

Home Price Prediction Model

By
Garrett J.A. Hass



Problem Statement

- Need to predict the price of a home to assist clients
- Competitive advantage over other companies



What traits have
a high correlation with the
price?



Can we predict the price
accurately?





- ★ Realtors
- ★ Buyers
- ★ Sellers
- ★ Potential Clients of Real Estate Companies
- ★ NAR
- ★ Zillow
- ★ Redfin
- ★ Investors
- ★ Housing Market Companies
- ★ HOAs
- ★ Mortgage Companies
- ★ Banks
- ★ Real Estate Companies
- ★ Appraisers

Stakeholders



Solution to our Problem

- ❖ Create a model to predict the price of a home
- ❖ Finding variables that are correlated with price
- ❖ Identifying unique outliers
- ❖ Using four types of Models:
 - Linear Regression
 - Ridge Regression
 - Random Forest
 - XGB Regression

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The screenshot shows a Microsoft Excel spreadsheet titled "REAL ESTATE PROJECT DATABASE". The spreadsheet is filled with data organized in columns. The first column contains names, and subsequent columns contain various details such as addresses, phone numbers, and other project-related information. The data is presented in a structured, tabular format typical of a database export into Excel.



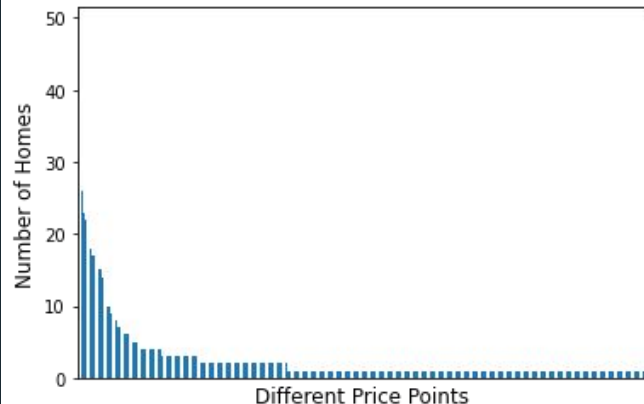
Data Wrangling

1. Started with 17 columns and 4,600 rows
2. Dropped certain rows
3. Added three columns
4. Only two NaN values
5. Dummy variables for Condition & View
6. Ended up with 25 columns and 4,600 rows
7. Target Variable is Price

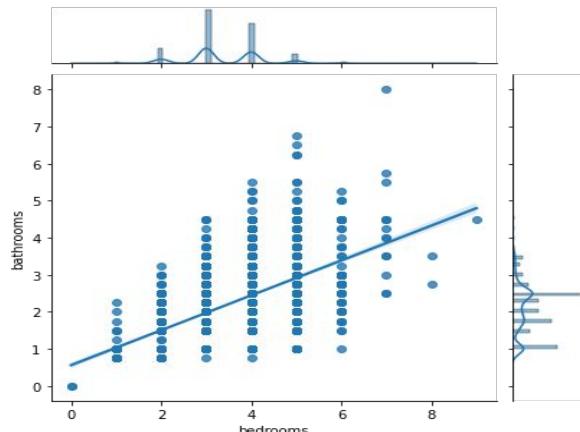


Exploratory Data Analysis

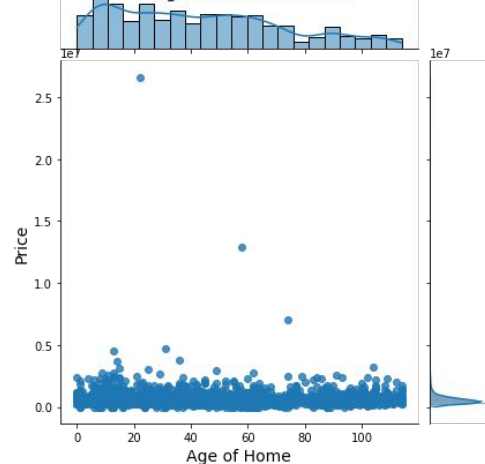
of Homes at different Price Points



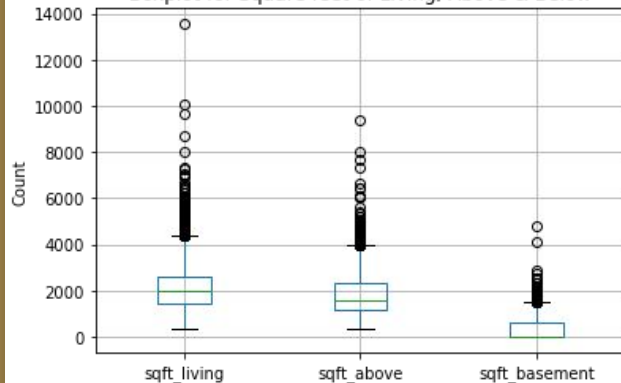
Bedrooms & Bathrooms per House



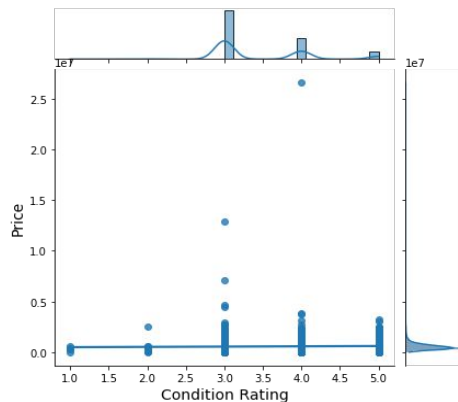
Age vs. Price of Home



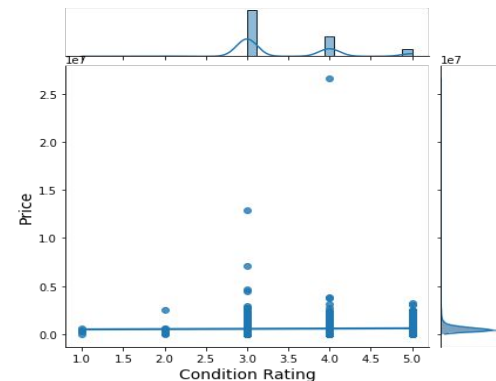
Boxplot for Square feet of Living, Above & Below



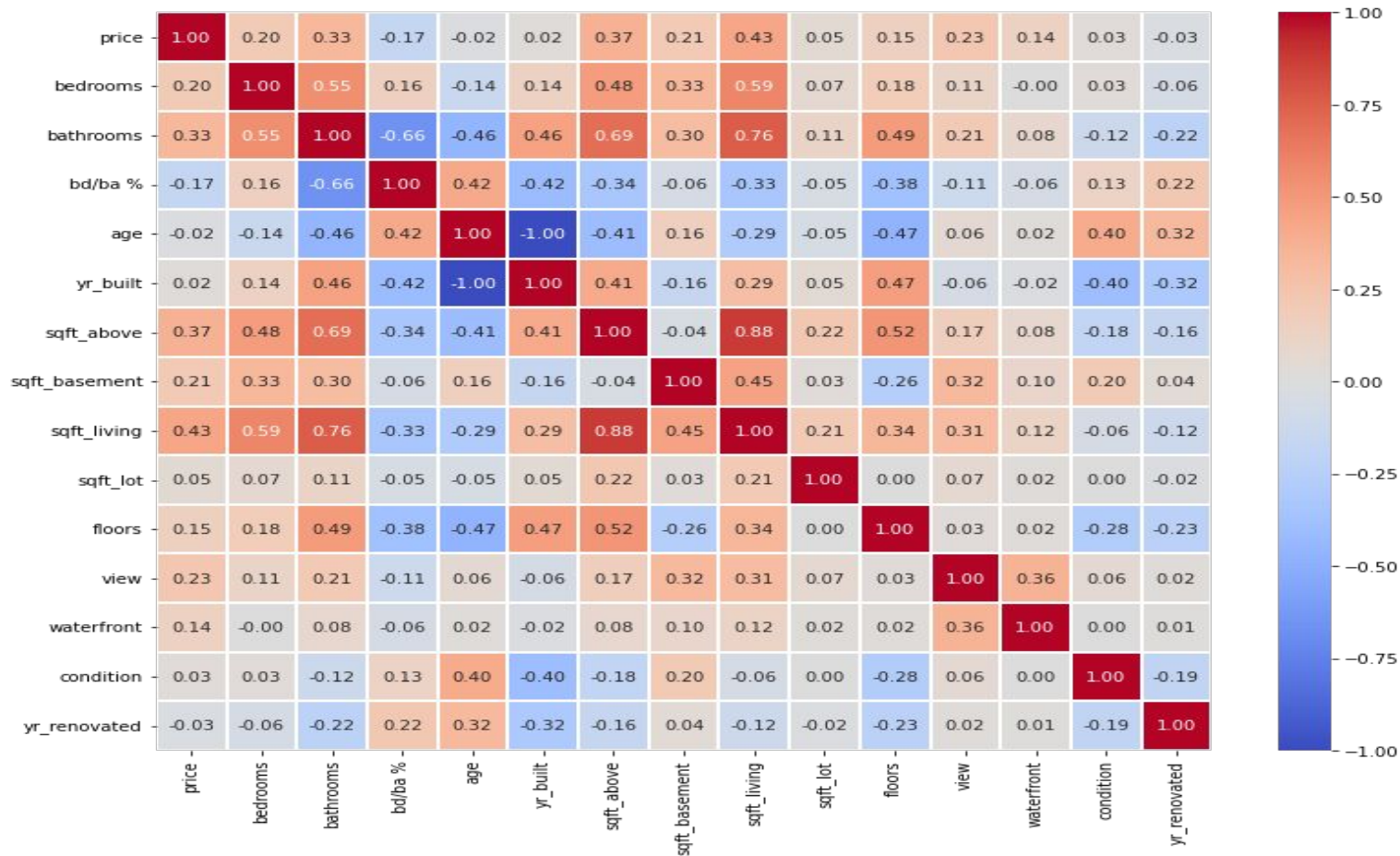
Condition Rating & Price



Condition Rating & Price



Correlations for the Housing Data

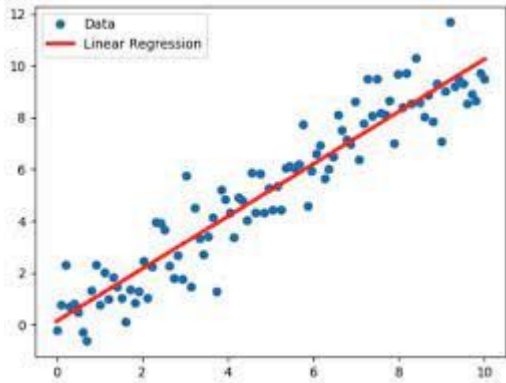


EDA Key Findings

- Many outliers for square feet of living, above and below.
- Most homes had a condition rating of 3-5.
- As # of bedrooms increased so did the # of bathrooms.
- Most of the homes had three to six bedrooms.
- Price and square feet of living space, positive correlation.
- One outlier home was under 2,000 sq ft and \$10 million.



Regression Model Selection



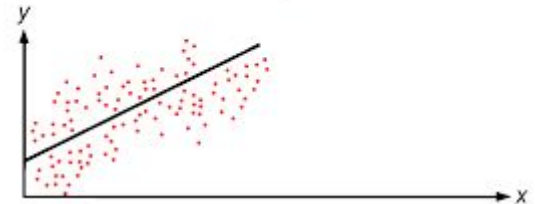
- ❖ Linear Regression
- ❖ Ridge Regression
- ❖ Random Forest Regression Model
- ❖ XGB Regression

Best Model was:

Linear Regression

- Typically best model for predicting price of real estate
- R2 score of .47
- MAE of 164,817.45

Linear Regression



Model Results Analysis

a. Linear Regression

- 1) $R^2 = .4696$
- 2) $MSE = 61,734,606,846.02$
- 3) $RMSE = 248,464.50$
- 4) $MAE = 164,817.45$

b. Ridge Regression

- 1) $R^2 = .467$
- 2) $MSE = 61,974,982,112.41$
- 3) $RMSE = 248,947.74$
- 4) $MAE = 165,587.70$

c. Random Forest Regression

- 1) $R^2 = .395$
- 2) $MSE = 70,352,537,281.89$
- 3) $RMSE = 248,947.74$
- 4) $MAE = 140,215.58$

d. XGB Regression

- 1) $R^2 = .303$
- 2) $MSE = 81,126,092,073.28$
- 3) $RMSE = 248,947.74$
- 4) $MAE = 145,701.31$

Recommendations

1) Need more data

- a) Could improve R^2 , MAE, etc. scores
- b) More X variables

2) Information to Clients

- a) Share with them our EDA findings
- b) Improve buyer morale
- c) Help sellers fix issues with their house

3) Unique Houses for Clients

- a) Our model could be a better comp than zillow
- b) Helping buyers if the house they like is over or under priced
- c) Find similar houses with certain traits





Whats next?

- Collect more data
- More variables for our model
- Improve our R^2 score
- Run models on different variables

