

Pricing Homes in King County

March 2023



Business Problem and Overview

Property values often fluctuate depending on certain features, which can make pricing a home difficult. The goal is to derive a **model that helps the general public determine how a certain property should sell compared to the average home.** The model should act as a **guide for property sellers in King County** so they can pinpoint attributes of a home to emphasize or amend when justifying and setting prices.



The Data

- ❖ To create a regression model, public data on homes sold is obtained from King County's government [website](#)
- ❖ The raw data contains 24 columns, but for the purpose of our regression we will focus on the variables listed to the right



"sqft_living": Square footage of living space in the home



"bathrooms": Number of bathrooms



"grade": Overall grade of the house (construction/design)

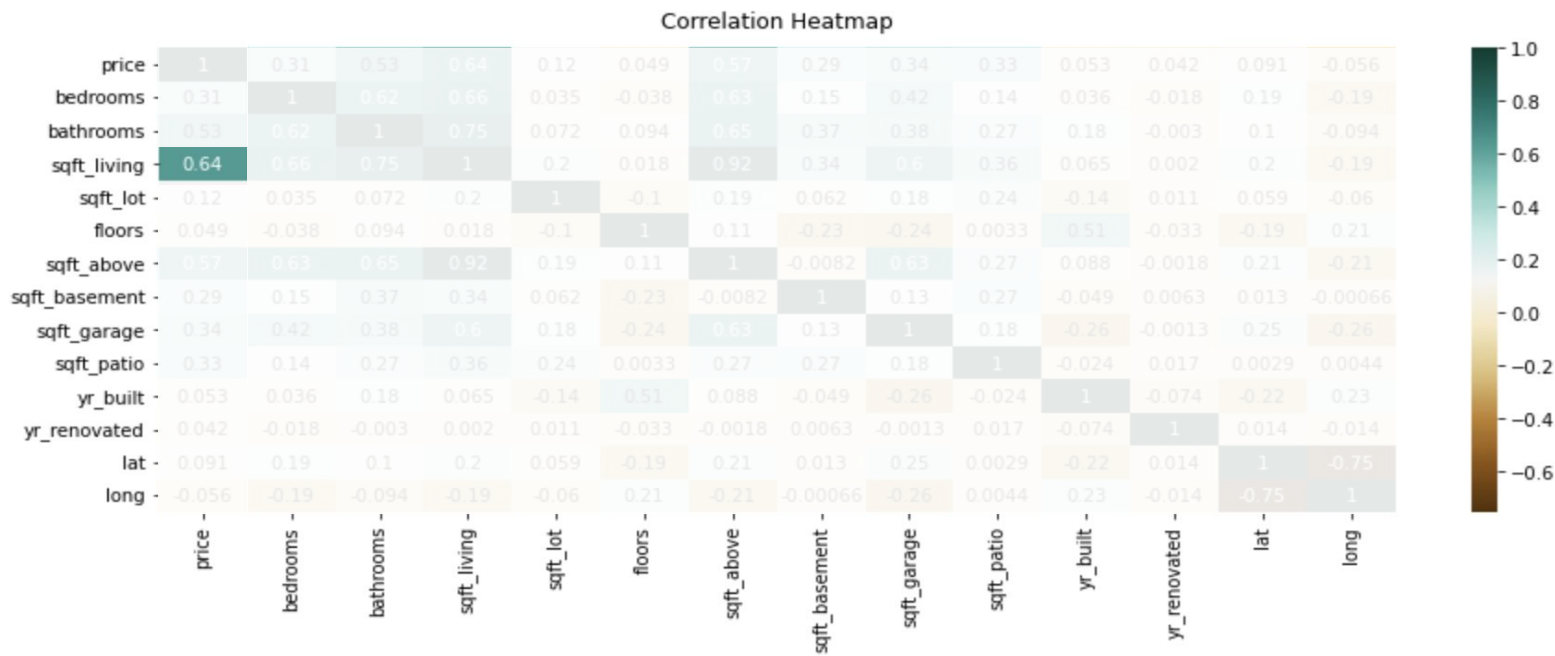


"waterfront": Whether the house is on a waterfront



"view": Quality of view from the house

Results: Initial Analysis and Baseline Model



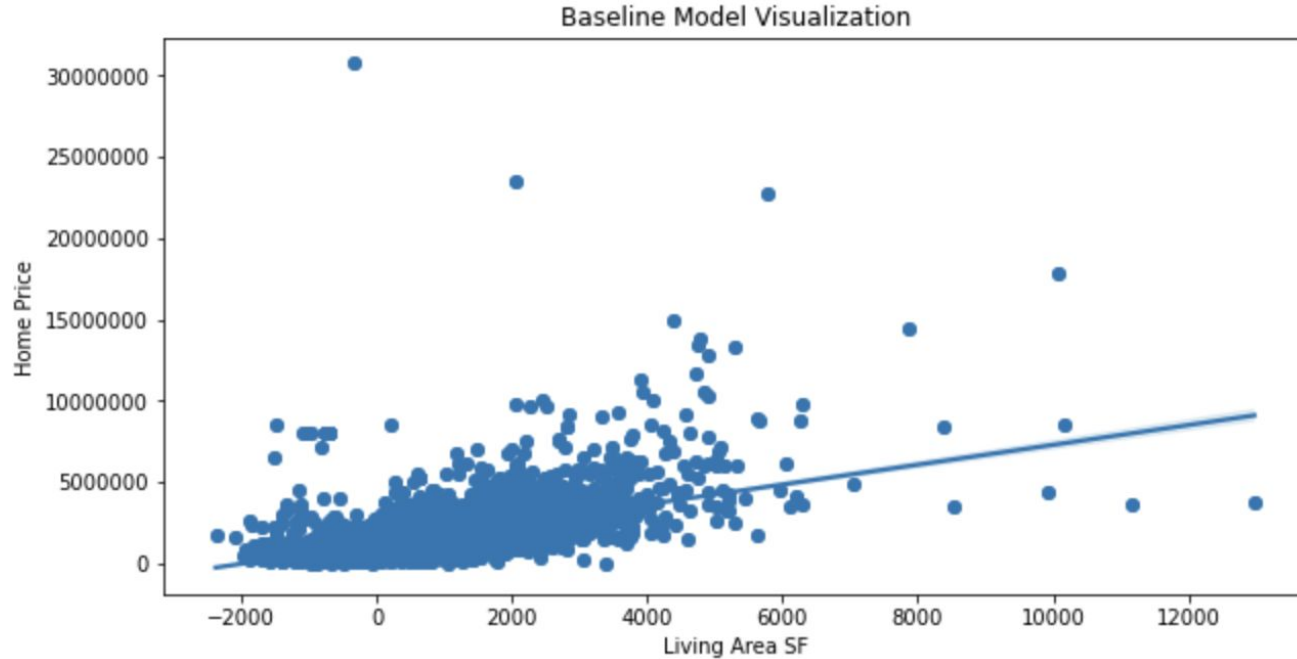
A seaborn heatmap shows that “**sqft_living**” is the **most correlated variable with price**

OLS Regression Results

Dep. Variable:	price	R-squared:	0.405			
Model:	OLS	Adj. R-squared:	0.405			
Method:	Least Squares	F-statistic:	9616.			
Date:	Mon, 06 Mar 2023	Prob (F-statistic):	0.00			
Time:	21:18:15	Log-Likelihood:	-2.1177e+05			
No. Observations:	14126	AIC:	4.235e+05			
Df Residuals:	14124	BIC:	4.236e+05			
Df Model:	1					
Covariance Type:	nonrobust					
=====						
	coef	std err	t	P> t	[0.025	0.975]

const	1.223e+06	6600.198	185.237	0.000	1.21e+06	1.24e+06
sqft_living	609.2048	6.213	98.059	0.000	597.027	621.382
=====						
Omnibus:	21552.244	Durbin-Watson:	1.806			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	31151969.336			
Skew:	9.101	Prob(JB):	0.00			
Kurtosis:	232.338	Cond. No.	1.06e+03			
=====						

The baseline model is statistically significant and accounts for 40.5% of price variance. With a unit increase in living area, we can expect price to rise by \$609 and a home of average area to sell for ~\$1.22M



The baseline model is statistically significant and accounts for **40.5%** of price variance. With a unit increase in living area, we can expect price to rise by **\$609**.

Results: Overview of Processes

Model 2

- Adding and standardizing the discrete variable (“bathrooms”) to create a multiple regression model

Model 3

- Re-formatting and standardizing “grade” data to transform it from a text into numbers the model can recognize

Model 4

- Incorporating categorical variables (“view” and “waterfront”) to assess the impact that amenities have on average properties



Results: Summary of Models Run

	Model	Independent Variables	R-squared
0	Baseline Model	sqft_living	0.408050
1	Second Model	sqft_living, bathrooms	0.412473
2	Third Model	sqft_living, bathrooms, grade	0.470337
3	Fourth Model	sqft_living, bathrooms, grade, waterfront, view	0.510179

Recall that “independent variables” refers to inputs and “R-squared” tells us the percentage of price variance that the model can account for. **As “R-squared” increases, our models can explain a greater proportion of the price data.**

Results: Final Regression Model

```
=====
                        OLS Regression Results
=====
Dep. Variable:          price      R-squared:          0.524
Model:                  OLS        Adj. R-squared:       0.524
Method:                 Least Squares  F-statistic:        1942.
Date:                   Mon, 13 Mar 2023  Prob (F-statistic):    0.00
Time:                   07:20:33      Log-Likelihood:      -2.1020e+05
No. Observations:      14126         AIC:                4.204e+05
Df Residuals:          14117         BIC:                4.205e+05
Df Model:              8
Covariance Type:       nonrobust
=====
                        coef      std err      t      P>|t|      [0.025      0.975]
-----
const                1.162e+06    2.29e+04    50.775    0.000    1.12e+06    1.21e+06
sqft_living           269.3194      10.008     26.910    0.000    249.702    288.937
bathrooms             1.301e+05    1.16e+04    11.255    0.000    1.07e+05    1.53e+05
grade                2.961e+05    7637.279     38.772    0.000    2.81e+05    3.11e+05
waterfront_YES       4.139e+05    7.02e+04     5.893    0.000    2.76e+05    5.52e+05
view_EXCELLENT       8.172e+05    5.94e+04    13.766    0.000    7.01e+05    9.34e+05
view_FAIR             3.001e+05    8.98e+04     3.342    0.001    1.24e+05    4.76e+05
view_NONE            -7.401e+04    2.35e+04     -3.156    0.002    -1.2e+05    -2.8e+04
sqft_living x waterfront_YES  629.9451     31.741     19.846    0.000    567.728    692.162
=====
Omnibus:              21308.665    Durbin-Watson:      1.773
Prob(Omnibus):        0.000      Jarque-Bera (JB):    33610538.388
Skew:                 8.842      Prob(JB):            0.00
Kurtosis:             241.309    Cond. No.            1.63e+04
=====
```

The final model experiments with the relationship between the inputs in the fourth model, and interpretations of the numbers highlighted above are outlined in the following slide

Results: Final Regression Model (Data Interpretation)

- ❖ The model accounts for **52.4% of the variance in sale price** and models against a reference home with:
 - Average living area
 - Average number of bathrooms
 - Average grade
 - Average views
 - No waterfront
- ❖ The model prices the **typical home** with the aforementioned features at **~\$1.2 Million**
- ❖ We expect a unit increase in "sqft_living" to raise the value of an average home by \$269
- ❖ We expect a unit increase in "grade" to raise the value of an average home by ~\$300K
- ❖ Adjusting variables within a home with "nice-to-have" amenities, such as a waterfront, has a greater impact on value
 - A unit increase in "sqft_living" for an average-sized home with a waterfront adds ~\$629 instead of \$269

Takeaway and Recommendation

Homesellers interested in maximizing the value of their properties should aim to **include at least one extra “nice-to-have” amenity** (e.g. waterfront, excellent views), as additional features can greatly elevate values of the average property



Future Considerations

- ❖ Map geographic distribution of homes with “lat” and “long”
- ❖ Leverage other public data from King County’s website such as population and socioeconomic information





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