation
