# Model to Predict House Prices

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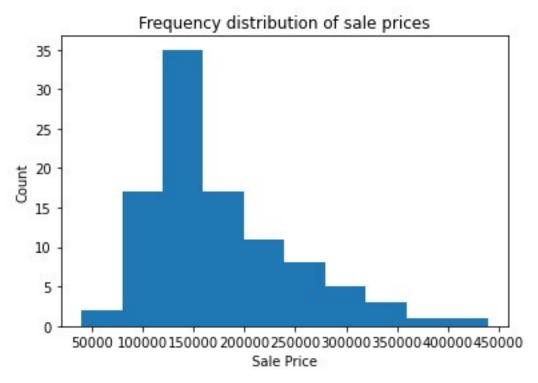
#### Introduction

The problem I am trying to solve is predicting the sale price of houses.

- 1. Analyze the dataset
- 2. Clean the data
- 3. Find the correlation of different variables against the sale price
- 4. Train a linear regression model with variables that have a high correlation to sale price
- 5. Test the different models
- 6. Find what model has the best score (the best chance of predicting the sale price)

## The Data

- The dataset to explore has 82 columns
- The sale price column has a log normal distribution with a 1.17 skew
- The graph below shows the distribution of the sale price column from the training dataset



# Data Exploration

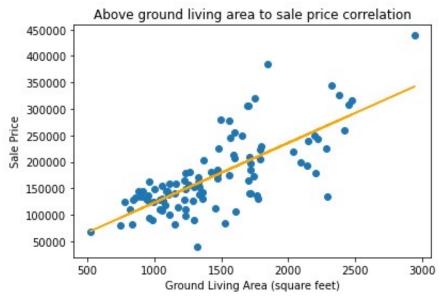
- The overall quality of the house had the strongest correlation to the sale price
- The graph below shows the relationship between the two columns



## Data Exploration

The above ground living area and exterior quality of the house had a strong positive correlation to the sale price





## Data Preparation

- 1. Convert categorical data to a numeric value
  - Convert the exterior quality rating and kitchen quality rating to a 5-star rating system
  - Excellent is equivalent to 5 and poor is equivalent to 1
- 2. Extract only numeric columns
- 3. Replace null values with the average
- 4. Drop columns with only null values

#### Correlation

- Overall quality of the house has a very strong positive correlation to the sale price of 0.86
- Exterior quality of the house and ground living area in square footage has a strong positive correlation of 0.74 to the sale price.



# **Project Description**

- The columns with the strongest correlation to the sale price are OverallQual, ExterQual, GrLivArea, GarageArea, KitchenQual, GarageCars, YearBuilt, and HeatingQC
- All columns except HeatingQC had a correlation value above 0.65
- Used the linear regression algorithm to predict the sale price column
- Used a variation of the following columns OverallQual, ExterQual, GrLivArea, GarageArea, KitchenQual, GarageCars, YearBuilt, and HeatingQC to create different linear regression models
- Compared each model against the test dataset
- Found the set of independent variables with the highest R squared value

# Analysis and Results

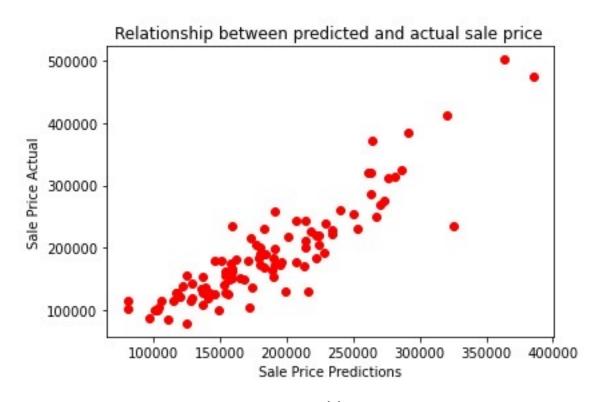
- I analyzed 6 different models with different variations for the independent variables
  - Set 1 = 'OverallQual', 'ExterQual'
  - Set 2 = 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual'
  - Set 3 = 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual', 'GarageCars'
  - Set 4 = 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual', 'GarageCars', 'YearBuilt'
  - Set 5 = 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual', 'GarageCars', 'YearBuilt', 'HeatingQC'
  - Set 6 = 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual','YearBuilt'

# Analysis and Results

- I discovered increasing the number of independent variables did not mean the models score was going to increase
- The models score on the training dataset made a huge jump from set 1 to set 2
  - Set 1 = 0.769
  - Set 2 = 0.823
- The models score went down on the test dataset for column set 4 through 6
  - Set 3 = 0.78
  - Set  $4-6 = \sim 0.75$
- The models score from column set 4 through 6 stayed consistent
  - Score in the *training* dataset = 0.84
  - Score in the *test* dataset = .75

## Verification

- The final columns used for my model were 'OverallQual', 'ExterQual', 'GrLivArea', 'GarageArea', 'KitchenQual', and 'GarageCars'
- The score of the model on the training dataset was ~0.824
- The score on the test dataset was ~0.781



• The final equation for my linear regression model to predict the sale price of a house is the following:

```
SalePrice = (17665.26545183 * OverallQual) + (25624.99501943 * ExterQual) + (41.07489011 * GrLivArea) + (79.59400665 * GarageArea) + (10179.51947658 * KitchenQual) + (-8875.89573741 * GarageCars) + - 132883.48735087246
```

- I learned that adding more independent variables to the model did not always increase the score of the model
- It sometimes decreased the models score

#### References

- Dhingra, Deepanshi. "All you need to know about your first Machine Learning model – Linear Regression." Analytics Vidhya. Data Science Blogathon, May 25, 2021. <a href="https://www.analyticsvidhya.com/blog/2021/05/all-you-need-to-know-about-your-first-machine-learning-model-linear-regression/#:~:text=In%20the%20most%20simple%20words,the%20dependent%20variable.</a>
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