



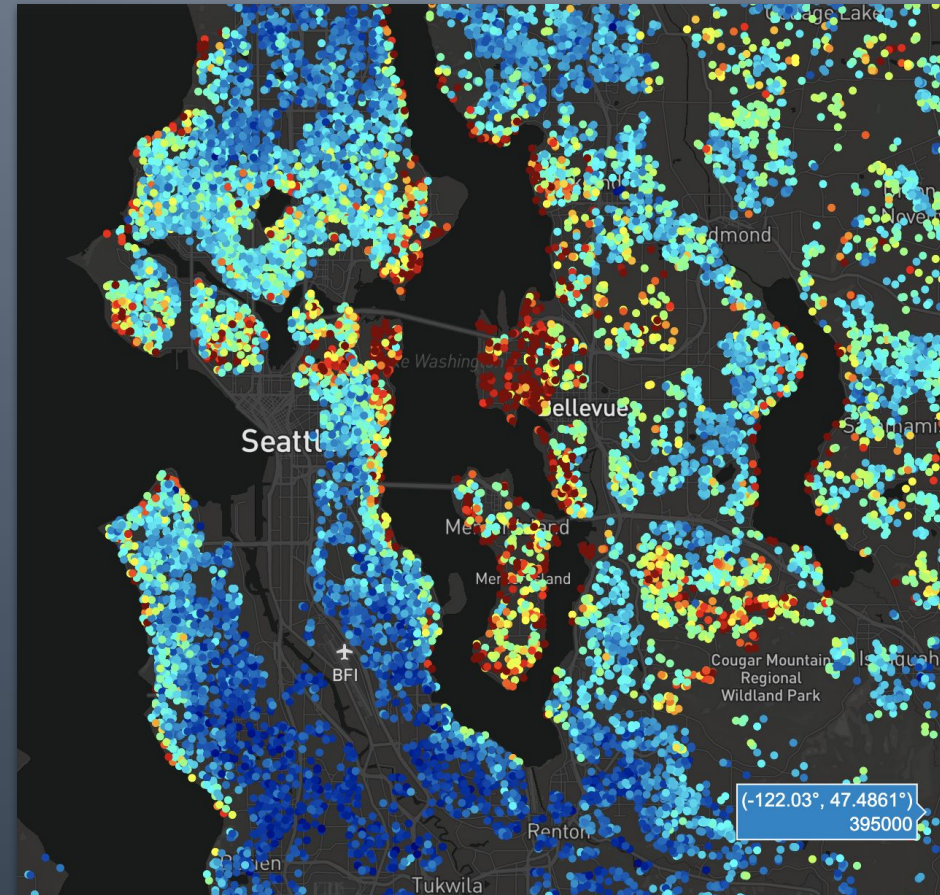
The Data Science of Selling your Home

Data-proven variables for maximizing profits



Background

- Find variables contributing to a higher selling profit.
- Data from houses sold between 2014-2015 in King County, Washington State, USA.
- Based on 21,596 data points with 15 variables each.

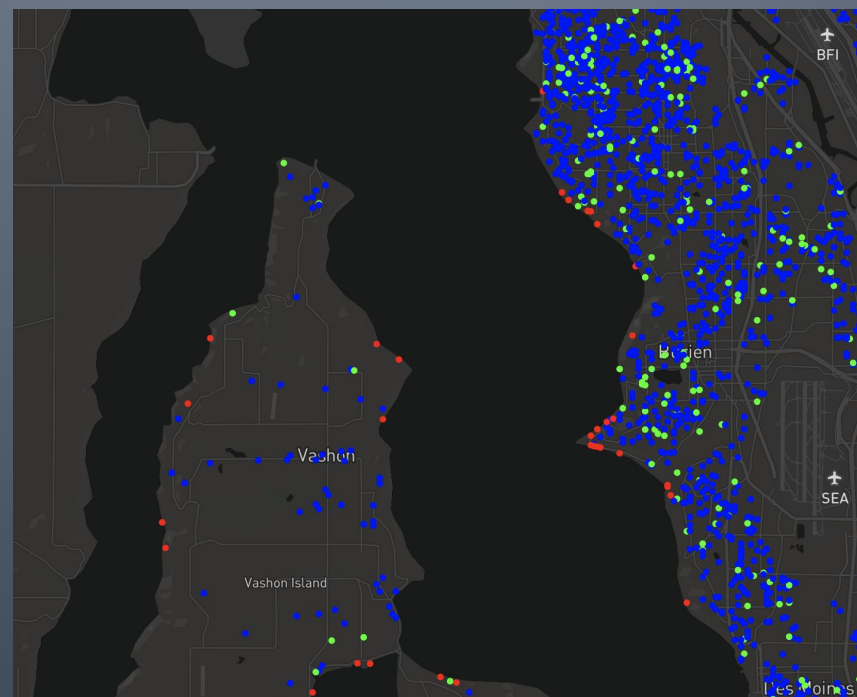


Map made with Mapbox and Plotly. Showing price distribution in King County, Washington



Data Cleaning

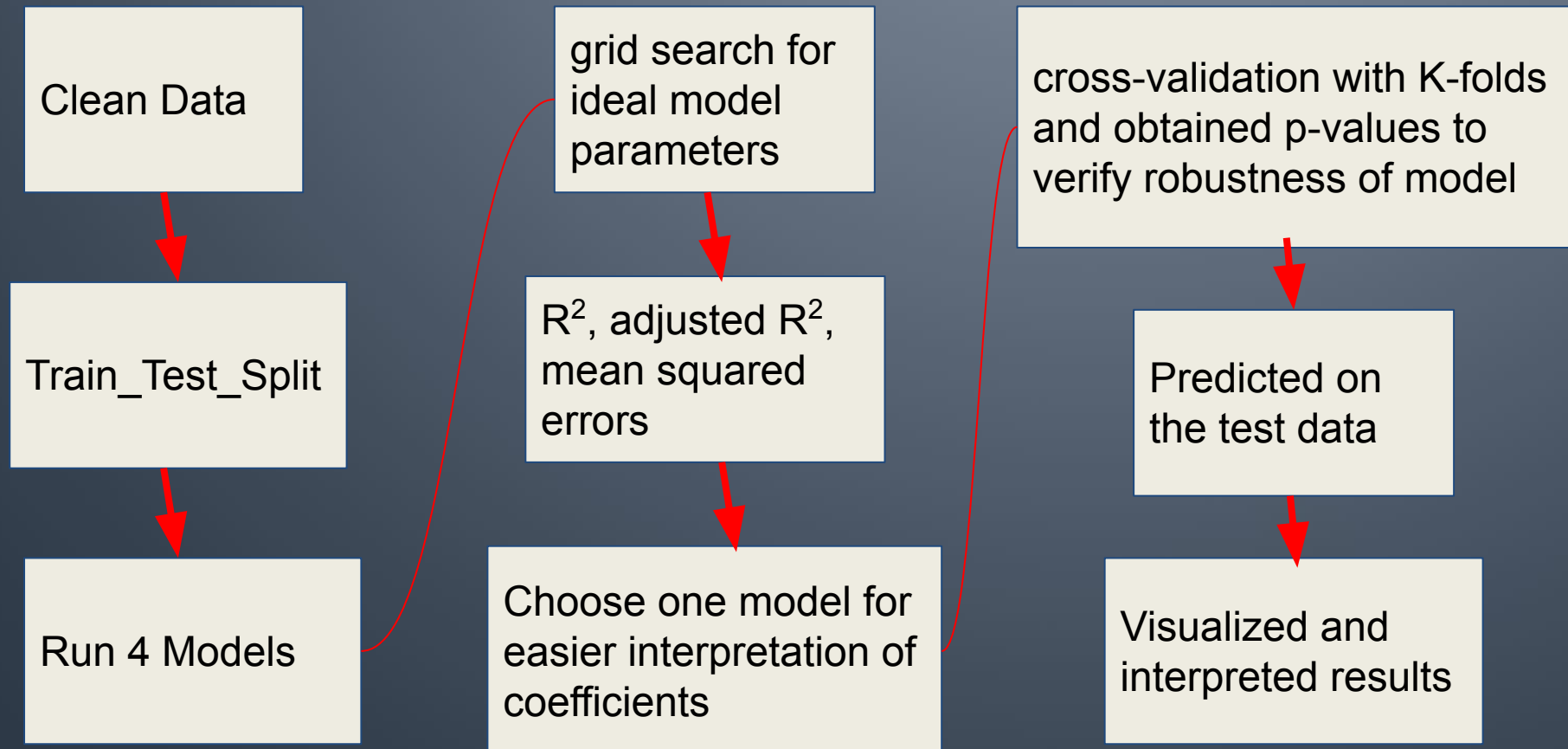
- Kept all variables, only dropped ID.
- Converted date to month, day, and year columns.
- Filled NaN's with the mode of column.
- Dropped incorrect values (i.e. 33 bedroom house with 1 bathroom)
- **Categorized**
 - waterfront, view, grade, condition, zip
- **Scaled**
 - sqft_living, sqft_above, sqft_lot, sqft_living15, sqft_lot15, sqft_basement (boolean transform)
 - yr_built
 - yr_renovated (boolean transform)
 - price (log scaled)



Map made with Mapbox and Plotly. Showing incorrectly labeled waterfront properties in green as NaN's.



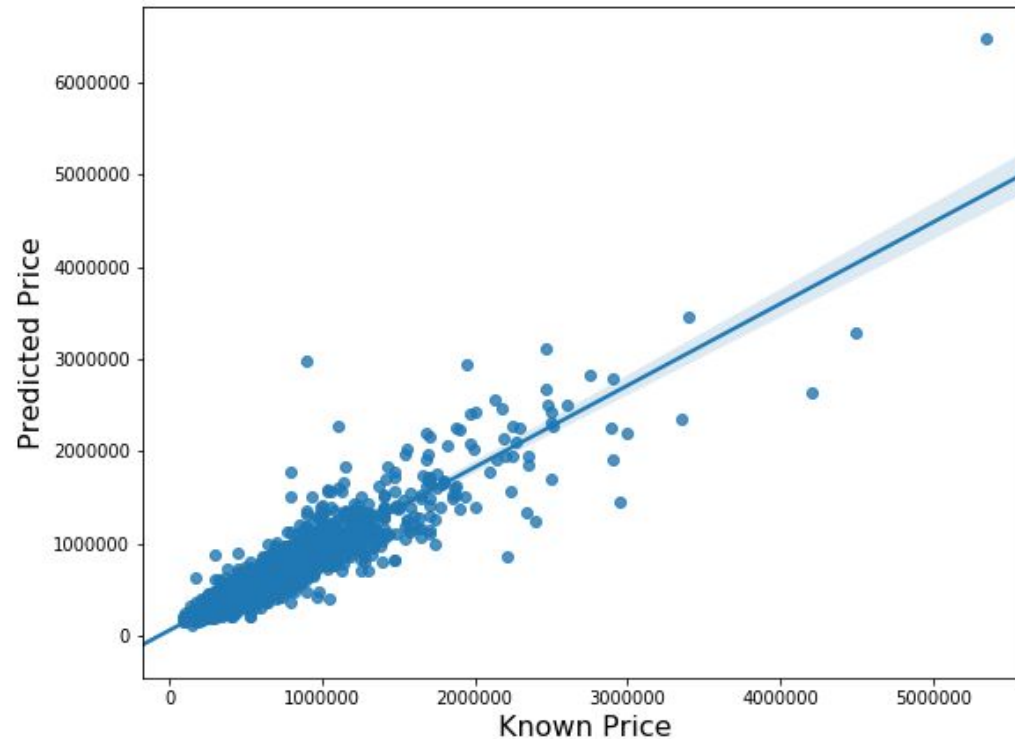
Methodology



Linear Regression
Ridge Regression
Lasso Regression
Elastic Regression

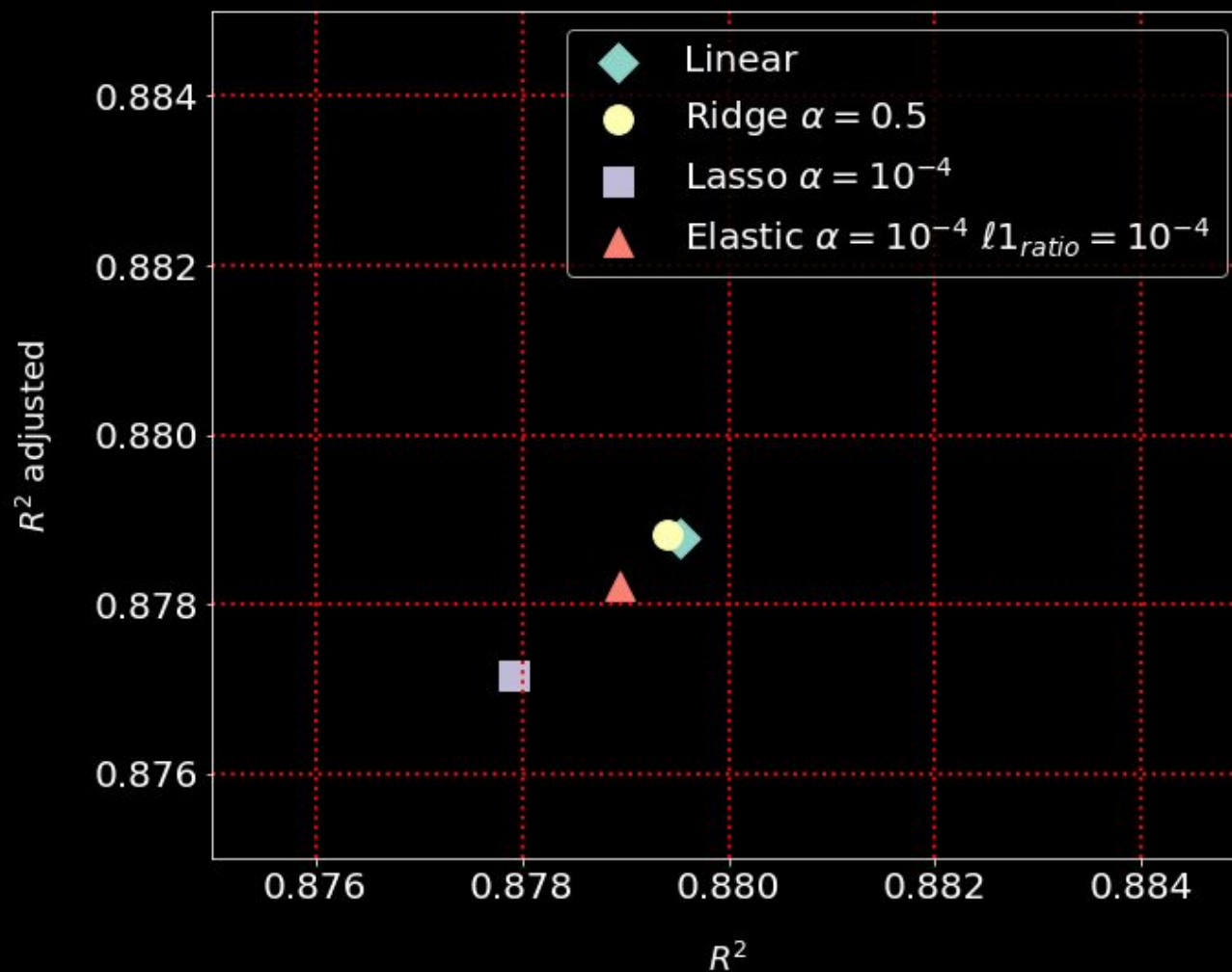
Models Used

Lasso Fit: Predicted vs. Known Prices





Model Comparison





Model predictions for higher price

- Higher latitude
- Living by the water
- Increasing square footage
- Having a bigger yard
- Living in the 98039 zip code (Medina, Washington)
- Having rich neighbors like Bill Gates

That's
MY zip!





Increasing Grade

Grade Scale is from 1-13 and can be found [here](#).

Initial Grade	Final Grade	% Increase in price
4	7	16%
4	10	51%
10	11	1.5%





Adding View

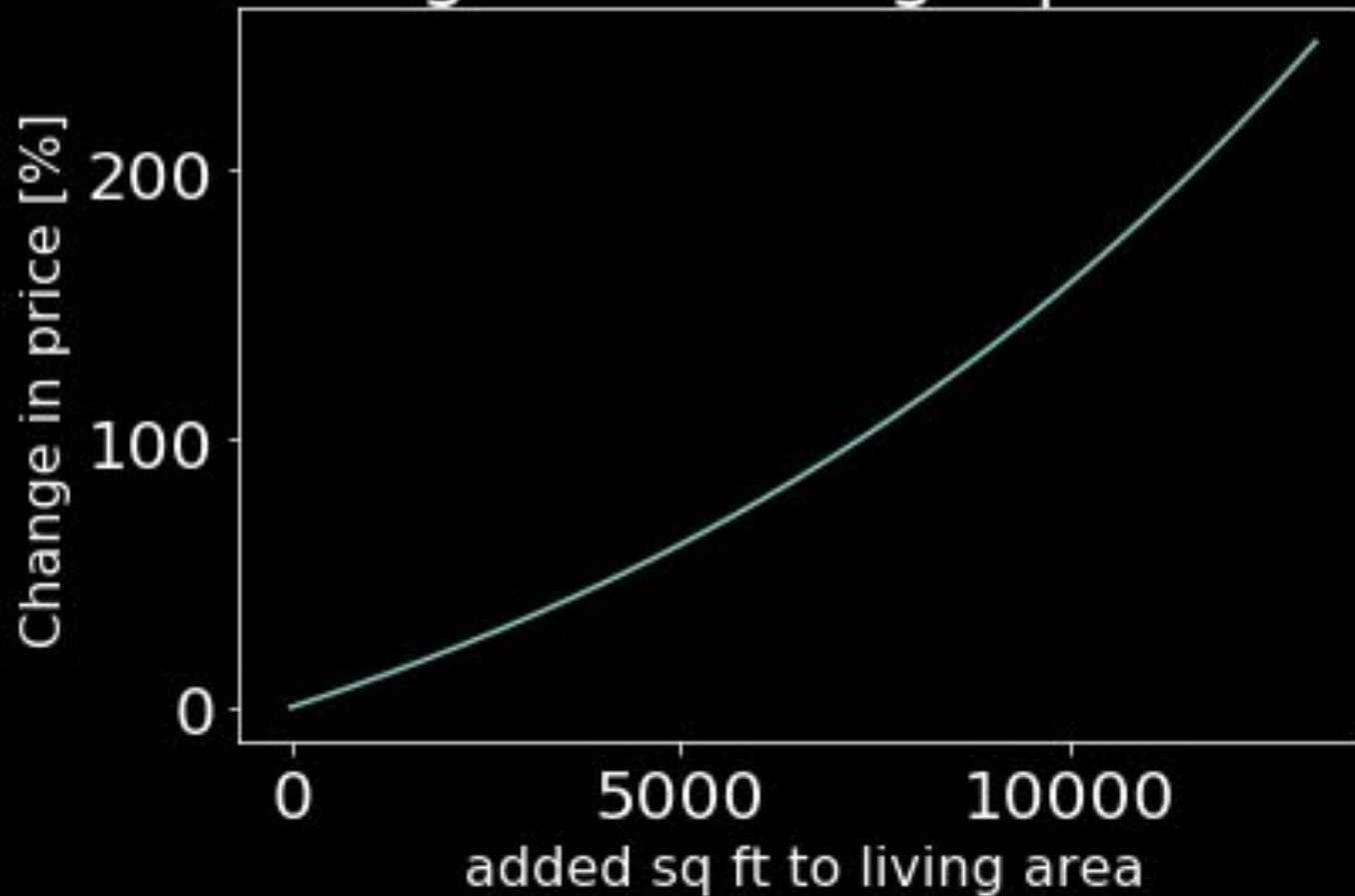
Initial View	Final View	% increase in price
0	1	13%
0	4	32%
1	3	8%
1	4	18%





Adding Square Footage

Price Change vs adding sq.ft. living area





Future Work

- Remove expensive houses from data (outliers).
- Try binning some of the categorical data.
- Separate regression for different subsets.

Questions?