

# Module 2 Project:

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# Modeling Housing Prices Around Seattle

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- Why?



# Approach

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- Create a model using the King County Housing dataset (Seattle & Seattle metro area)
- Answer these questions:
  - Which housing features drive housing prices?
  - How can I get the best and worst deals?



# The Model: Features

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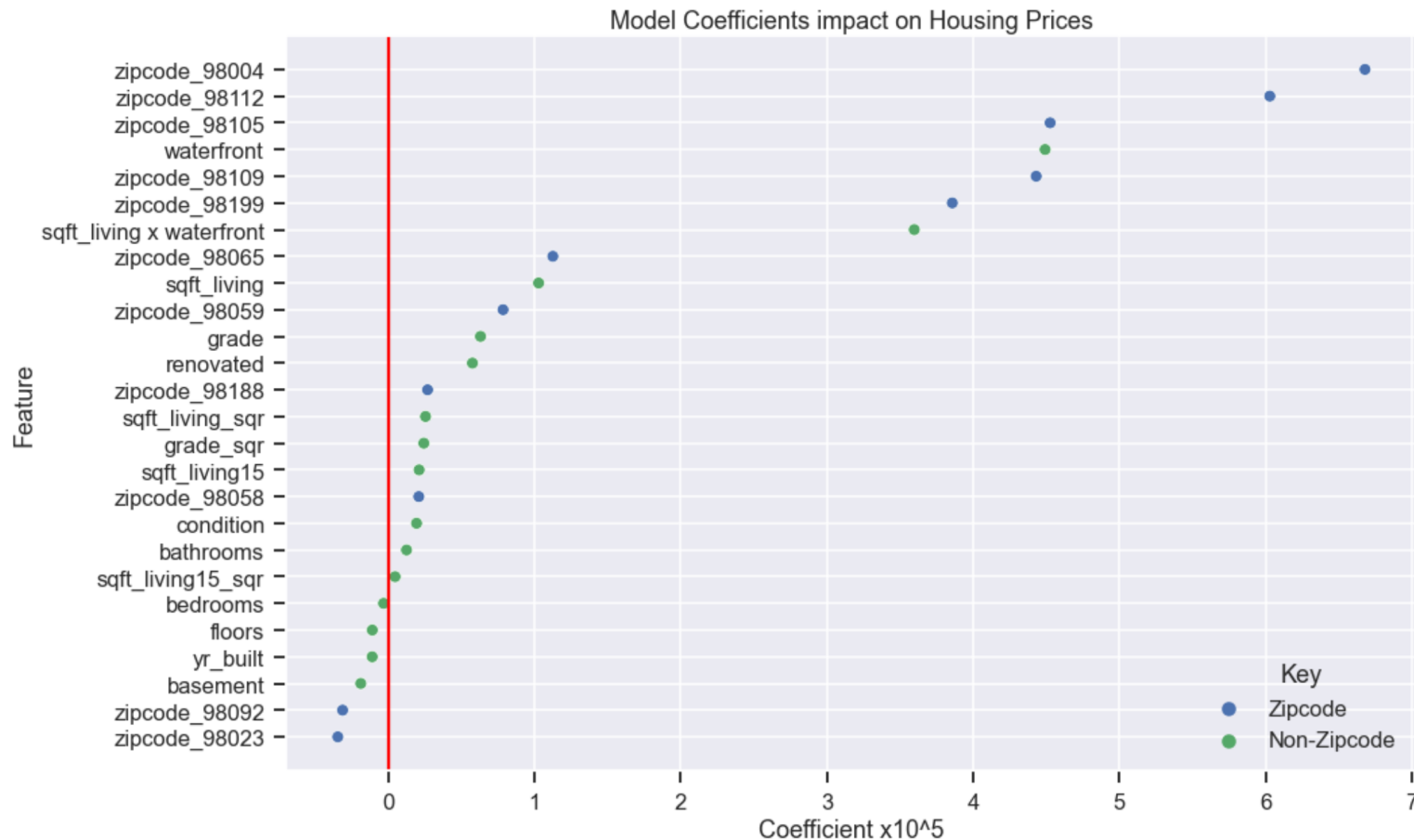
- Predictors:

- Bedrooms
- Bathrooms,
- Square feet
- Floors
- Waterfront
- Condition
- Grade
- Year built
- Square feet of nearest 15 neighbors
- Renovated
- Basement
- Zipcode

- Outcome Variable: Price



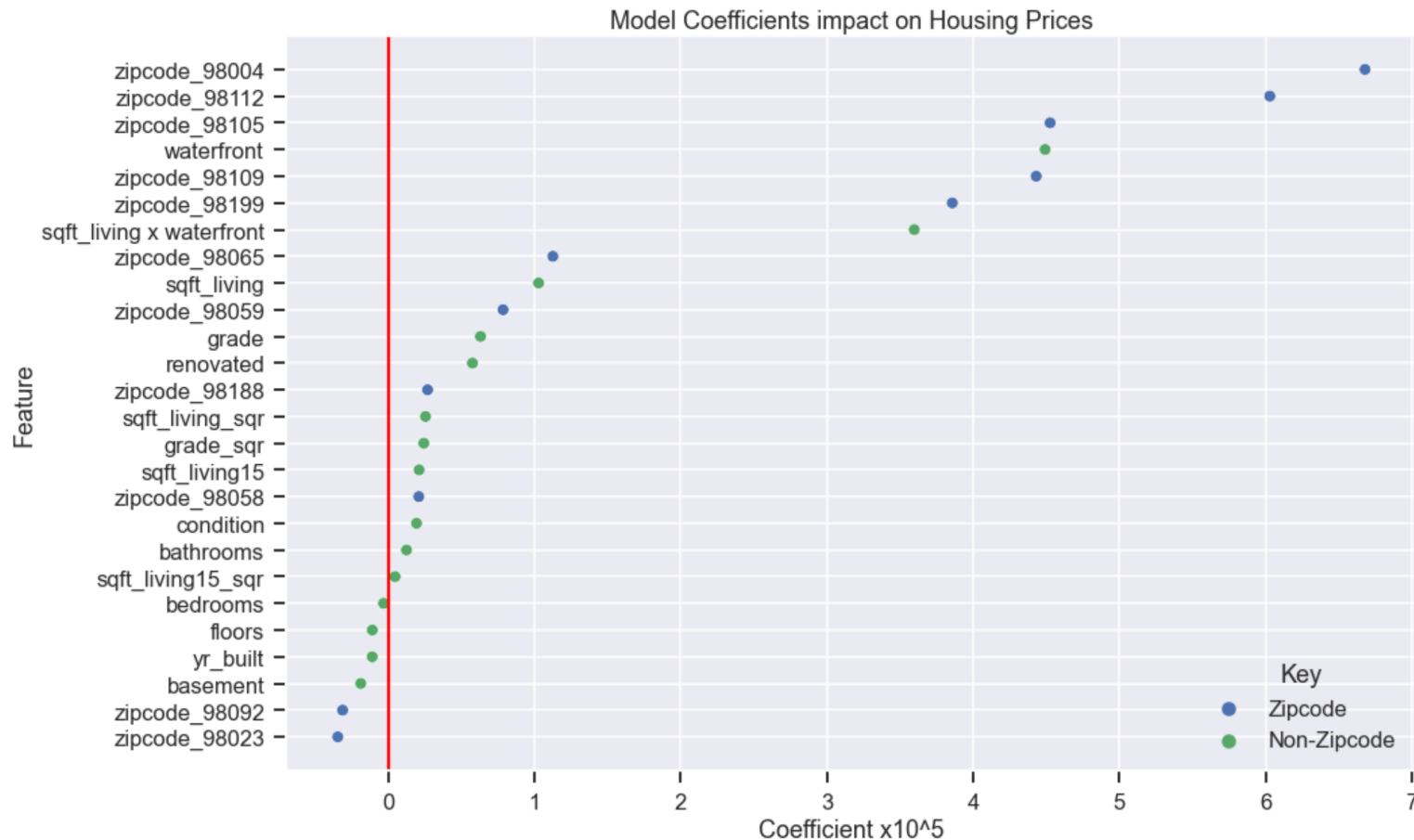
# The Model: Driving Features



Rank	Feature	Coefficient
1	zipcode_98004	6.680633
2	waterfront	4.493111
3	sqft_living x waterfront	3.598411
4	sqft_living	1.029954
5	grade	0.632588
6	grade_sqr	0.244879
7	sqft_living15	0.212852
8	condition	0.195142
9	basement	-0.186253
10	bathrooms	0.126754
11	yr_built	-0.108136
12	sqft_living15_sqr	0.048538
13	bedrooms	-0.030189



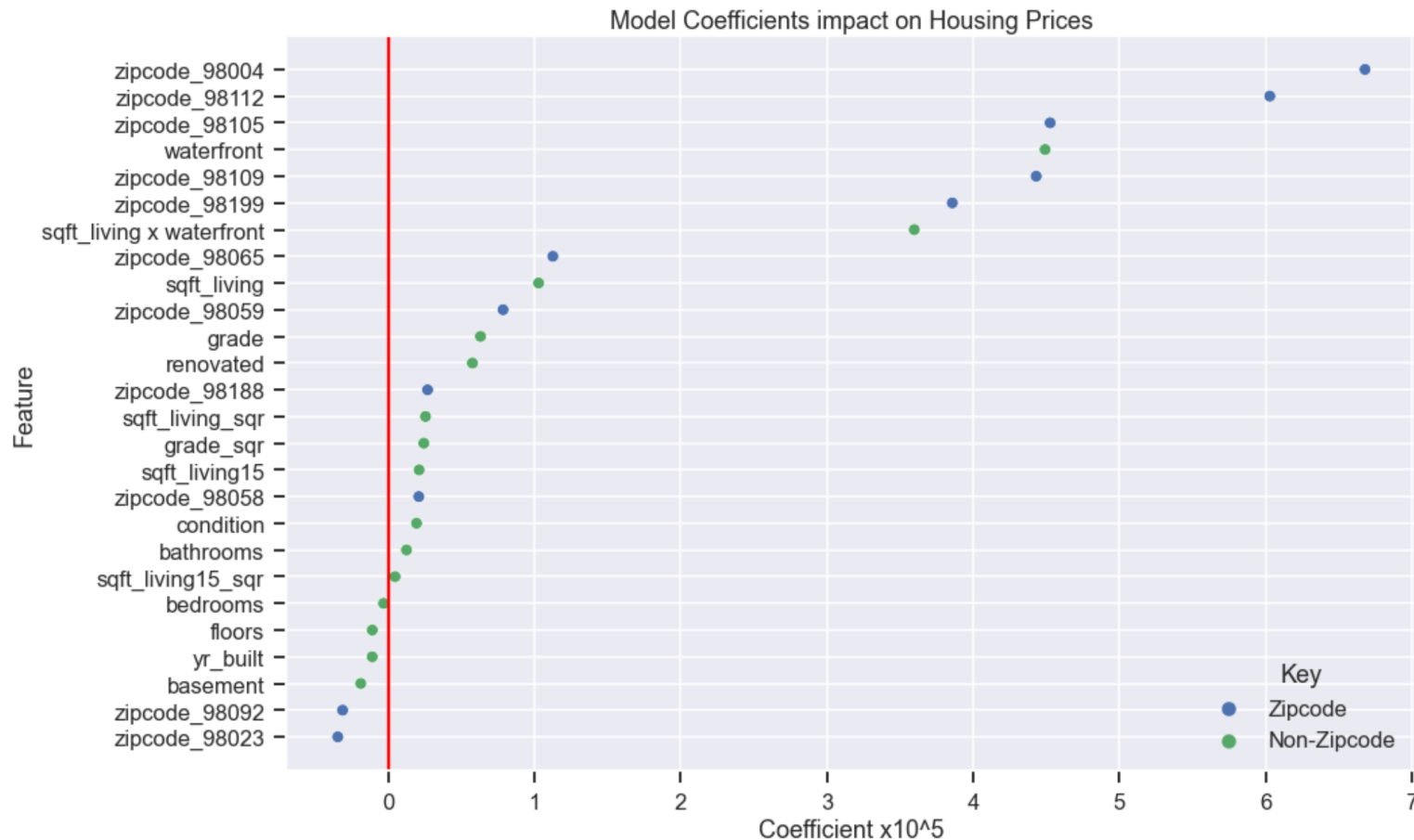
# The Model: Bedrooms and Sqft



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- 81%
- Model does not perform well where prices are very high.



# The Model: Best/Worst Deals

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- Top 10 “best deals” (data points with the most negative errors) v. dataset averages
  - Square Feet
  - Bathrooms
  - Zipcodes: 98006, 98107, 98199
- Top 100 “worst deals” (data points with the most positive errors) v. dataset averages
  - Waterfront
  - Renovated houses
  - Zipcodes: 98112, 98004, 98070



# Summary / Recommendations

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- The most important indicator of housing prices is location
- If you are looking for a good price on a house, look for something with a lot of square feet and but yet fewer bedrooms, avoid waterfront properties or any properties that have been newly renovated

## Future Work

- Expand the model to include data from larger geographical regions.



# The End

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- Any questions?