

Predicting Home Prices

King County, Washington





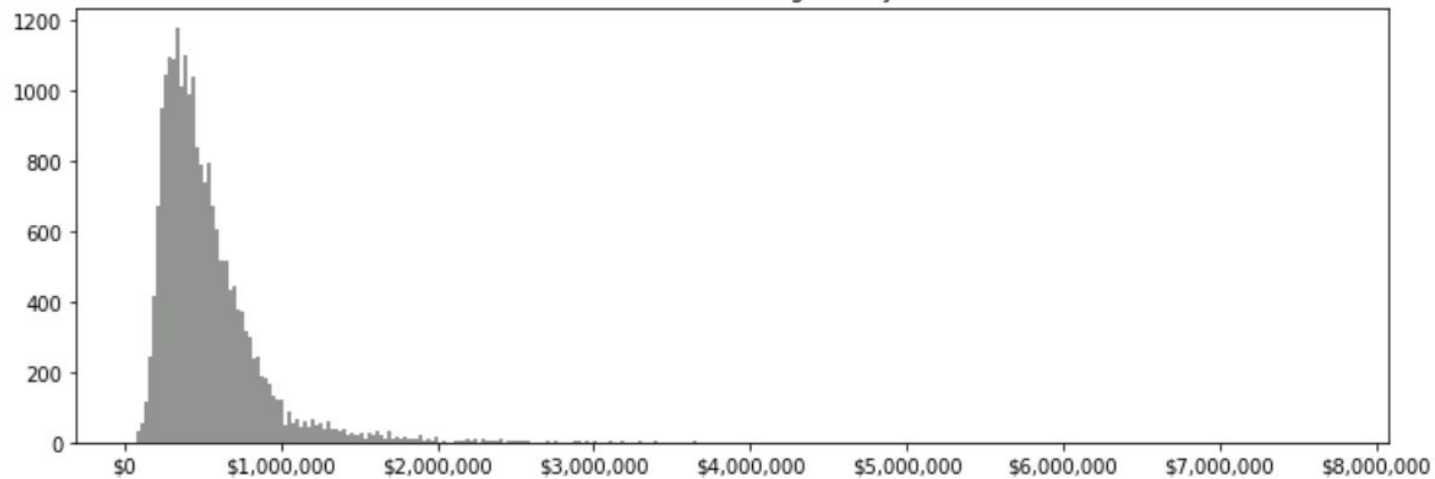
- What are the greatest influencers in the pricing of homes?
- Explore data visually
- Create linear regression model to predict home prices

The Dataset

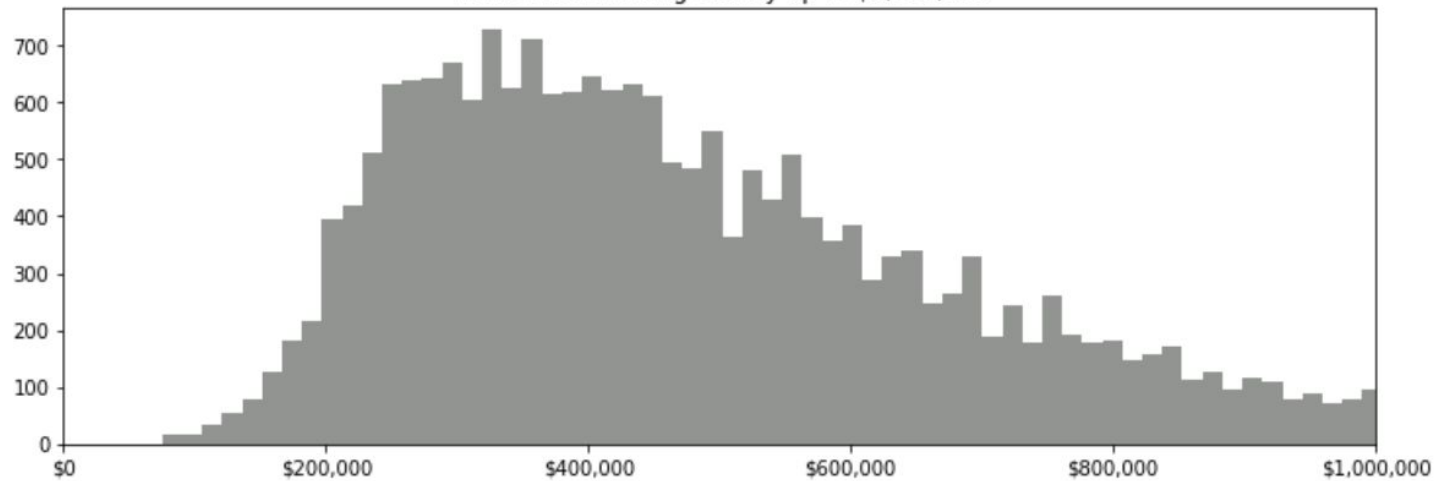
The dataset was made available on Kaggle.com by user harlfoxem. The data is clean and ready for analysis.

<https://www.kaggle.com/harlfoxem/housesalesprediction/data>

Home Prices in King County



Home Prices in King County up to \$1,000,000

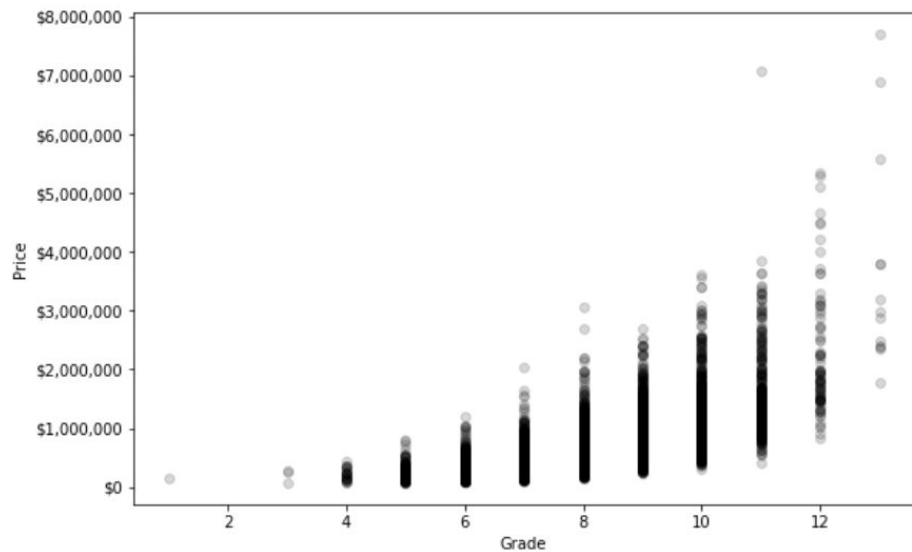
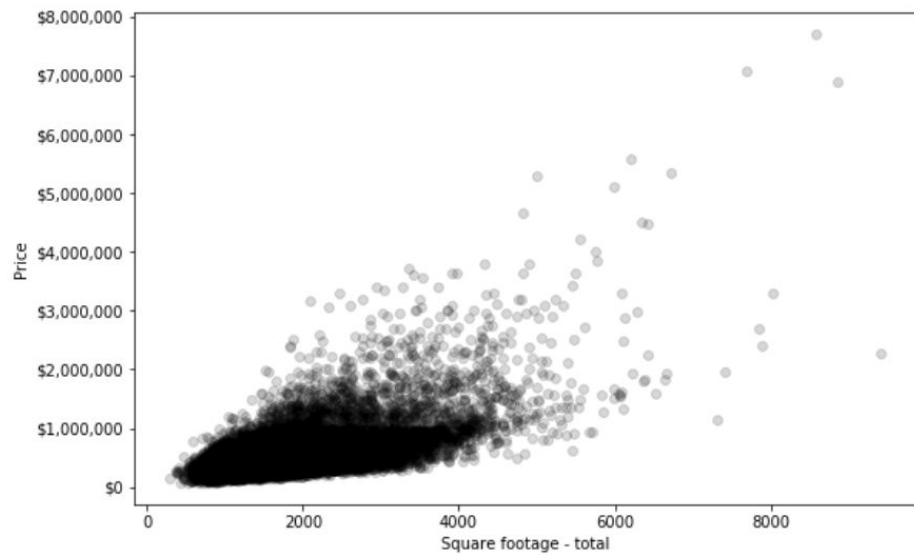


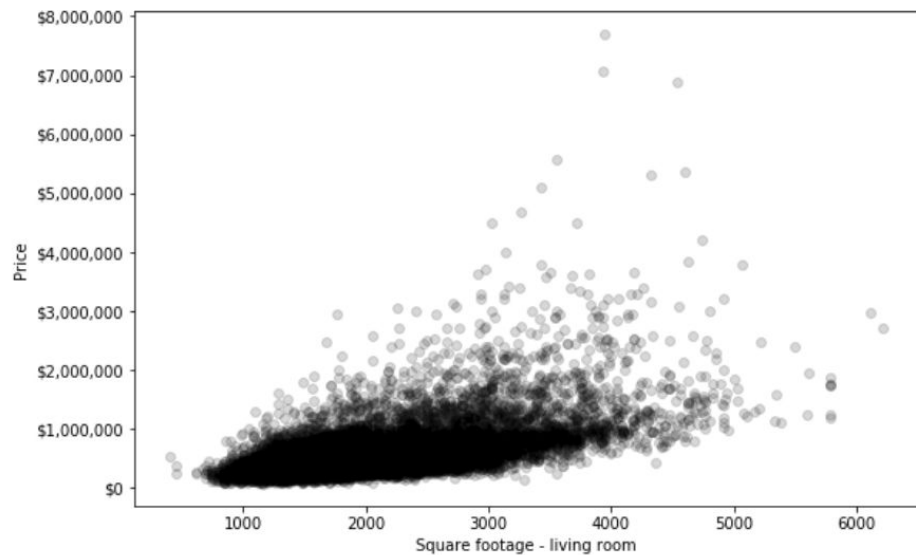
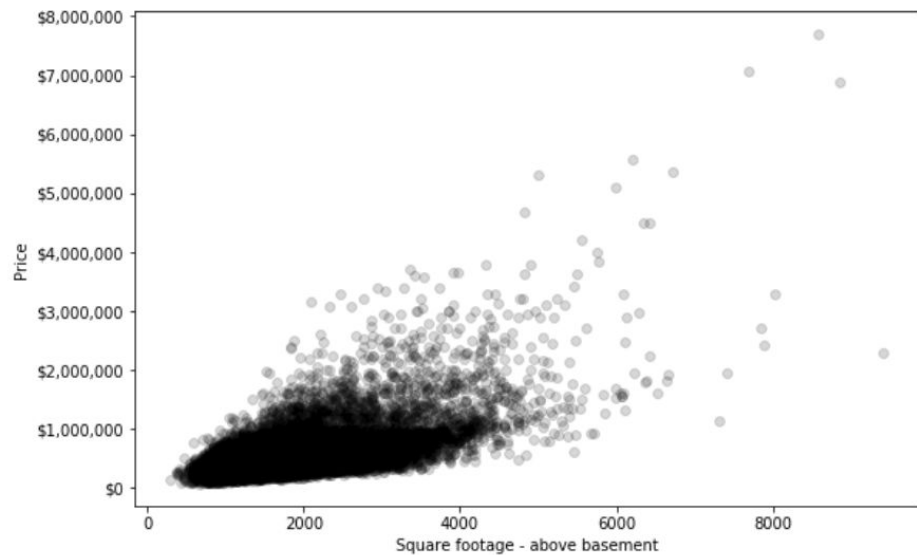
id	1	-0.017	0.0013	0.0052	-0.012	-0.13	0.019	-0.0027	0.012	-0.024	0.0081	-0.011	-0.0052	0.021	-0.017	-0.0082	-0.0019	0.021	-0.0029	-0.14
price	-0.017	1	0.31	0.53	0.7	0.09	0.26	0.27	0.4	0.036	0.67	0.61	0.32	0.054	0.13	-0.053	0.31	0.022	0.59	0.082
bedrooms	0.0013	0.31	1	0.52	0.58	0.032	0.18	-0.0066	0.08	0.028	0.36	0.48	0.3	0.15	0.019	-0.15	-0.0089	0.13	0.39	0.029
bathrooms	0.0052	0.53	0.52	1	0.75	0.088	0.5	0.064	0.19	-0.12	0.66	0.69	0.28	0.51	0.051	-0.2	0.025	0.22	0.57	0.087
sqft_living	-0.012	0.7	0.58	0.75	1	0.17	0.35	0.1	0.28	-0.059	0.76	0.88	0.44	0.32	0.055	-0.2	0.053	0.24	0.76	0.18
sqft_lot	-0.13	0.09	0.032	0.088	0.17	1	-0.0052	0.022	0.075	-0.009	0.11	0.18	0.015	0.053	0.0076	-0.13	-0.086	0.23	0.14	0.72
floors	-0.019	0.26	0.18	0.5	0.35	-0.0052	1	0.024	0.029	-0.26	0.46	0.52	-0.25	0.49	0.0063	-0.059	0.05	0.13	0.28	-0.011
waterfront	-0.0027	0.27	-0.0066	0.064	0.1	0.022	0.024	1	0.4	0.017	0.083	0.072	0.081	-0.026	0.093	0.03	-0.014	-0.042	0.086	0.031
view	0.012	0.4	0.08	0.19	0.28	0.075	0.029	0.4	1	0.046	0.25	0.17	0.28	-0.053	0.1	0.085	0.0062	-0.078	0.28	0.073
condition	-0.024	0.036	0.028	-0.12	-0.059	-0.009	-0.26	0.017	0.046	1	-0.14	-0.16	0.17	-0.36	-0.061	0.003	-0.015	-0.11	-0.093	-0.0034
grade	0.0081	0.67	0.36	0.66	0.76	0.11	0.46	0.083	0.25	-0.14	1	0.76	0.17	0.45	0.014	-0.18	0.11	0.2	0.71	0.12
sqft_above	-0.011	0.61	0.48	0.69	0.88	0.18	0.52	0.072	0.17	-0.16	0.76	1	-0.052	0.42	0.023	-0.26	-0.0082	0.34	0.73	0.19
sqft_basement	-0.0052	0.32	0.3	0.28	0.44	0.015	-0.25	0.081	0.28	0.17	0.17	-0.052	1	-0.13	0.071	0.075	0.11	-0.14	0.2	0.017
yr_built	0.021	0.054	0.15	0.51	0.32	0.053	0.49	-0.026	-0.053	-0.36	0.45	0.42	-0.13	1	-0.22	-0.35	-0.15	0.41	0.33	0.071
yr_renovated	-0.017	0.13	0.019	0.051	0.055	0.0076	0.0063	0.093	0.1	-0.061	0.014	0.023	0.071	-0.22	1	0.064	0.029	-0.068	-0.0027	0.0079
zipcode	-0.0082	-0.053	-0.15	-0.2	-0.2	-0.13	-0.059	0.03	0.085	0.003	-0.18	-0.26	0.075	-0.35	0.064	1	0.27	-0.56	-0.28	-0.15
lat	-0.0019	0.31	-0.0089	0.025	0.053	-0.086	0.05	-0.014	0.0062	-0.015	0.11	-0.00082	0.11	-0.15	0.029	0.27	1	-0.14	0.049	-0.086
long	-0.021	-0.022	0.13	0.22	0.24	0.23	0.13	-0.042	-0.078	-0.11	0.2	0.34	-0.14	0.41	-0.068	-0.56	-0.14	1	0.33	0.25
sqft_living15	-0.0029	0.59	0.39	0.57	0.76	0.14	0.28	0.086	0.28	-0.093	0.71	0.73	0.2	0.33	-0.0027	-0.28	0.049	0.33	1	0.18
sqft_lot15	-0.14	0.082	0.029	0.087	0.18	0.72	-0.011	0.031	0.073	-0.0034	0.12	0.19	0.017	0.071	0.0079	-0.15	-0.086	0.25	0.18	1

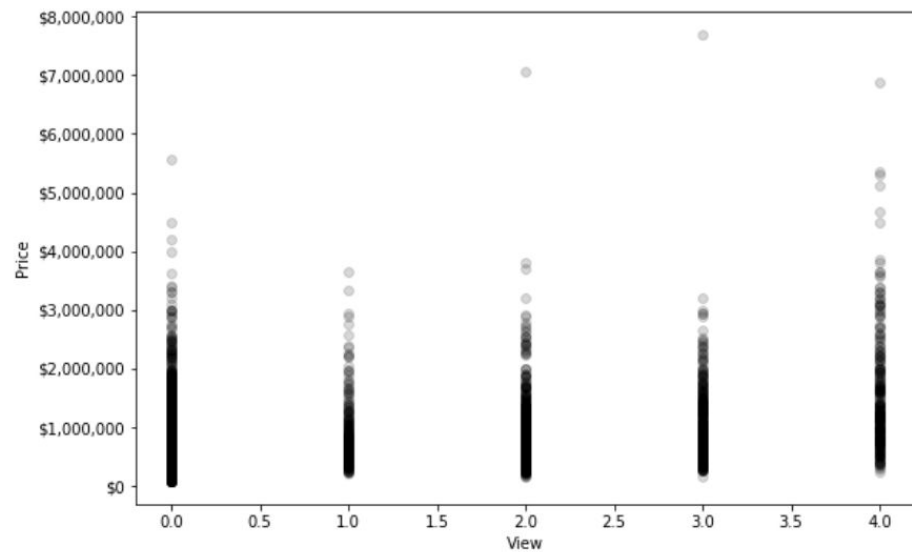
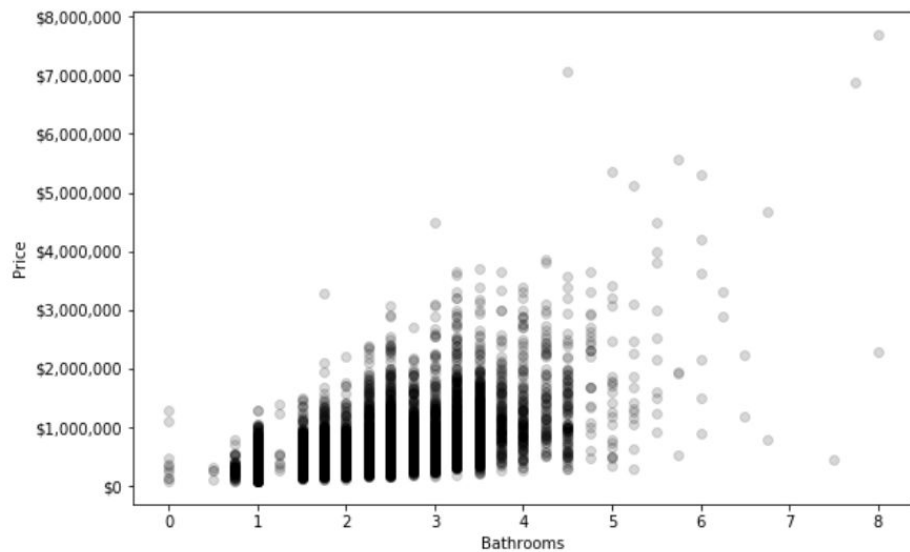


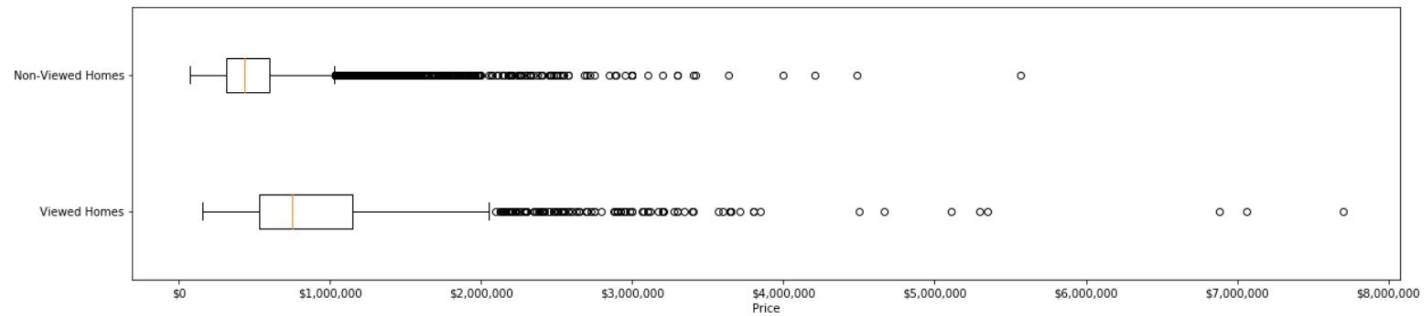
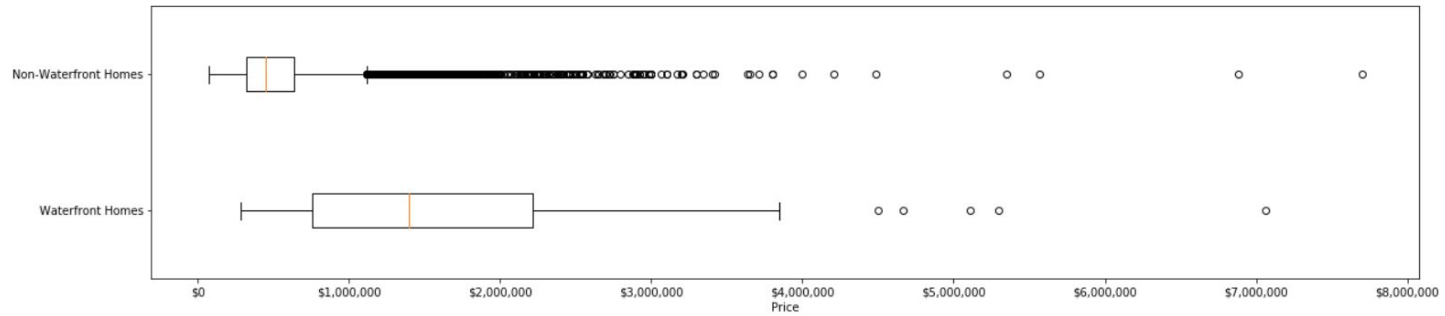
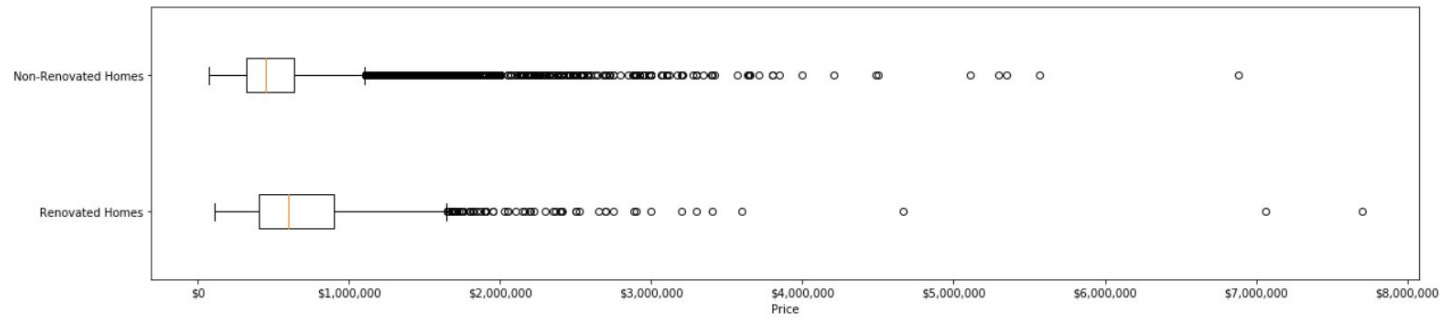
Correlates to price

sqft_living	0.702035
grade	0.667434
sqft_above	0.605567
sqft_living15	0.585379
bathrooms	0.525138
view	0.397293
sqft_basement	0.323816
bedrooms	0.308350
lat	0.307003
waterfront	0.266369





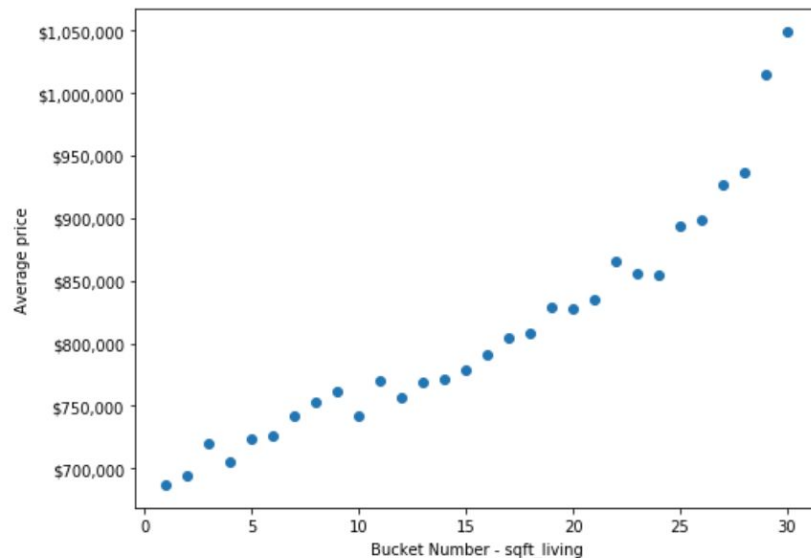




Feature Encoding

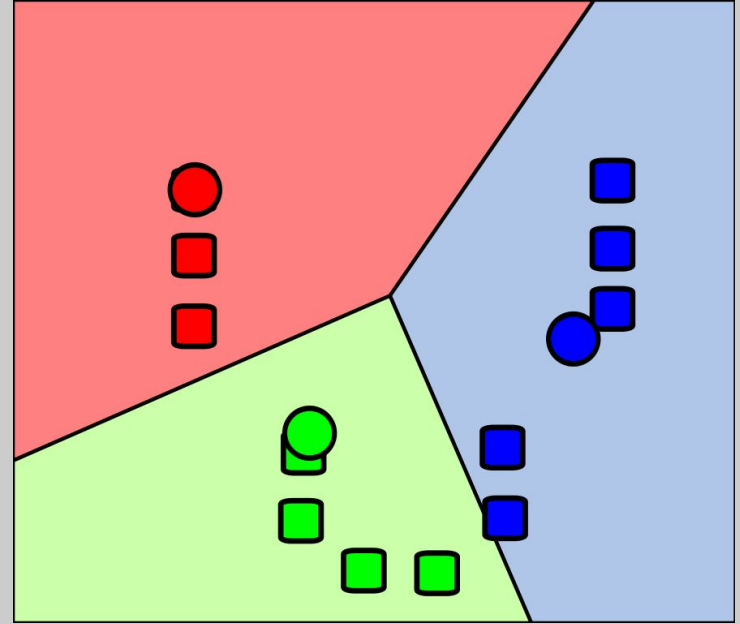
To eliminate some of the variability of a feature, we can group each observation into intervals of similarly priced homes. After each observation is in its bucket, we calculate the average of each bucket, and then assign that average value to a new column in the dataframe.

For example, when I encode the feature 'sqft_living', each home will be grouped into 30 equally populated buckets corresponding to their square footage. I'll calculate the average of each bucket, and it will be assigned to a new feature, which can be used to predict the price.

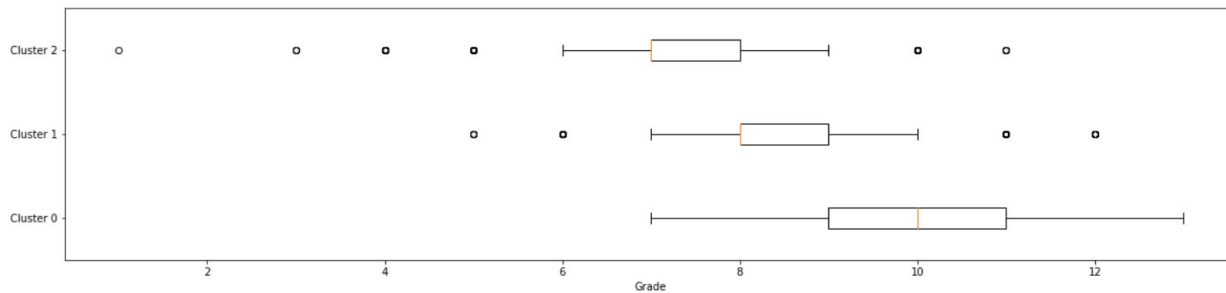
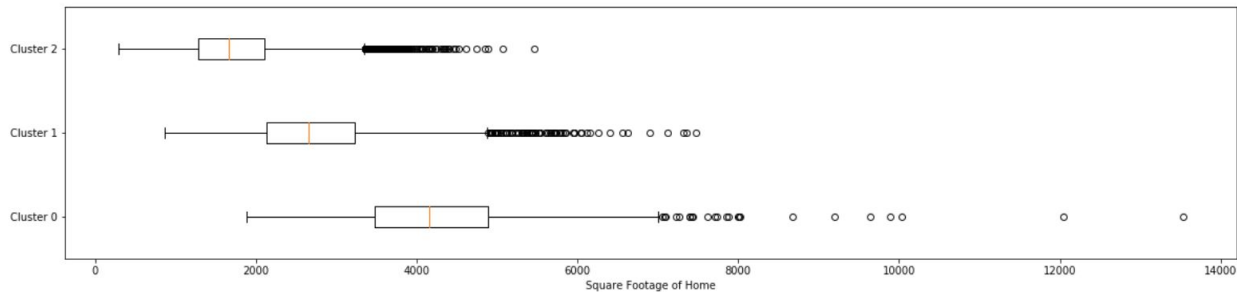
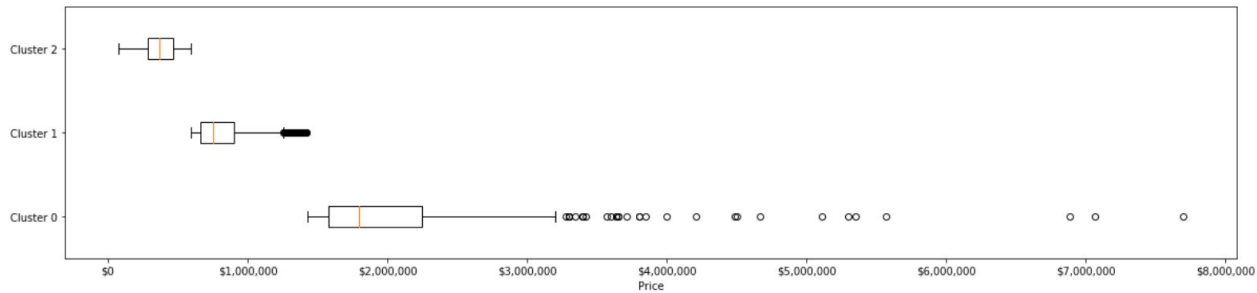


Clustering

Kmeans clustering is an unsupervised learning technique used to identify similar groupings in a dataset. We can use clustering to find groups of homes that have similar characteristics, and create models of these smaller subgroups to obtain a more accurate overall model.



Results of Clustering



Model using all features

	Features	Estimated Coefficients
0	bedrooms	-40546.613566
1	bathrooms	45969.896289
2	sqft_living	117.451295
3	sqft_lot	0.118722
4	floors	5087.980474
5	waterfront	581027.588474
6	view	52480.450568
7	condition	24334.440552
8	grade	93012.019086
9	sqft_above	72.997345
10	sqft_basement	44.453950
11	yr_built	-2702.168375
12	yr_renovated	20.869308
13	zipcode	-586.456083
14	lat	603594.802672
15	long	-209172.985549
16	sqft_living15	19.398367
17	sqft_lot15	-0.390268

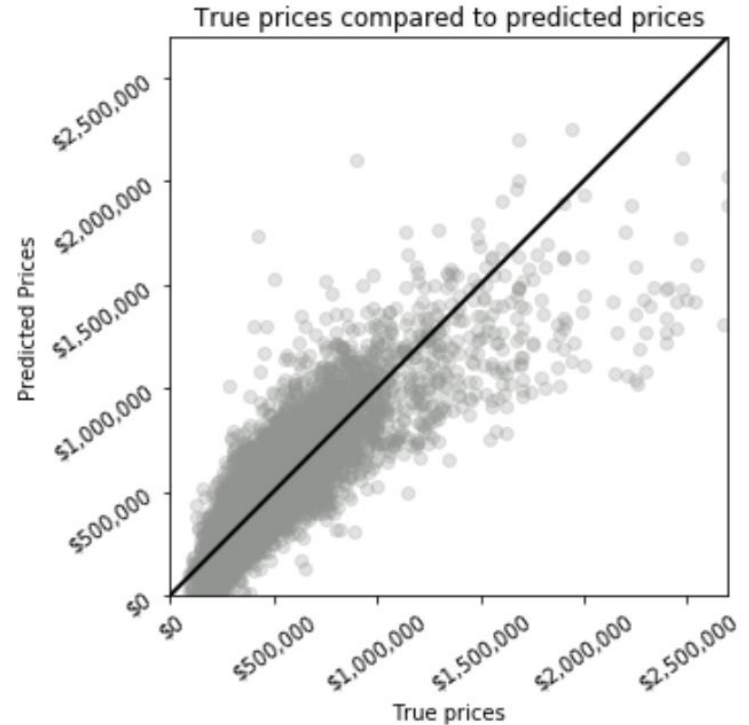
Estimated intercept coefficient: 7909720.118896149

Summary Statistics

R-squared value: 0.6962965895543749

Root Mean Squared Error: 193088.65760593285

Mean Absolute Percentage Error (MAPE): 25.838850273124415



Cluster 0 Model

	Features	Estimated Coefficients
0	sqft_living_bucket_price_squared	1.298238e-07
1	waterfront	5.954120e+05
2	condition	6.721346e+04
3	grade	6.211902e+04
4	sqft_living	9.376187e+01

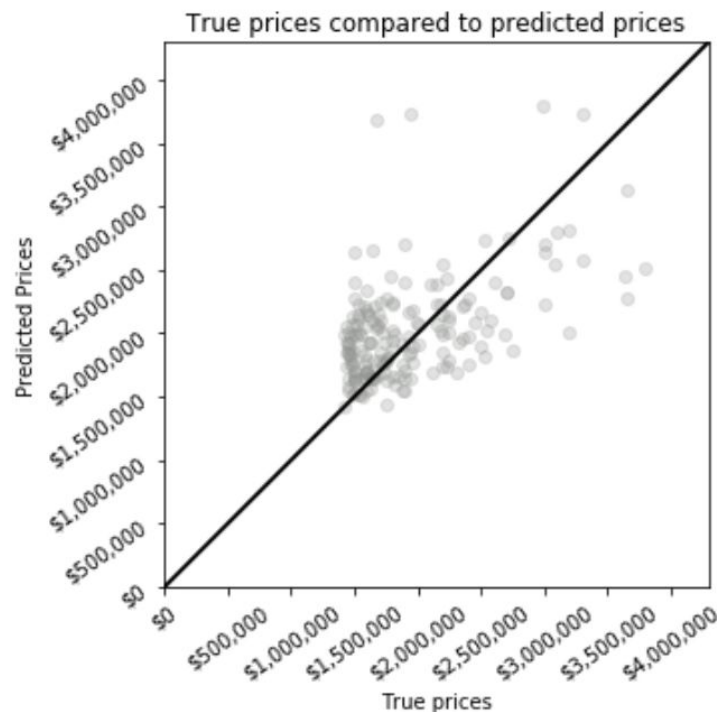
Estimated intercept coefficient: 135601.22614349728

Summary Statistics

R-squared value: 0.3651722230592157

Root Mean Squared Error: 534305.9167235789

Mean Absolute Percentage Error (MAPE): 18.801434341787772



Cluster 1 Model

	Features	Estimated Coefficients
0	bathrooms_bucket_price_squared	4.058957e-07
1	grade_bucket_price	2.143519e-01
2	bathrooms	-4.196916e+04
3	waterfront	2.167078e+05
4	condition	1.528197e+04
5	grade	4.867775e+04
6	yr_built	-1.160438e+03
7	zipcode	-6.985400e+02
8	lat	2.144938e+05
9	long	-5.242644e+05
10	sqft_living	6.546359e+01
11	sqft_living15	4.725114e+01

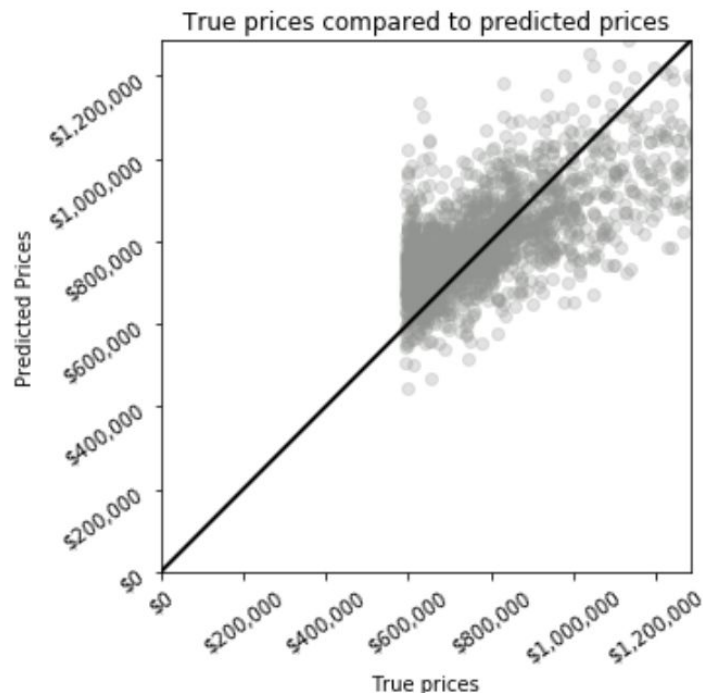
Estimated intercept coefficient: -3770611.0657177963

Summary Statistics

R-squared value: 0.44342874856398595

Root Mean Squared Error: 139180.63909621062

Mean Absolute Percentage Error (MAPE): 12.935603453041134



Cluster 2 Model

	Features	Estimated Coefficients
0	bedrooms	-1.229909e+04
1	grade_bucket_price_sqrt	3.231988e+02
2	sqft_living_bucket_price_squared	7.410774e-07
3	bathrooms	2.437804e+04
4	waterfront	1.295705e+05
5	condition	1.240750e+04
6	grade	3.218932e+04
7	yr_built	-8.241704e+02
8	zipcode	-4.188423e+01
9	lat	4.057994e+05
10	long	3.945118e+04
11	sqft_living	1.685261e+01

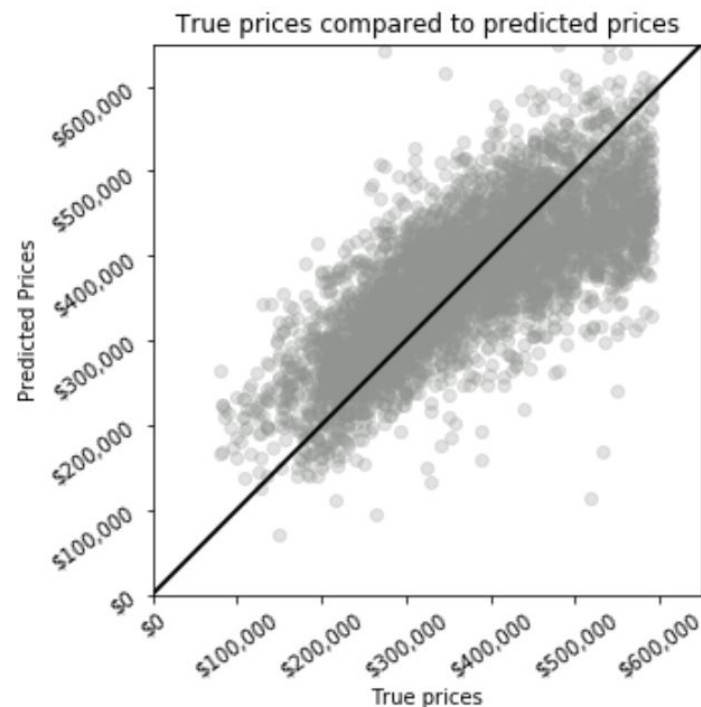
Estimated intercept coefficient: -8974715.60852908

Summary Statistics

R-squared value: 0.5664798337761238

Root Mean Squared Error: 75550.38559401885

Mean Absolute Percentage Error (MAPE): 17.925521103105808



Clustering and Feature Encoding

9.28% improvement

	MAPE
All Features - no clustering	25.84%
Cluster 0	18.80%
Cluster 1	12.94%
Cluster 2	17.93%
Average of clusters	16.56%

Conclusion

It should be noted that this model is only trained on home sales in King County, Washington between 2014 and 2015. It's use is limited to the same geographical area and time bounds, and should not be used to make predictions of home prices far into the future.