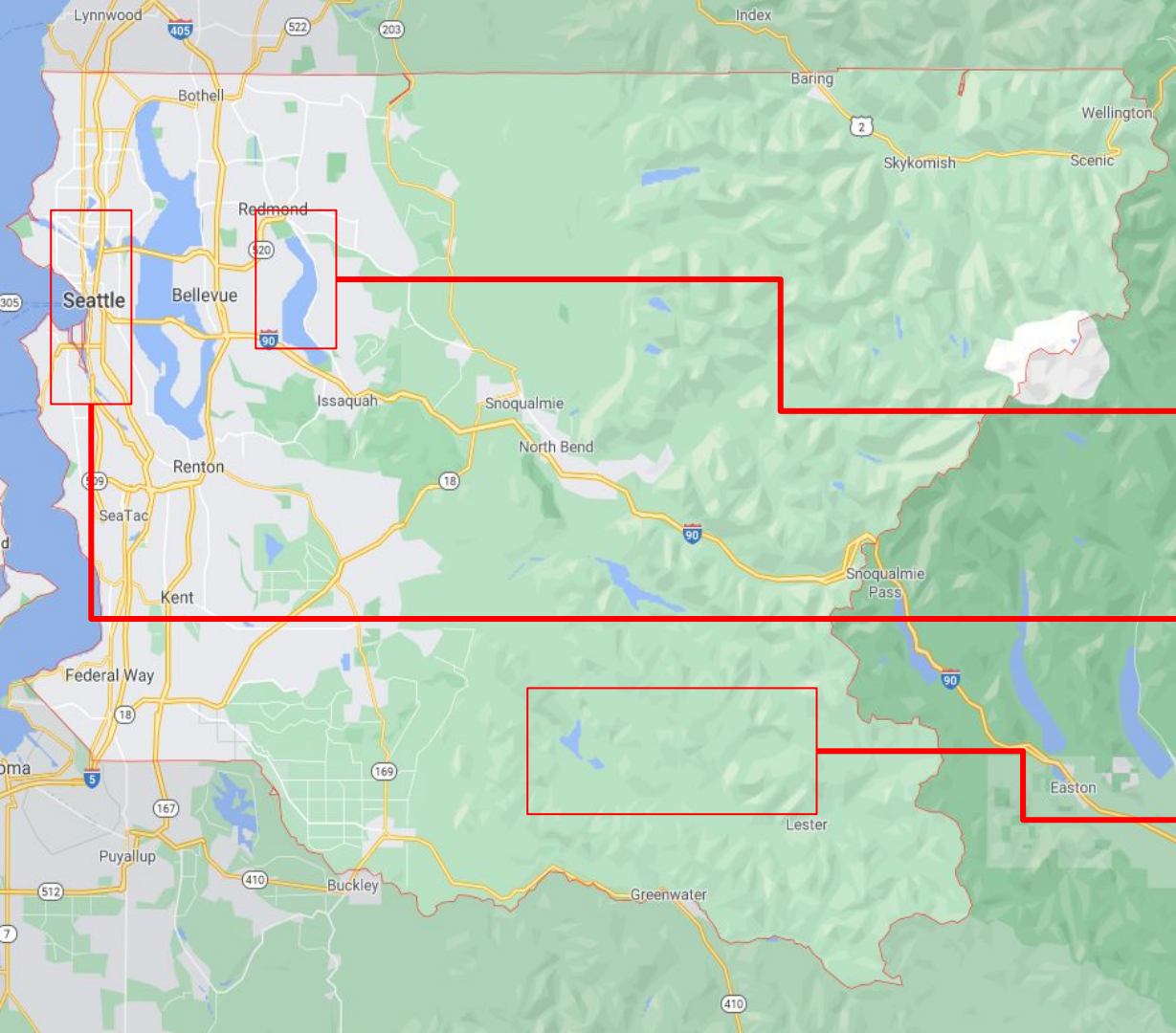


# Predicting House Prices in King County, Washington

Leonard Boes

# First look on map

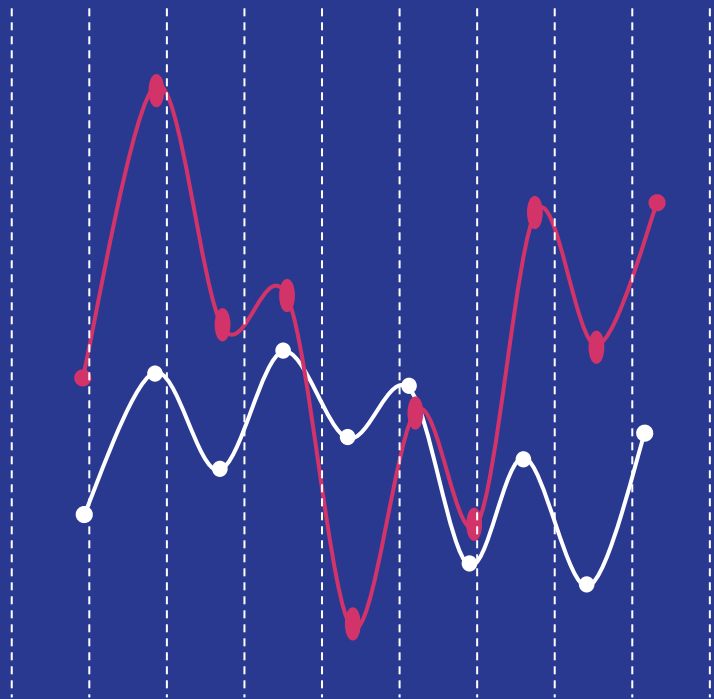


Unique areas like waterview

Dense Seattle Metropolis

Rural areas far away from city  
and interstate

# First look on the data



- Price and date of individual sales
- Number of bedrooms, bathrooms and floors
- Rank of condition and information about waterview
- Geographical location and zipcode
- Information about size of lot and living area
- Information about size of lot and living area of 15 neighboring houses
- Year of construction and renovation
- Roughly 21,600 observations

# Condition of the data

Data columns (total 21 columns):

id	21597	non-null	int64
date	21597	non-null	object
price	21597	non-null	float64
bedrooms	21597	non-null	int64
bathrooms	21597	non-null	float64
sqft_living	21597	non-null	int64
sqft_lot	21597	non-null	int64
floors	21597	non-null	float64
waterfront	19221	non-null	float64
view	21534	non-null	float64
condition	21597	non-null	int64
grade	21597	non-null	int64
sqft_above	21597	non-null	int64
sqft_basement	21597	non-null	object
yr_built	21597	non-null	int64
yr_renovated	17755	non-null	float64
zipcode	21597	non-null	int64
lat	21597	non-null	float64
long	21597	non-null	float64
sqft_living15	21597	non-null	int64
sqft_lot15	21597	non-null	int64

Date needed to be converted

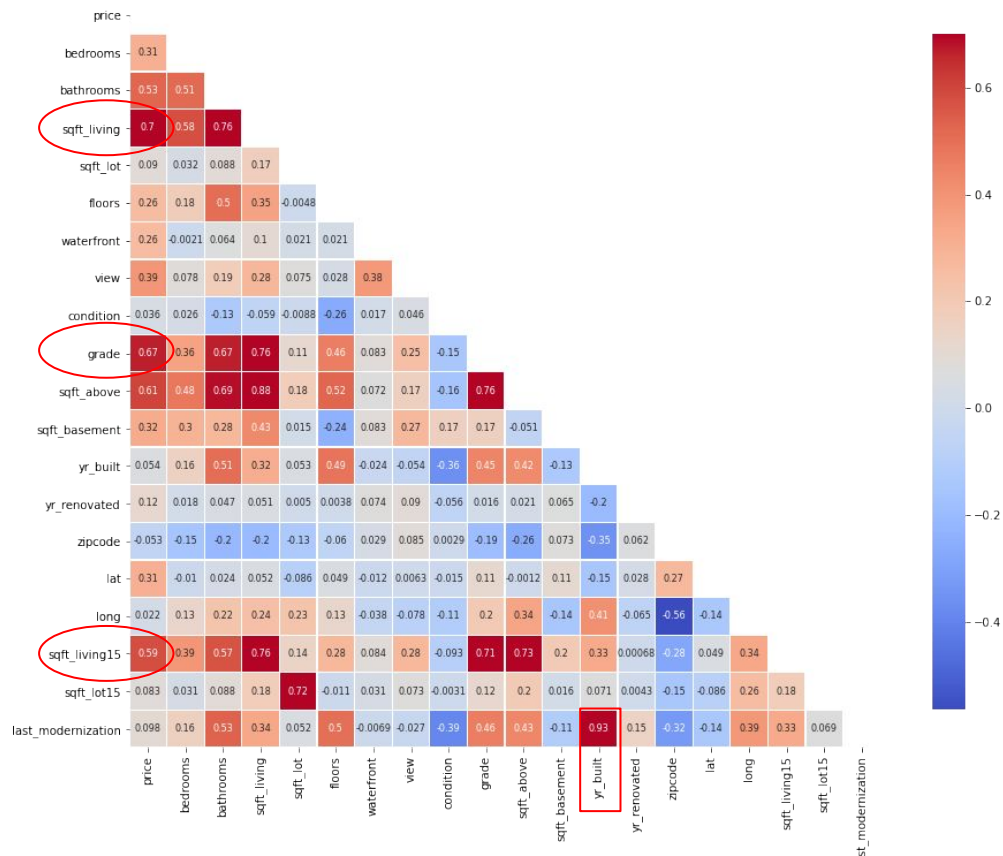
Missing values needed to be accounted for

Something is wrong with the basement measures

Fortunately, only few values were missing and could logically be replaced with zeros

# Exploring the data

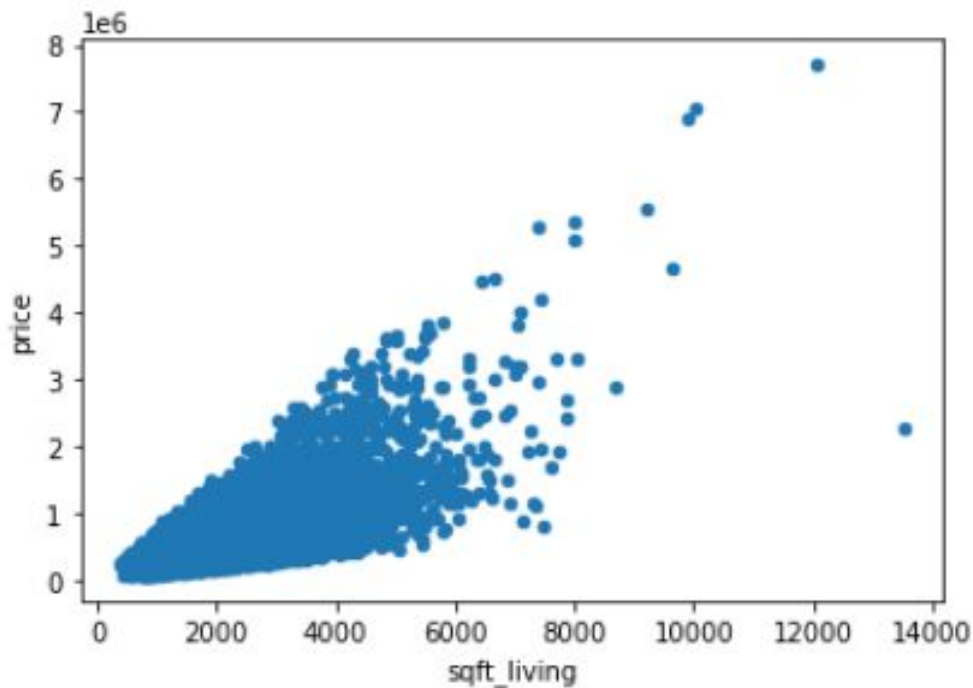
# Exploring the data



Highest correlation with price can be observed at:

Size of living area,  
grade  
& neighbors size of living area

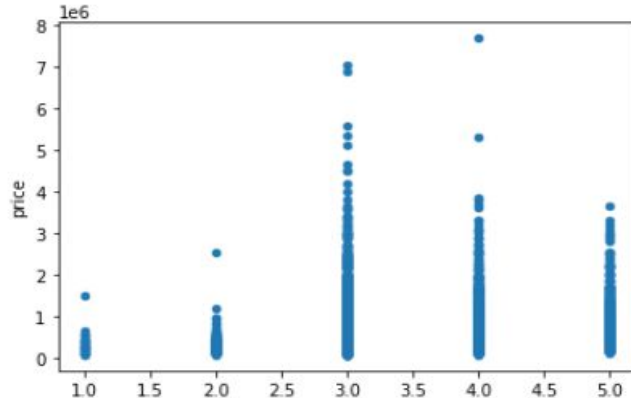
# Exploring the data



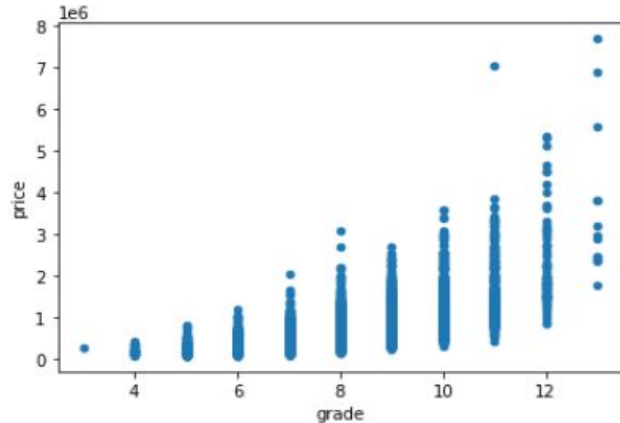
Highest correlation with price can be observed at:

Size of living area,  
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& neighbors size of living area

# Exploring the data



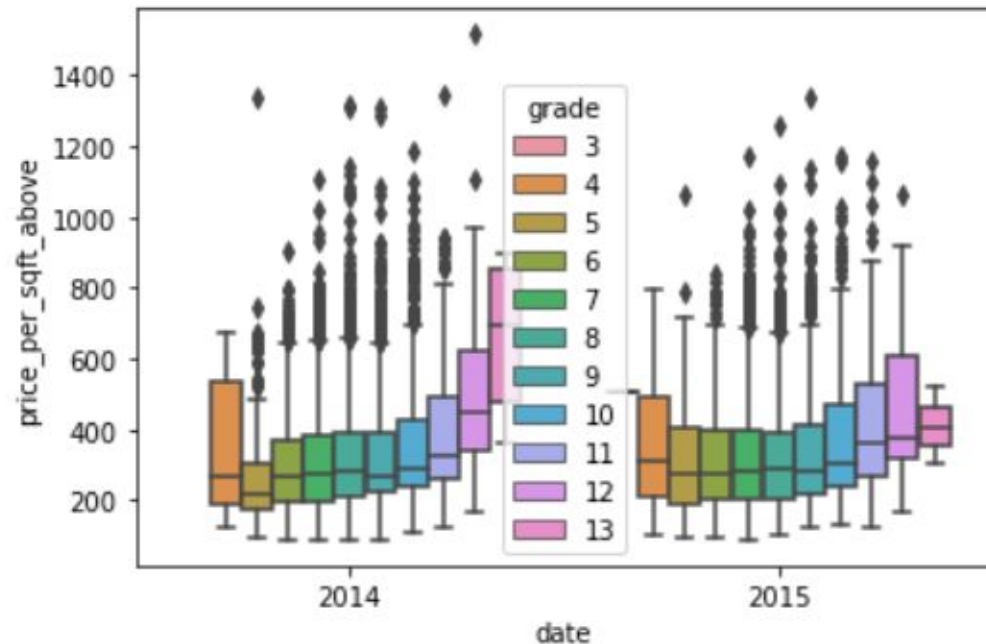
No linear relationship between price and condition rating. Condition just needs to be  $> 2$



Positive correlation trend between grade and price

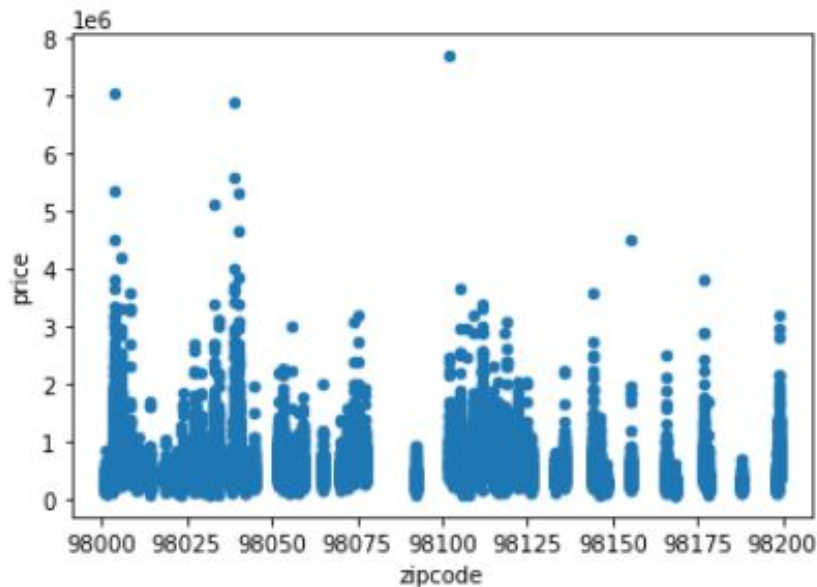


# Exploring the data



If you want to buy luxury real estate, you might want to put a special emphasis on market timing

# Exploring the data

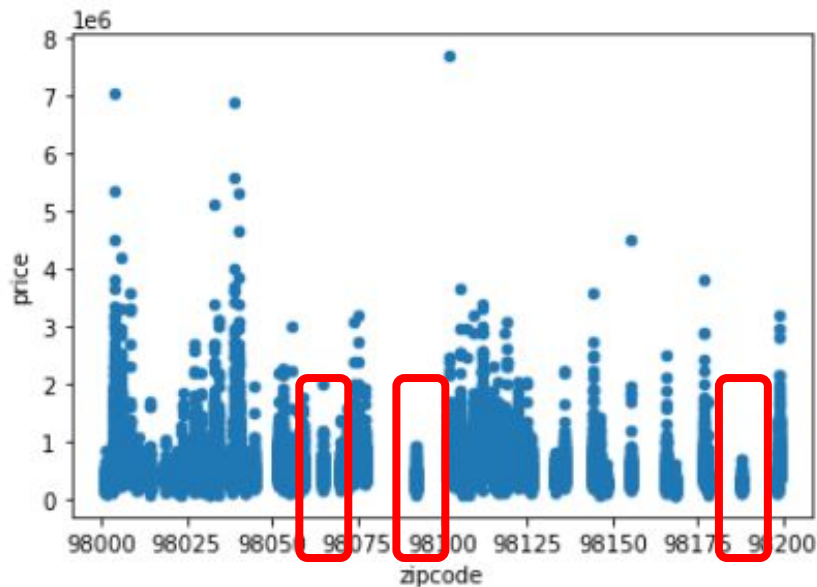


Some zipcode areas seem to have really cheap house prices.

The zip code areas with best price to footage ratio are:

98092, 98002, 98030, 98001, 98023,  
98042, 98003 & 98038,

# Exploring the data



Some zipcode areas seem to have really cheap house prices.

The zip code areas with best price to footage ratio are:

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98042, 98003 & 98038,

# Building new features

# Building new features

- Size of living area seemed to be important
- I wanted to add geographical dimension as well

```
#calculate how many squarefeet the rooms have on average
df['rooms_per_sqft'] = (df.bedrooms + df.bathrooms) / df.s

#calculate how many bedrooms per squarefoot are there
df['bedrooms_per_sqft'] = df.bedrooms / df.sqft_living

#calculate how many bathroomy per squarefoot there are
df['bathrooms_per_sqft'] = df.bathrooms /df.sqft_living

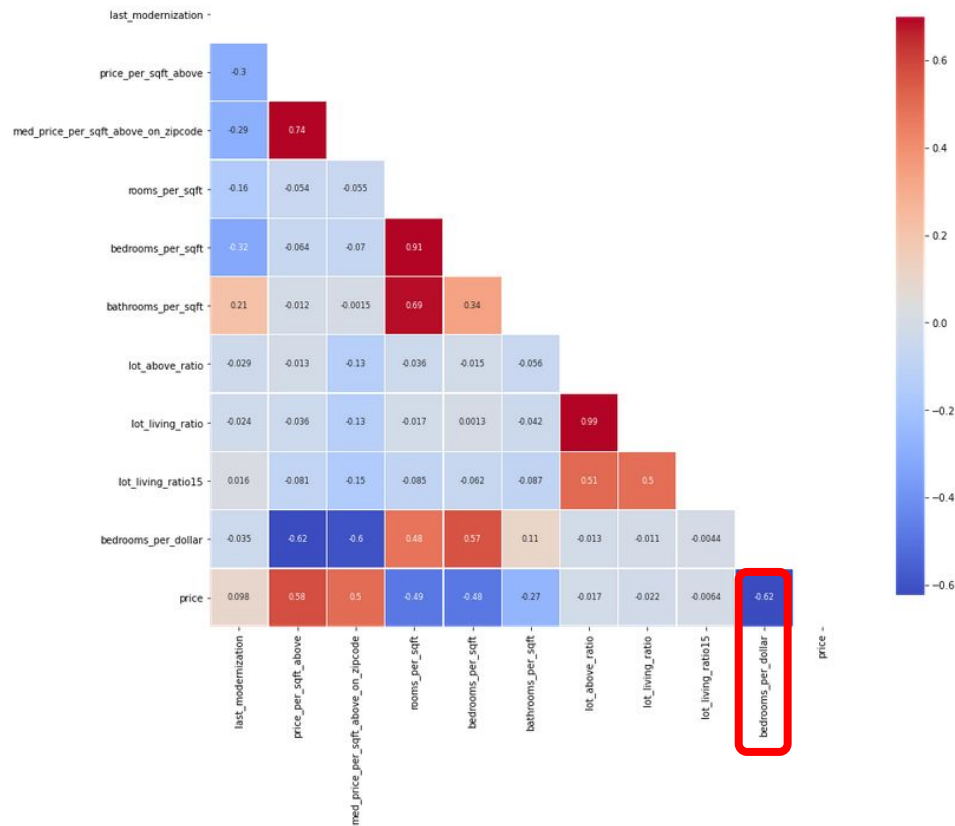
#create lot to above ratio for individual home
df['lot_above_ratio'] = df.sqft_lot / df.sqft_above

#create lot to living ratio for individual home
df['lot_living_ratio'] = df.sqft_lot / df.sqft_living

#create lot to living ratio for nearest 15 neighbors
df['lot_living_ratio15'] = df.sqft_lot15 / df.sqft_living15

#calculate how many rooms you get for a dollar
df['bedrooms_per_dollar'] = df.bedrooms / df.price
```

# Exploring the data



The more bedrooms you get for a dollar, the cheaper the house price.

Sellers should keep the number of rooms normal and not try to squeeze more rooms in a property that neccessary

# Building a predictive model

# Looking at different model qualities

## Best Single Predictor

**Square footage of living area**

RSquared: 0.469

RMSE: \$ 249,948

Price = -52,451 + 285 x area

## Best Feature Combination

**Square footage of living area,  
grade & year of construction**

RSquared: 0.598

RMSE: \$ 217,556

Price = 6319009  
+ 182 x area  
+ 144080 x grade  
- 3684 x year of construction



# Looking at different model qualities

## Old Best Feature Combination

**Square footage of living area,  
grade & year of construction**

RSquared: 0.598

RMSE: \$ 217,556

Price = 6319009

+ 182 x area

+ 144080 x grade

- 3684 x year of construction

## New Best Feature Combination

**Square footage of living area,  
median price per sqft in zip-area,  
waterfront**

RSquared: 0.758

RMSE: \$ 168,863

Price = -497856

+ 267 x area

+ 1594 x median price per sqft in zip area

+ 839,473 x waterfront

# Looking at different model qualities

## Old Best Feature Combination

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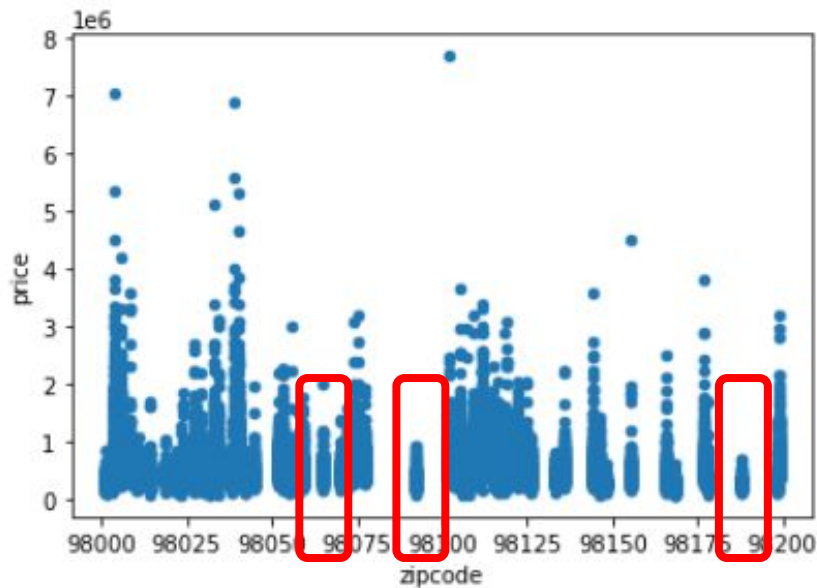
+ 1594 x median price per sqft in zip area

+ 839,473 x waterfront

Not the margin  
of error a house  
buyer wants to  
see

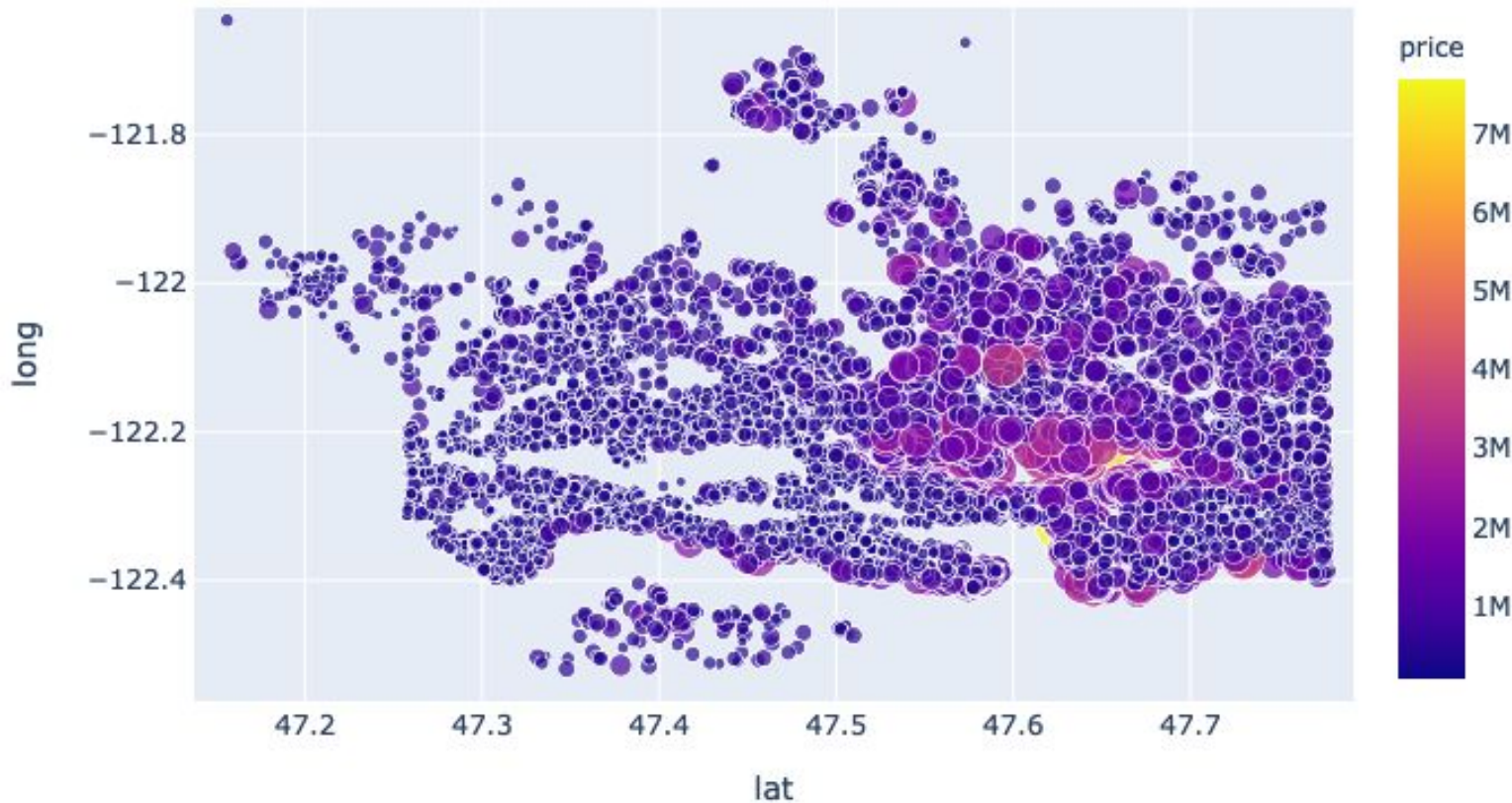
# Outlook

## Remember this chart?

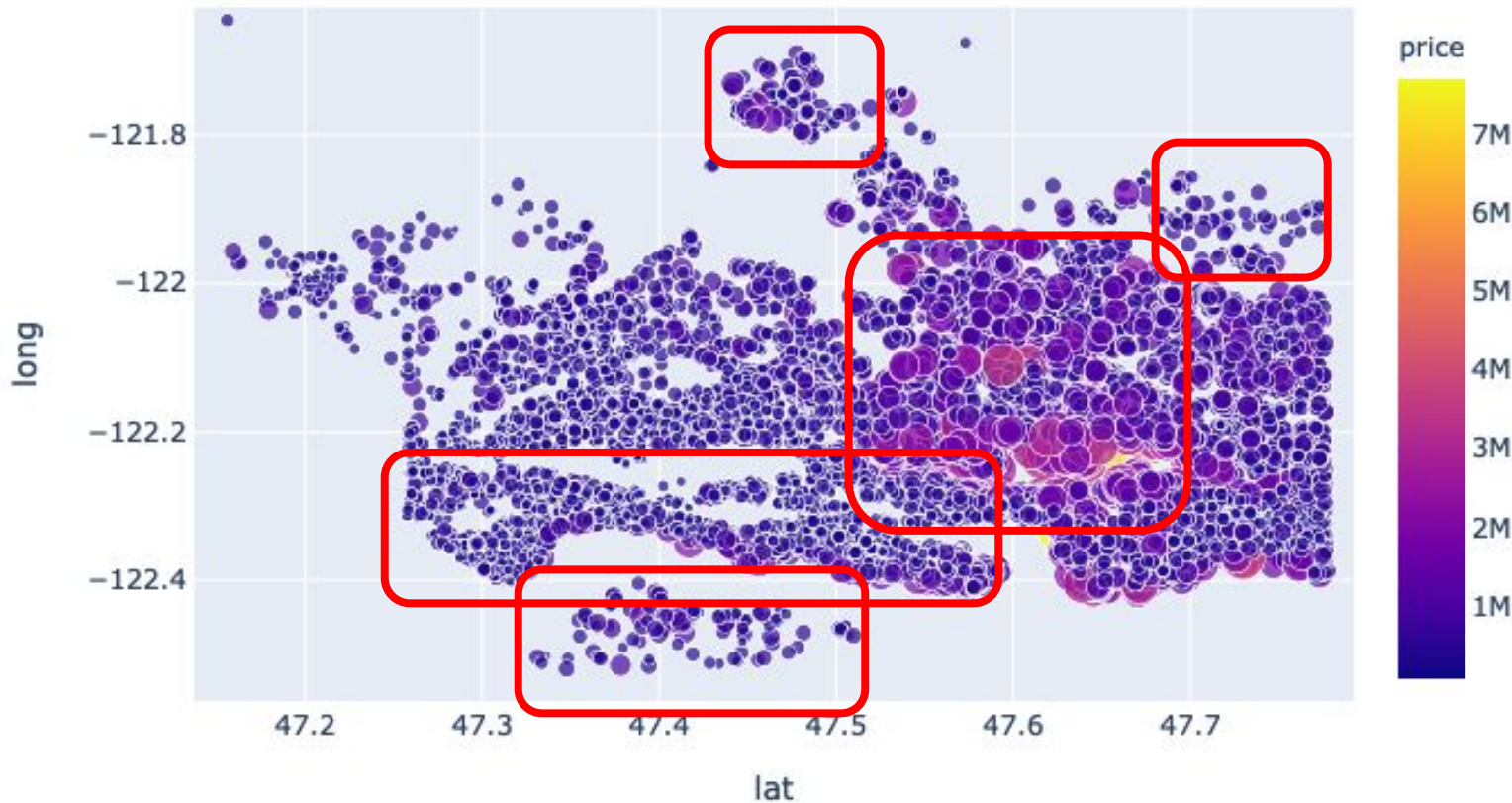


Some zipcode areas seem to have really cheap house prices.

# Clustering by region could be even better feature



# Clustering by region could be even better feature





Thanks for your attention