

---

# How renovations may increase estimated value of homes.

Linear Regression of King County House-Sales Dataset

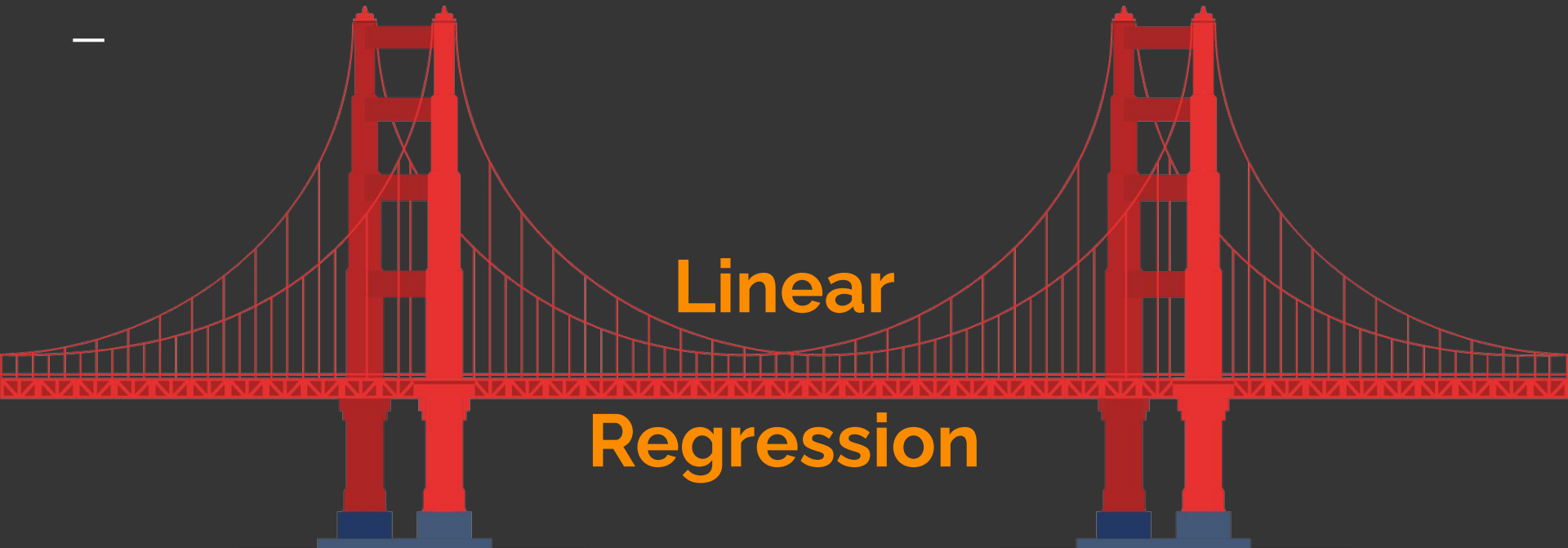
---

# Business Problem

The Real Estate Company needs to be able to advise homeowners on how to increase the value of their homes.

They need at least three features for which home renovations can accomplish this, and by how much.





## **NEED:**

Provide Advice to  
Homeowners

## **GOAL:**

3 Features that  
Increase the Value of  
the Home

# This Regression Analysis uses the King County House-Sales Dataset



## The Data

| Column        | Description  |
|---------------|--|
| id            | Unique identifier for a house  |
| date          | Date house was sold  |
| price         | Sale price (prediction target)   |
| bedrooms      | Number of bedrooms   |
| bathrooms     | Number of bathrooms  |
| sqft_living   | Square footage of living space in the home                                       |
| sqft_lot      | Square footage of the lot  |
| floors        | Number of floors (levels) in house   |
| waterfront    | Whether the house is on a waterfront   |
| view          | Quality of view from house   |
| condition     | How good the overall condition of the house is. Related to maintenance of house. |
| grade         | Overall grade of the house. Related to the construction and design of the house. |
| sqft_above    | Square footage of house apart from basement                                      |
| sqft_basement | Square footage of the basement   |
| yr_built      | Year when house was built  |
| yr_renovated  | Year when house was renovated  |
| zipcode       | ZIP Code used by the United States Postal Service                                |
| lat           | Latitude coordinate  |
| long          | Longitude coordinate   |
| sqft_living15 | The square footage of interior housing living space for the nearest 15 neighbors |
| sqft_lot15    | The square footage of the land lots of the nearest 15 neighbors                  |



Frequency

Distribution of Residuals

Normality of Residuals

QQ Plot

Frequency

Distribution of Residuals

Normality of Residuals

QQ Plot

Frequency

Distribution of Residuals

Normality of Residuals

QQ Plot

Frequency

Distribution of Residuals

Normality of Residuals

QQ Plot

Sample Quantiles

Theoretical Quantiles

$r^2$

$r^2_{adj}$

$r\_sq$ : 0.4517919301514499.  $r\_sq\_adjusted$ : 0.4517663428436389.  $k\_fold\_r$ : 0.4517066812587575

|             | coef    | std err  | t        | P> t  | [0.025 | 0.975] |
|-------------|---------|----------|----------|-------|--------|--------|
| const       | 12.2369 | 0.007    | 1871.209 | 0.000 | 12.224 | 12.250 |
| sqft_living | 0.0004  | 2.92e-06 | 132.879  | 0.000 | 0.000  | 0.000  |

# UNTIL...

# FINAL MODEL

## OLS Regression Results

Dep. Variable: price\_log  
Model: OLS  
Method: Least Squares  
Date: Tue, 28 Jun 2022  
Time: 13:51:02  
No. Observations: 21596  
Df Residuals: 21561  
Df Model: 34  
Covariance Type: nonrobust

R-squared: 0.760  
Adj. R-squared: 0.760  
F-statistic: 2011.  
Prob (F-statistic): 0.00  
Log-Likelihood: -1968.1  
AIC: 2006.  
BIC: 2086.

|                             | coef      | std err  | t       | P> t  | [0.025   | 0.975]   |
|-----------------------------|-----------|----------|---------|-------|----------|----------|
| const                       | 10.9794   | 0.020    | 548.982 | 0.000 | 10.940   | 11.019   |
| bedrooms                    | -0.0083   | 0.003    | -3.281  | 0.001 | -0.013   | -0.003   |
| bathrooms                   | 0.0145    | 0.004    | 3.623   | 0.000 | 0.007    | 0.022    |
| sqft_living                 | 0.0002    | 3.98e-06 | 48.154  | 0.000 | 3.95e-07 | 5.77e-07 |
| sqft_lot                    | 4.851e-07 | 4.64e-08 | 10.476  | 0.000 | 0.048    | 0.066    |
| floors                      | 0.0570    | 0.004    | 12.891  | 0.000 | 0.146    | 0.156    |
| grade                       | 0.1514    | 0.003    | 58.162  | 0.000 | 0.094    | 0.169    |
| view_EXCELLENT              | 0.1315    | 0.019    | 6.880   | 0.000 | 0.013    | 0.066    |
| view_GOOD                   | 0.0397    | 0.014    | 2.922   | 0.003 | -0.190   | -0.100   |
| view_NONE                   | -0.1151   | 0.020    | -6.989  | 0.000 | -0.180   | -0.101   |
| condition_Very Good         | -0.0834   | 0.004    | -19.241 | 0.000 | 0.075    | 0.092    |
| has_basement_YES            | 0.1803    | 0.007    | 26.318  | 0.000 | -0.243   | -0.054   |
| waterfront_YES              | 0.0355    | 0.004    | 8.218   | 0.000 | 0.167    | 0.194    |
| season_solid_spring         | 0.3303    | 0.026    | 12.512  | 0.000 | 0.279    | 0.382    |
| Location/Area_Bellevue Area | 0.0466    | 0.004    | 12.159  | 0.000 | 0.039    | 0.054    |
| Location/Area_Black Diamond | 0.6932    | 0.008    | 83.146  | 0.000 | 0.677    | 0.710    |
| Location/Area_Bothell Area  | 0.2726    | 0.027    | 10.231  | 0.000 | 0.220    | 0.326    |
| Location/Area_Carnation     | 0.4356    | 0.013    | 32.518  | 0.000 | 0.024    | 0.024    |
| Location/Area_Duvall        | 0.3326    | 0.024    | 13.858  | 0.000 | 0.284    | 0.382    |

Distribution of Residuals

Normality of Residuals

QQ Plot

Frequency

Sample Quantiles

Theoretical Quantiles

Residual

rs\_sq: 0.441701214938702, rs\_sq\_adjusted: 0.441674866418948, k\_fold

coef std err t P>|t| [0.025 0.975]

const 8.814e+04 3498.528 25.194 0.000 8.13e+04 9.5e+04

sqft\_living 206.0861 1.592 129.475 0.000 202.866 209.206

## OLS Regression Results

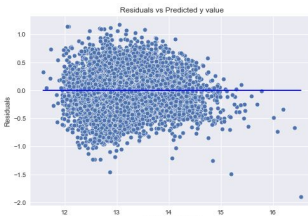
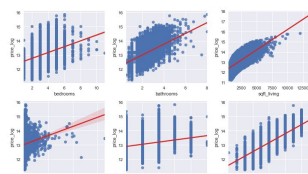
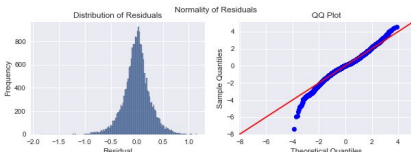
Dep. Variable: price\_log  
Model: OLS  
Method: Least Squares  
Date: Tue, 28 Jun 2022  
Time: 13:51:02  
No. Observations: 21596  
Df Residuals: 21561  
Df Model: 34  
Covariance Type: nonrobust

R-squared: 0.760  
Adj. R-squared: 0.760  
F-statistic: 2011.  
Prob (F-statistic): 0.00  
Log-Likelihood: -1968.1  
AIC: 2006.  
BIC: 2086.

|                             | coef      | std err  | t       | P> t  | [0.025   | 0.975]   |
|-----------------------------|-----------|----------|---------|-------|----------|----------|
| const                       | 10.9794   | 0.020    | 548.982 | 0.000 | 10.940   | 11.019   |
| bedrooms                    | -0.0083   | 0.003    | -3.281  | 0.001 | -0.013   | -0.003   |
| bathrooms                   | 0.0145    | 0.004    | 3.623   | 0.000 | 0.007    | 0.022    |
| sqft_living                 | 0.0002    | 3.98e-06 | 48.154  | 0.000 | 3.95e-07 | 5.77e-07 |
| sqft_lot                    | 4.851e-07 | 4.64e-08 | 10.476  | 0.000 | 0.048    | 0.066    |
| floors                      | 0.0570    | 0.004    | 12.891  | 0.000 | 0.146    | 0.156    |
| grade                       | 0.1514    | 0.003    | 58.162  | 0.000 | 0.094    | 0.169    |
| view_EXCELLENT              | 0.1315    | 0.019    | 6.880   | 0.000 | 0.013    | 0.066    |
| view_GOOD                   | 0.0397    | 0.014    | 2.922   | 0.003 | -0.190   | -0.100   |
| view_NONE                   | -0.1151   | 0.020    | -6.989  | 0.000 | -0.180   | -0.101   |
| condition_Very Good         | -0.0834   | 0.004    | -19.241 | 0.000 | 0.075    | 0.092    |
| has_basement_YES            | 0.1803    | 0.007    | 26.318  | 0.000 | -0.243   | -0.054   |
| waterfront_YES              | 0.0355    | 0.004    | 8.218   | 0.000 | 0.167    | 0.194    |
| season_solid_spring         | 0.3303    | 0.026    | 12.512  | 0.000 | 0.279    | 0.382    |
| Location/Area_Bellevue Area | 0.0466    | 0.004    | 12.159  | 0.000 | 0.039    | 0.054    |
| Location/Area_Black Diamond | 0.6932    | 0.008    | 83.146  | 0.000 | 0.677    | 0.710    |
| Location/Area_Bothell Area  | 0.2726    | 0.027    | 10.231  | 0.000 | 0.220    | 0.326    |
| Location/Area_Carnation     | 0.4356    | 0.013    | 32.518  | 0.000 | 0.024    | 0.024    |
| Location/Area_Duvall        | 0.3326    | 0.024    | 13.858  | 0.000 | 0.284    | 0.382    |

|                              |        |       |        |       |       |       |
|------------------------------|--------|-------|--------|-------|-------|-------|
| Location/Area_Sammamish Area | 0.4829 | 0.020 | 15.500 | 0.000 | 0.454 | 0.521 |
| Location/Area_Kent Area      | 0.0668 | 0.010 | 6.800  | 0.000 | 0.038 | 0.071 |
| Location/Area_Rainy Ridge    | 0.1912 | 0.010 | 19.500 | 0.000 | 0.167 | 0.206 |
| Location/Area_Medina         | 0.2126 | 0.037 | 32.517 | 0.000 | 0.139 | 0.286 |
| Location/Area_Mercer Island  | 0.8252 | 0.017 | 48.783 | 0.000 | 0.780 | 0.858 |
| Location/Area_Newcastle Area | 0.2514 | 0.007 | 27.856 | 0.000 | 0.234 | 0.269 |
| Location/Area_North Bend     | 0.3397 | 0.019 | 18.270 | 0.000 | 0.303 | 0.376 |
| Location/Area_Sammamish Area | 0.5556 | 0.006 | 66.889 | 0.000 | 0.538 | 0.572 |
| Location/Area_Sammamish Area | 0.5544 | 0.007 | 78.612 | 0.000 | 0.541 | 0.568 |
| Location/Area_Snoqualmie     | 0.4142 | 0.016 | 25.891 | 0.000 | 0.382 | 0.446 |
| Location/Area_Washou         | 0.3660 | 0.020 | 18.443 | 0.000 | 0.316 | 0.416 |
| Location/Area_Woodinville    | 0.4721 | 0.014 | 34.728 | 0.000 | 0.445 | 0.499 |

Omnibus: 789.100 Durbin-Watson: 2.004  
Prob(Omnibus): 0.000 Jarque-Bera (JB): 1990.545  
Skew: -0.151 Prob(R^2): 0.00  
Kurtosis: 4.458 Cond. No. 1.21e+06



OLS Regression Results

Dep. Variable: price  
Model: OLS  
Method: Least Squares  
Date: Tue, 28 Jun 2022  
Time: 13:49:30  
No. Observations: 21596  
Df Residuals: 21561  
Df Model: 21594  
Covariance Type: nonrobust

R-squared: 0.493  
Adj. R-squared: 0.493  
F-statistic: 2.097e+04  
Prob (F-statistic): 0.00  
Log-Likelihood: -3.000e+05  
AIC: 6.001e+05  
BIC: 6.001e+05

coef std err t P>|t| [0.025 0.975]

const -4.401e+04 4410.123 -9.980 0.000 -5.284e+04 -3.518e+04

sqft\_living 280.8688 1.929 144.820 0.000 276.980 284.758

Omnibus: 14801.482 Durbin-Watson: 0.000 Jarque-Bera (JB): 2.820 Prob(Omnibus): 0.000 Kurtosis: 26.901

Almost there

Distribution of Residuals

Normality of Residuals

QQ Plot

Frequency

Sample Quantiles

Theoretical Quantiles

Residual

Location/Area\_Sammamish Area

Location/Area\_Snoqualmie

Location/Area\_Washou

Location/Area\_Woodinville

Omnibus: 789.100 Durbin-Watson: 2.004  
Prob(Omnibus): 0.000 Jarque-Bera (JB): 1990.545  
Skew: -0.151 Prob(R^2): 0.00  
Kurtosis: 4.458 Cond. No. 1.21e+06

Drop 'sqft\_above'

# Findings & Recommendations

## CONDITION:

### Upgrade

From average or below to  
VERY GOOD, it will increase home  
value by 8.3%

To VERY GOOD, it will increase  
17.87%

## Renovate to GRADE:

For every increase in GRADE  
rating, the home value will  
increase by 14.71%

## FLOOR:

### Add a FLOOR

For every additional floor, home  
values increase 5.89%



## Grade:

Related to the construction and design of the house.  
Classification by construction quality which refers to the types of materials used and the quality of workmanship



## Condition:

Related to maintenance of the house



# Next Steps:

## Remove outliers

Starting with 'sqft\_lot' remove outliers from features one by one

## Find price per square foot of living

Trying to group homes with similar homes, improving the relevancy of recommendations

## Find bathrooms and/or bedrooms per square foot of living

Weeding out anomalies

Building on current model

Exploring new possibilities

## Change CONDITION to binary

Add 'condition\_above\_average' to make it more interpretable

## Divide homes by Urban vs Rural

Do Rural home buyers value different features compared to City buyers?

## Compare square foot lot to square foot living

---

---

# Thank You

**Email:** [cassigroesbeck@emailplace.com](mailto:cassigroesbeck@emailplace.com)

**GitHub:** [@AgathaZareth](#)

**LinkedIn:** [linkedin.com/in/cassarra-groesbeck-a64b75229](https://www.linkedin.com/in/cassarra-groesbeck-a64b75229)

---