King County House Prices Exploratory Data Analyses

Stakeholder : Bonnie Brown (Seller)

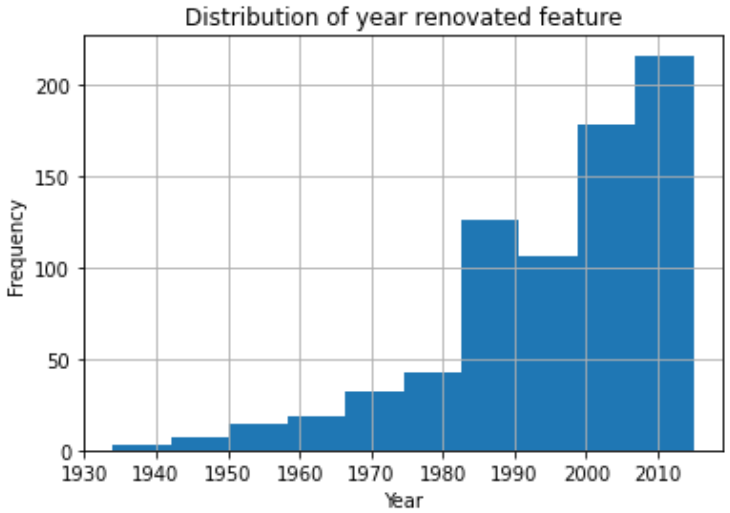
He as a house and wants to move soon (timing?), but wants high profit in middle class NH (neighborhood)

Hypothesis:

- House prices are high in city center
- waterfront Houses are expensive
- Best time for selling a house in mid of the year (Jun –July)
- After renovation the price will be increased

Data Cleaning

- **date** feature is converted to DateTime type
- **NaN** values are replaced:
  - waterfront 99% of the values are 0 therefore all NAN values in this data series are replaced by 0.
  - view 90% of the values are with 0's and this data is categorical variable data. In this analyses this data will be with 0's and 1's i.e. 0 = not viewed, 1 = viewed. Therefore all NAN values are replaced by 0's
    new feature = has_veiwed
  - yr_renovated this is also kind of categorical data like renovated = 1, not renovated = 0 in last 15 years. All renovations before 1990 are obsolete and assigned to 0. Similarly all NaNs are replaced with 0.
    - New feature = has_renovated



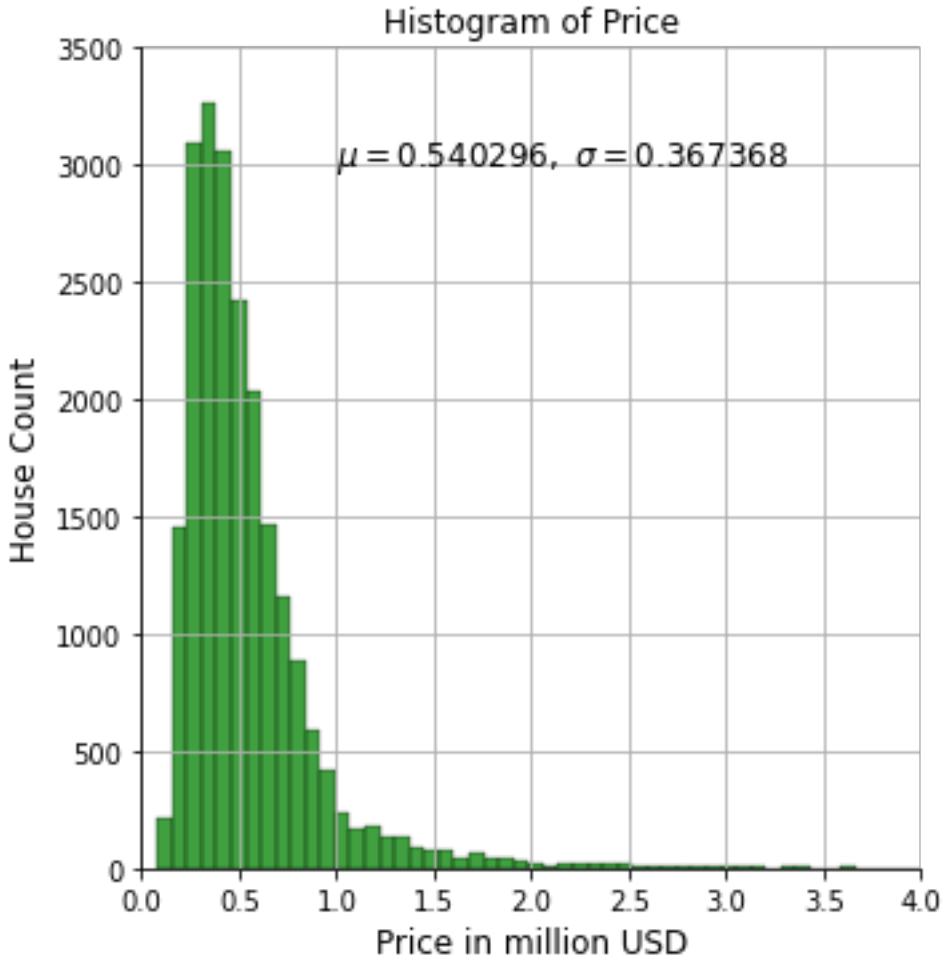
Data Cleaning

- The following analyses shows that sqft_basement feature is identified as object instead of float since it has `?` character in the dataset and 59% of the houses haven't basements.
- For quick analyses this feature also considered as categorical variable feature like house has basements = 1 or not = 0
- A new feature has_basement feature created and dropped sqft_basement feature

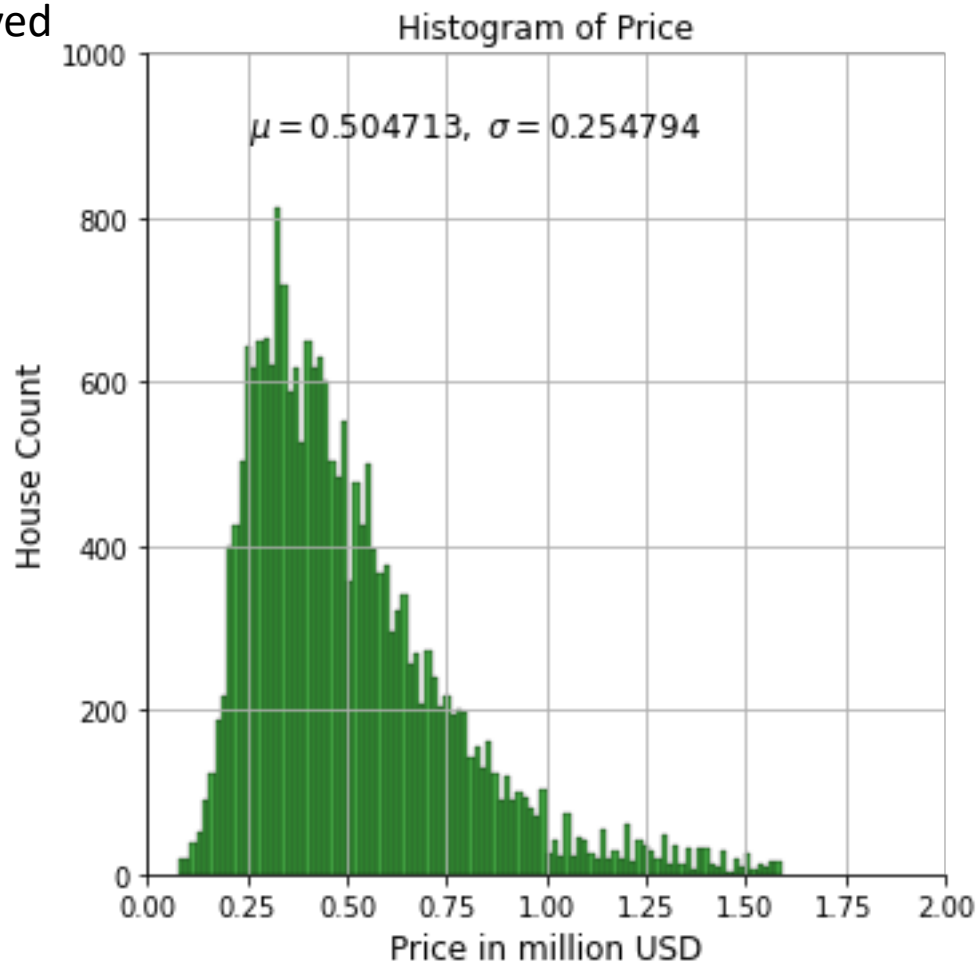
```
0.0       0.593879
?         0.021021
600.0     0.010048
500.0     0.009677
700.0     0.009631
            ...
1248.0    0.000046
283.0     0.000046
652.0     0.000046
3260.0    0.000046
276.0     0.000046
Name: sqft_basement, Length: 304, dtype: float64
```

```
count      21597
unique       304
top          0.0
freq       12826
Name: sqft_basement, dtype: object
```

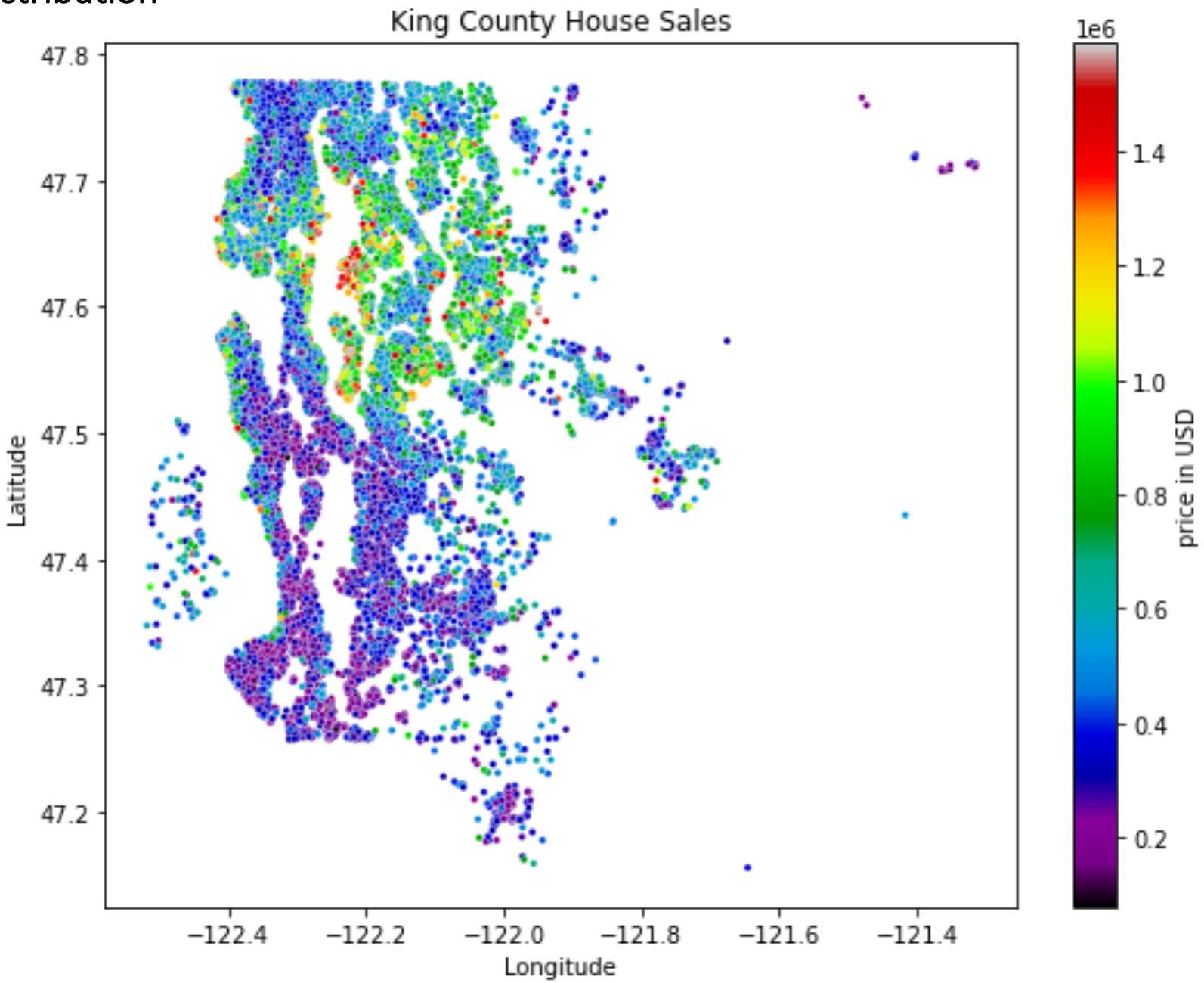
Analyses of price distribution



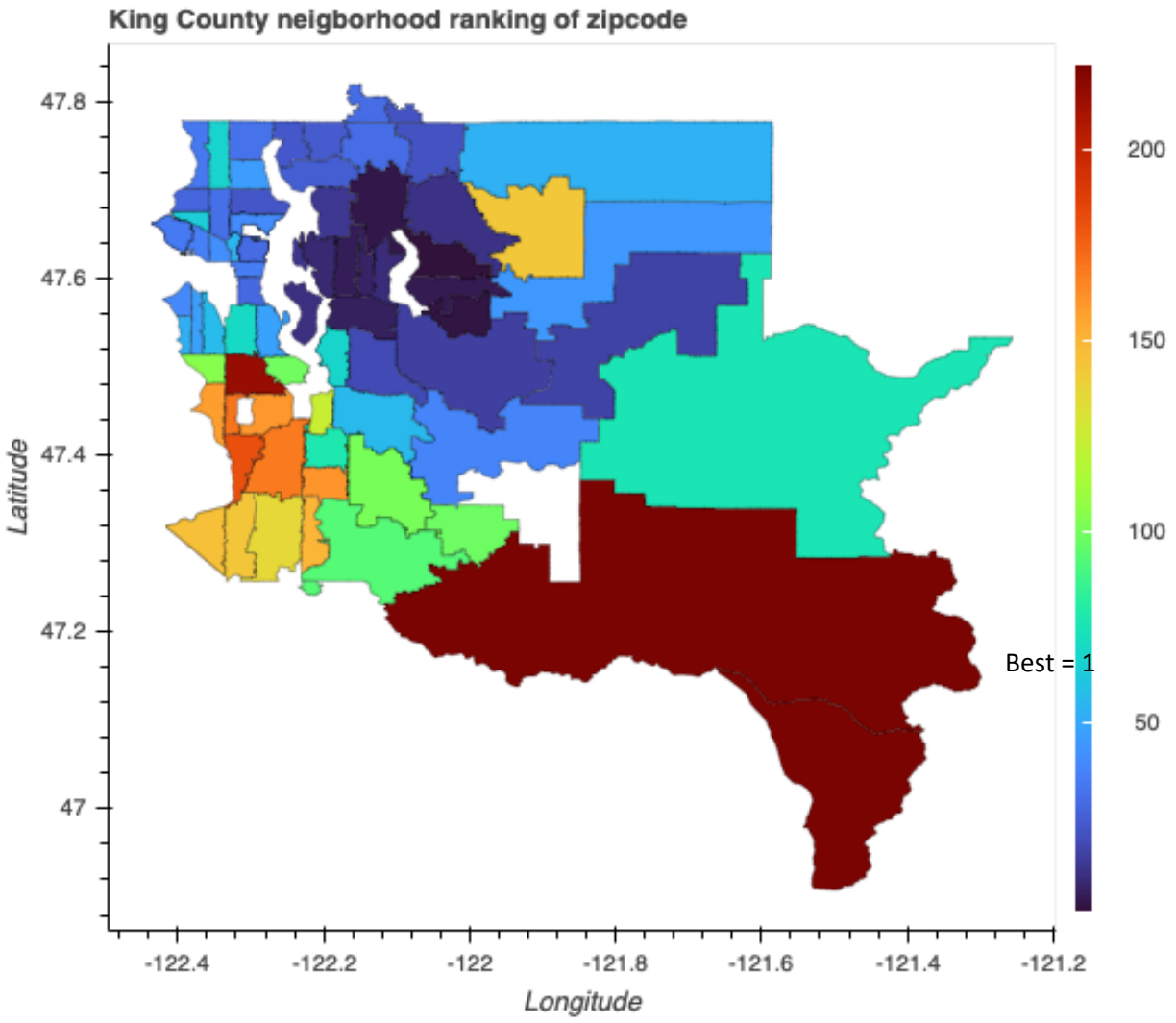
From price distribution graph, the houses with price above $1.6m are very few and can be considered as outliers. In this project for simplification the data of houses above 1.6m USD are removed

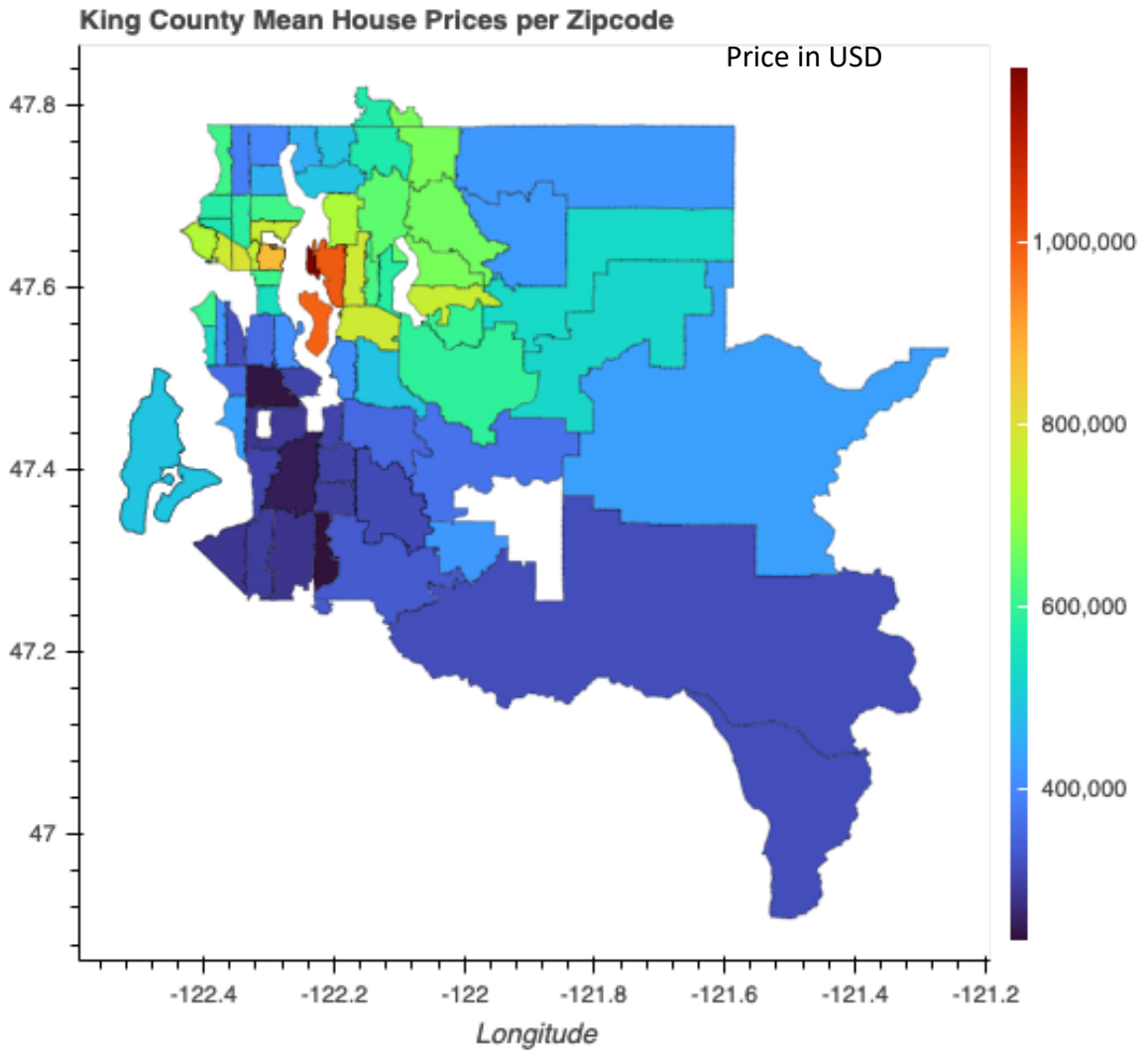


House price distribution



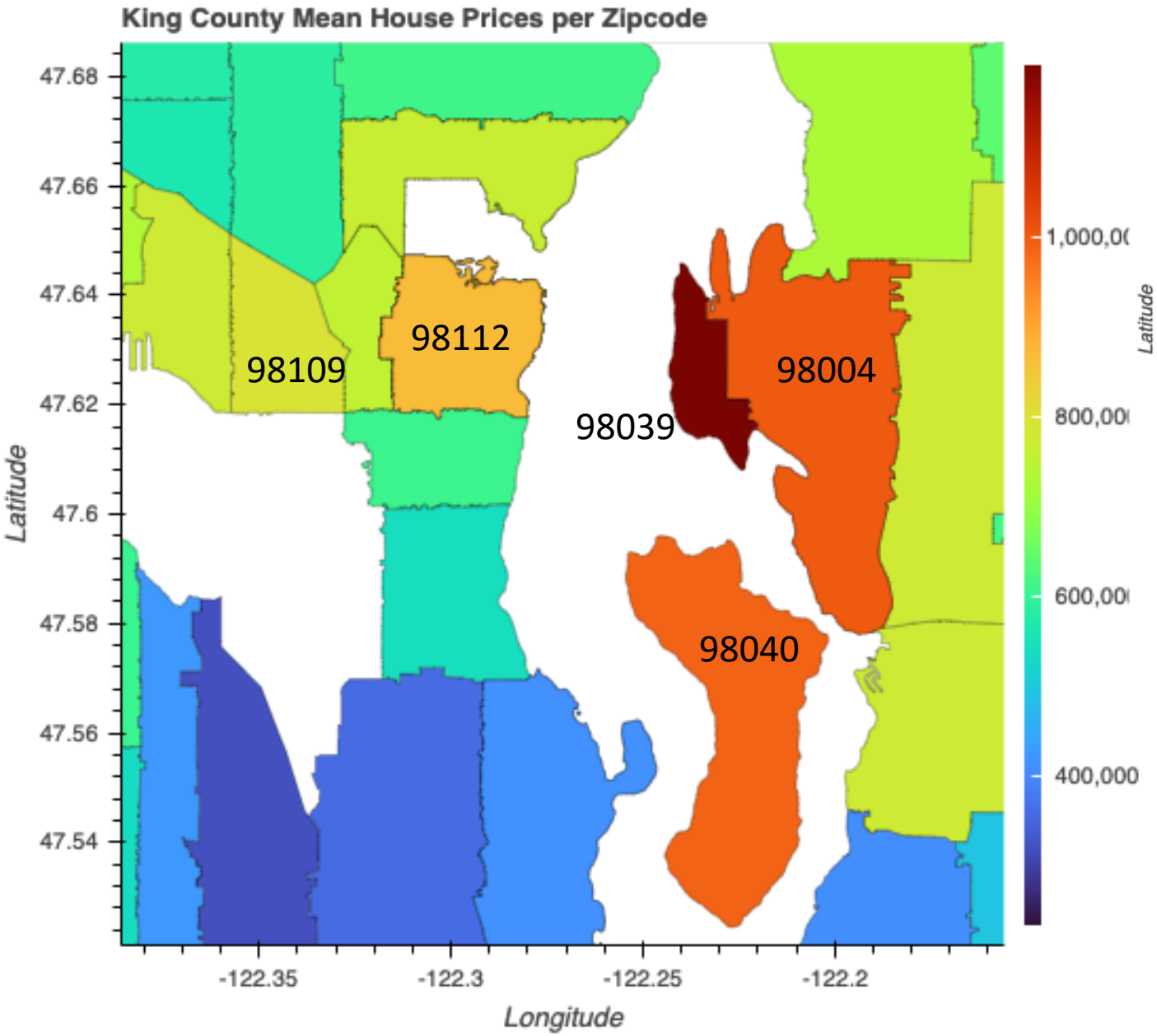
Analyses of price distribution and neighborhood

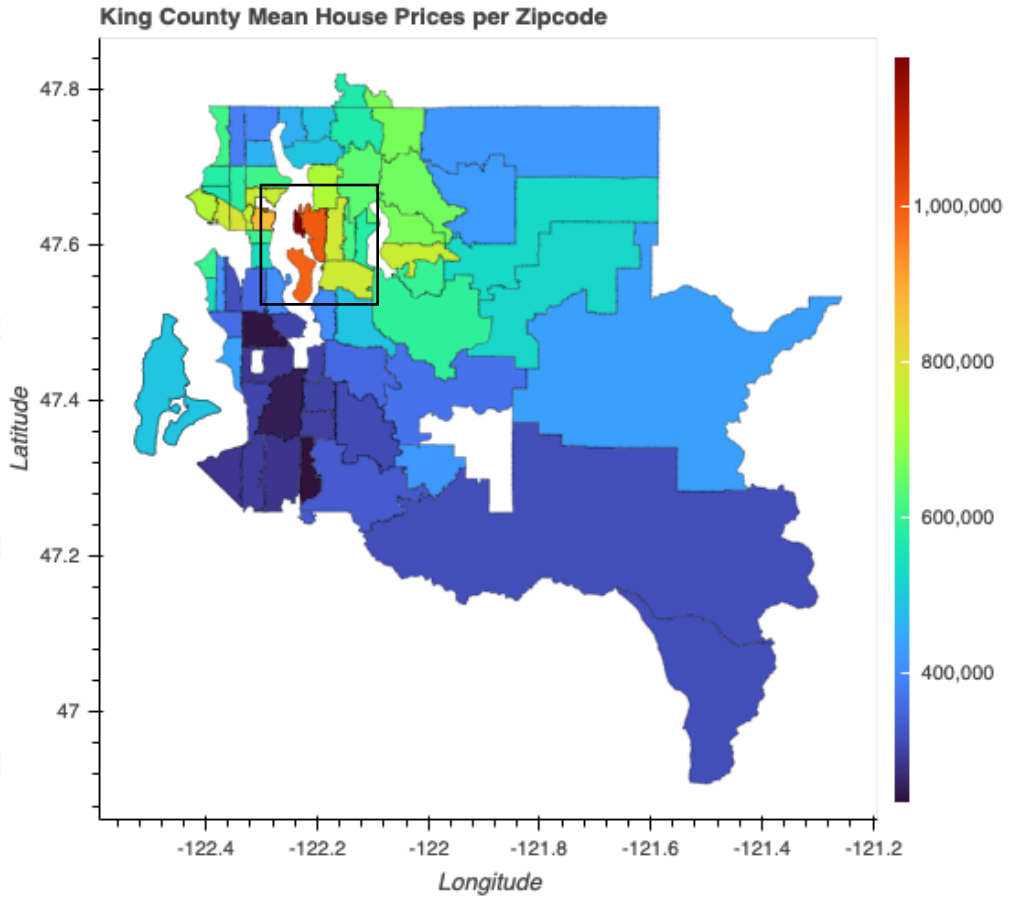




houses above 1.6mUSD are not considered in this analyses

Max mean price





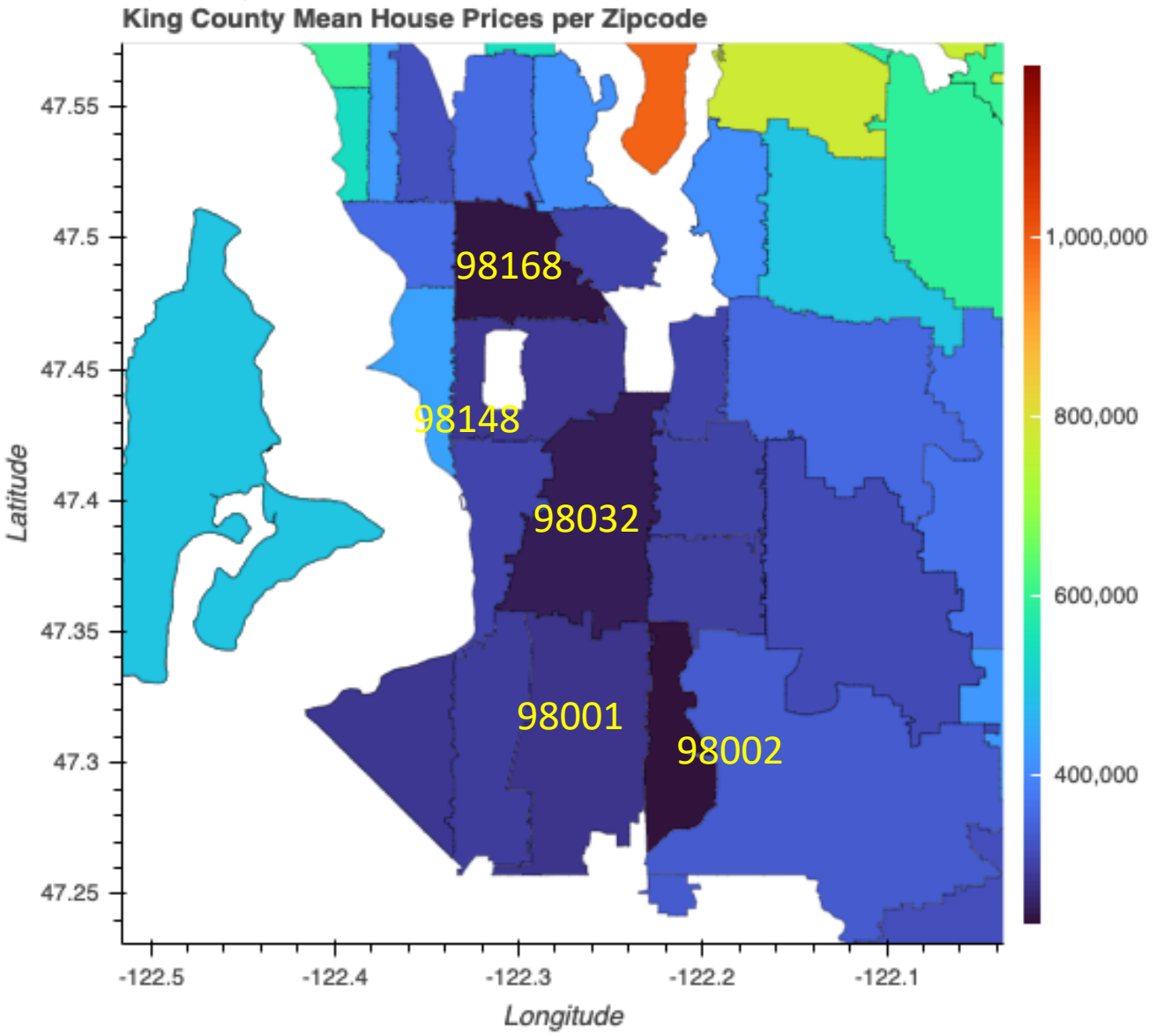
ZIPCODE	price
98039	1.193824e+06
98004	1.000978e+06
98040	9.909973e+05
98112	8.712536e+05
98109	7.955733e+05

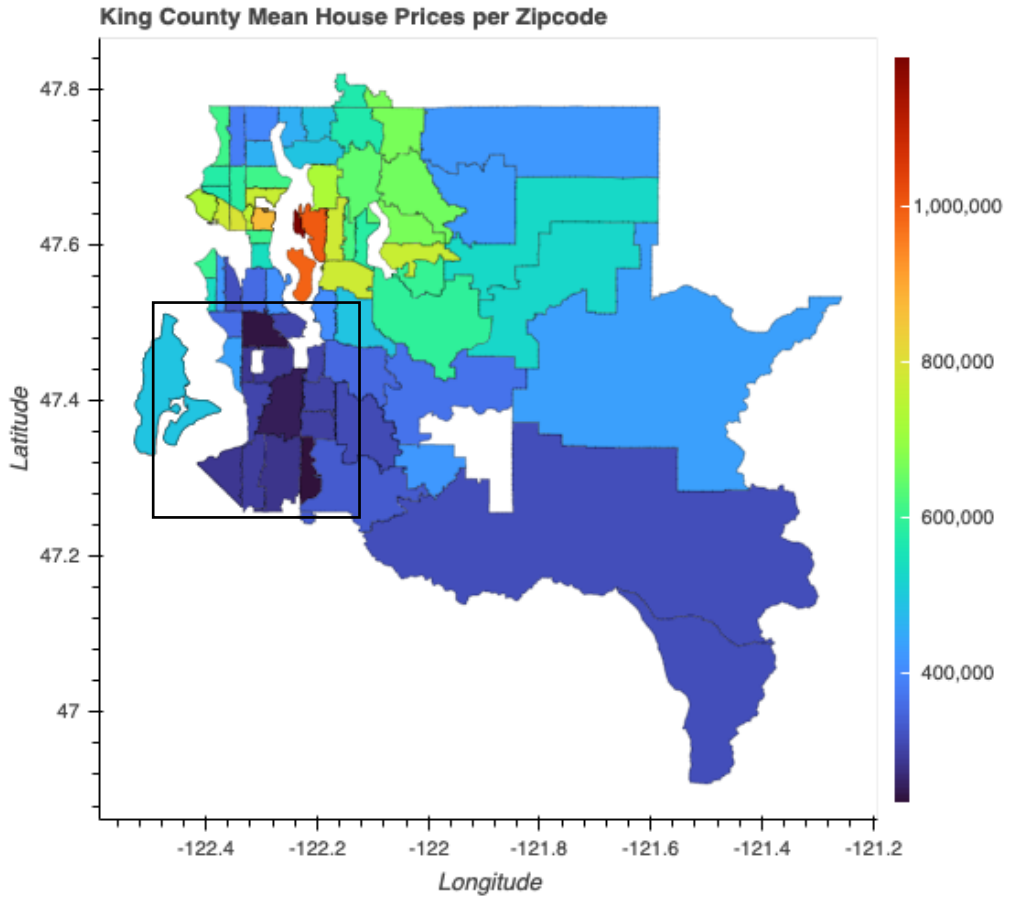
Top 5 places:

- Medina
- Clyde Hill /
- Mercer Island
- Madison Park
- Kenmore Air Harbor

houses above 1.6mUSD are not considered in this analyses

Min mean price





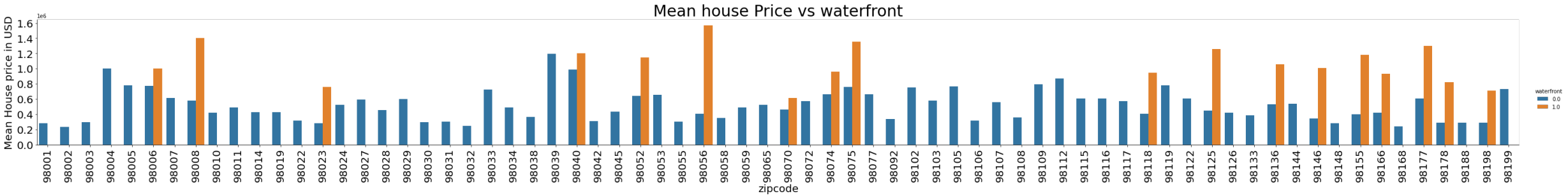
ZIPCODE	price
98002	234284.035176
98168	240328.371747
98032	251296.240000
98001	281194.869806
98148	284908.596491

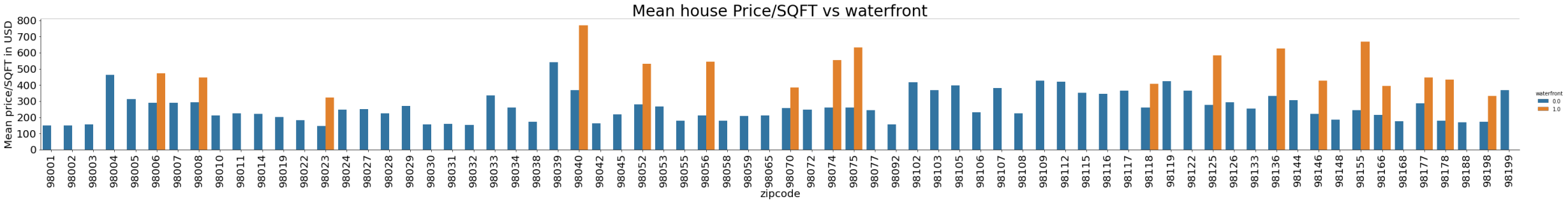
Bottom 5 places:

- South Auburn
- Burien
- Kent
- Auburn
- Burien

houses above 1.6mUSD are not considered in this analyses

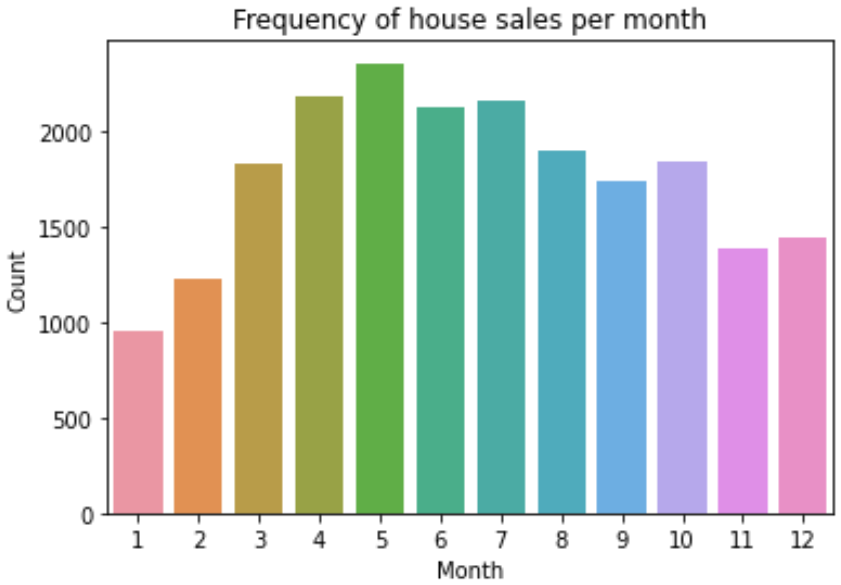
Sales, price vs waterfront

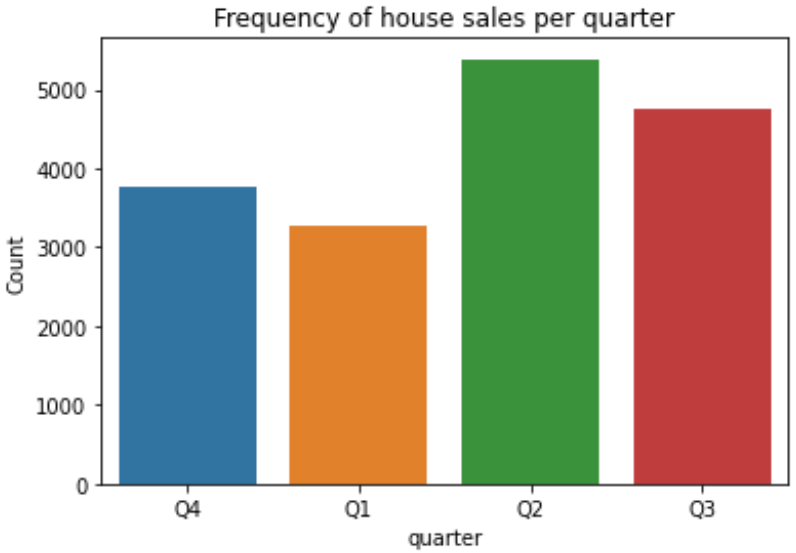


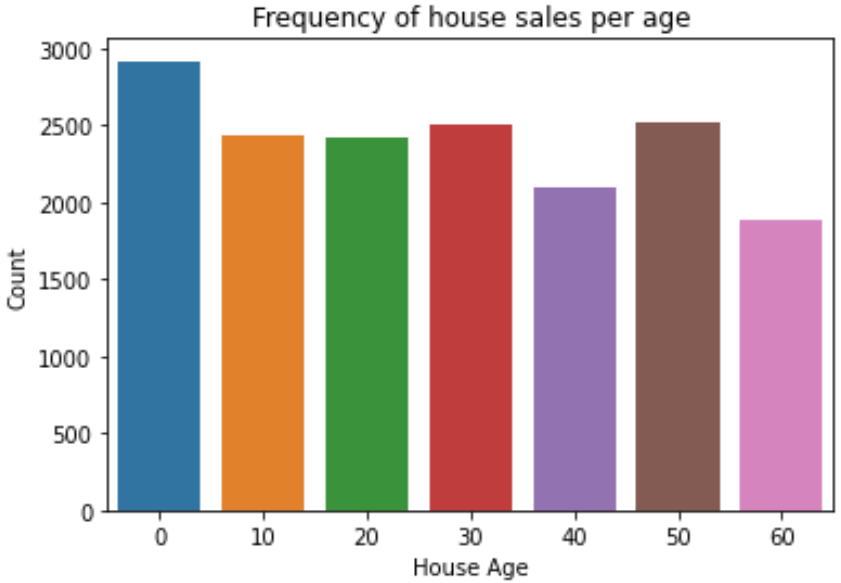


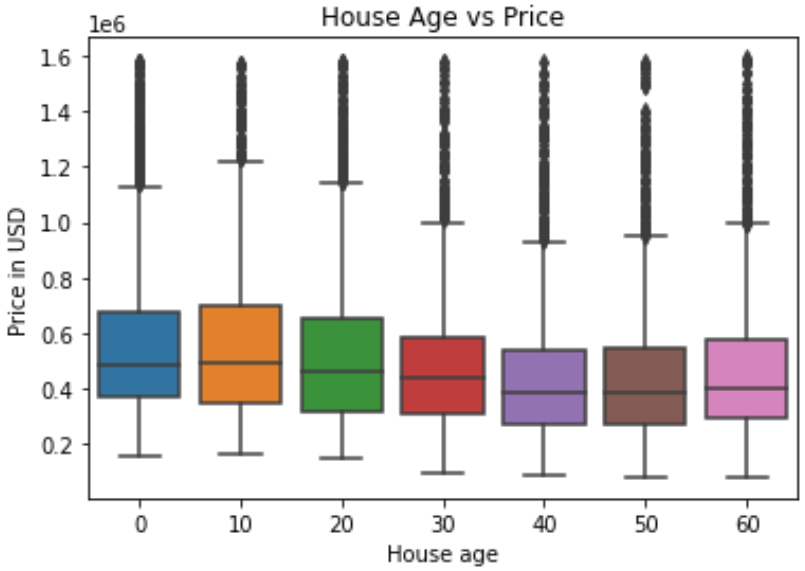
houses above 1.6mUSD are not considered in this analyses

House sales analyses



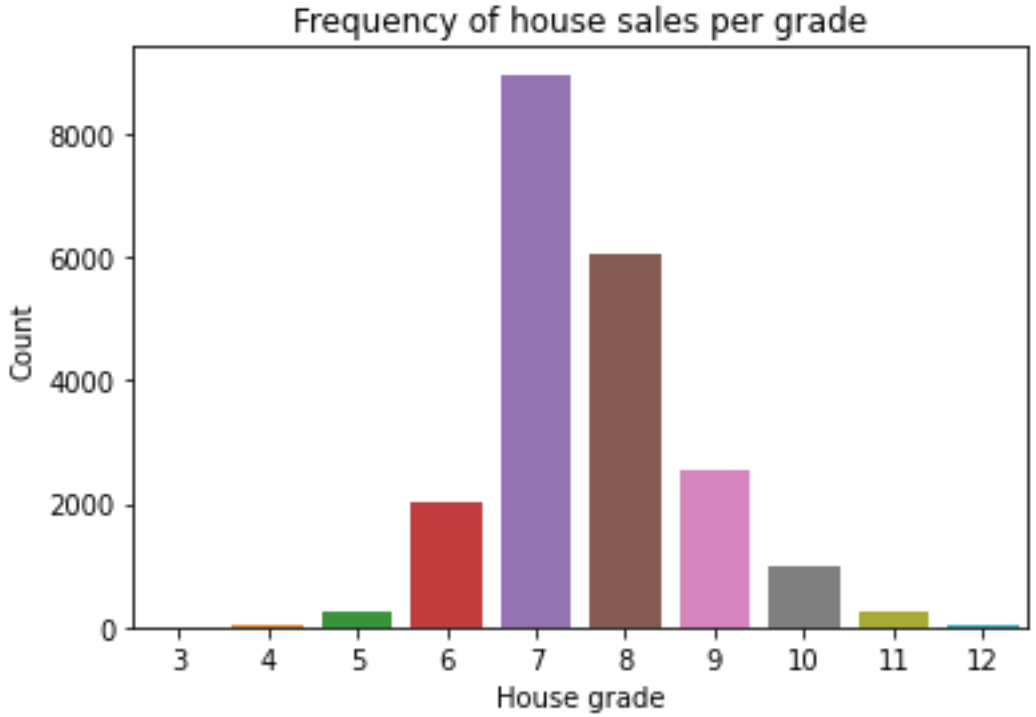




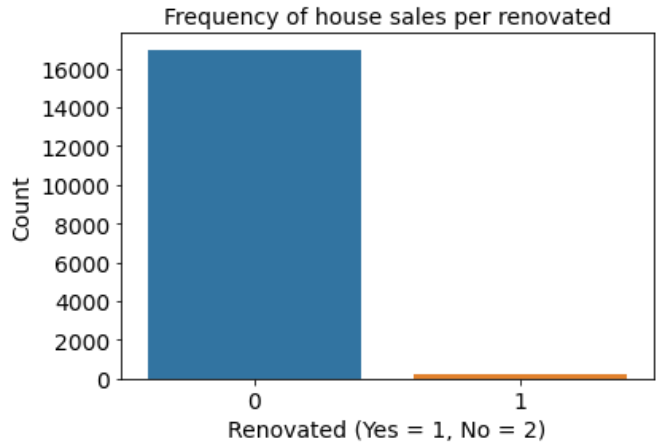


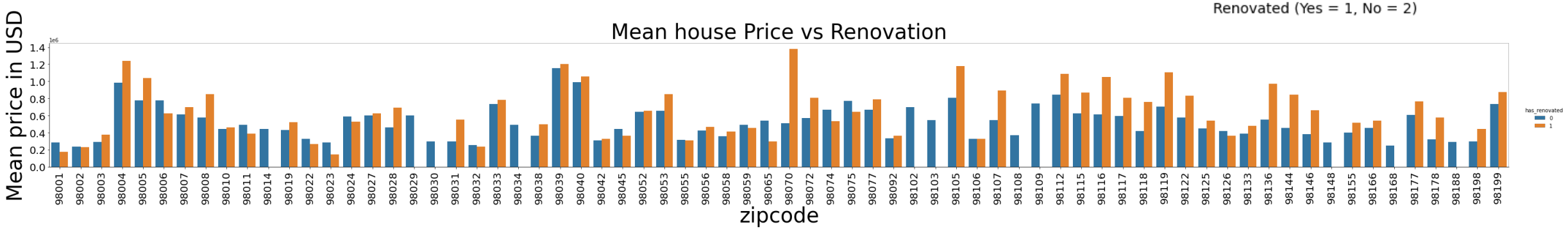
houses above 1.6mUSD are not considered in this analyses

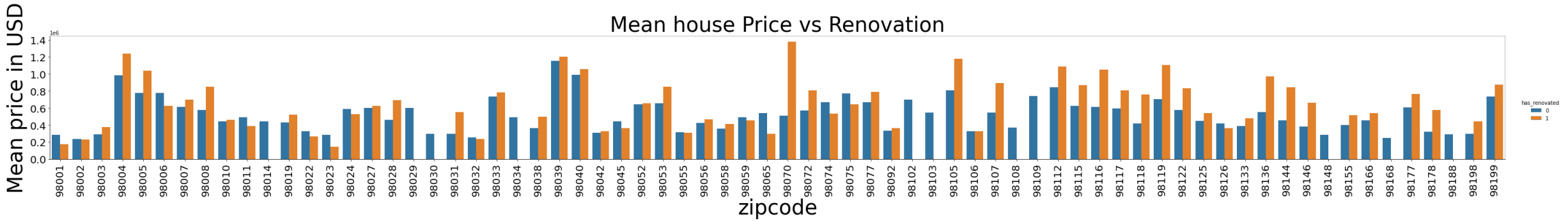
House sales per grade



Renovation vs sales







Post 1990renovations are obsolated, houses above 1.6mUSD are not considered in this analyses

Price vs waterfront

OLS Regression Results

Dep. Variable:	price	R-squared:	0.009
Model:	OLS	Adj. R-squared:	0.009
Method:	Least Squares	F-statistic:	153.2
Date:	Tue, 01 Jun 2021	Prob (F-statistic):	4.81e-35
Time:	10:43:31	Log-Likelihood:	-2.3826e+05
No. Observations:	17196	AIC:	4.765e+05
Df Residuals:	17194	BIC:	4.765e+05
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
Intercept	4.993e+05	1923.707	259.557	0.000	4.96e+05	5.03e+05
waterfront[T.1.0]	4.461e+05	3.6e+04	12.379	0.000	3.75e+05	5.17e+05

Omnibus:	4119.063	Durbin-Watson:	1.964
Prob(Omnibus):	0.000	Jarque-Bera (JB):	9083.646
Skew:	1.380	Prob(JB):	0.00
Kurtosis:	5.250	Cond. No.	18.8

Price vs renovation

OLS Regression Results

Dep. Variable:	price	R-squared:	0.008
Model:	OLS	Adj. R-squared:	0.008
Method:	Least Squares	F-statistic:	141.3
Date:	Tue, 01 Jun 2021	Prob (F-statistic):	1.87e-32
Time:	10:44:13	Log-Likelihood:	-2.3827e+05
No. Observations:	17196	AIC:	4.765e+05
Df Residuals:	17194	BIC:	4.766e+05
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
Intercept	4.979e+05	1934.494	257.399	0.000	4.94e+05	5.02e+05
has_renovated[T.1]	1.997e+05	1.68e+04	11.886	0.000	1.67e+05	2.33e+05

Price vs month

OLS Regression Results

Dep. Variable:	price	R-squared:	0.000
Model:	OLS	Adj. R-squared:	0.000
Method:	Least Squares	F-statistic:	3.092
Date:	Tue, 01 Jun 2021	Prob (F-statistic):	0.0787
Time:	10:53:02	Log-Likelihood:	-2.3834e+05
No. Observations:	17196	AIC:	4.767e+05
Df Residuals:	17194	BIC:	4.767e+05
Df Model:	1		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
Intercept	5.077e+05	4498.196	112.874	0.000	4.99e+05	5.17e+05
mth_sold	-1086.8705	618.081	-1.758	0.079	-2298.373	124.632

Omnibus:	4148.083	Durbin-Watson:	1.968
Prob(Omnibus):	0.000	Jarque-Bera (JB):	9163.102
Skew:	1.389	Prob(JB):	0.00
Kurtosis:	5.253	Cond. No.	17.2

Price vs all features (incl. dummies)

OLS Regression Results

Dep. Variable:	price	R-squared:	0.851
Model:	OLS	Adj. R-squared:	0.850
Method:	Least Squares	F-statistic:	996.2
Date:	Tue, 01 Jun 2021	Prob (F-statistic):	0.00
Time:	09:12:24	Log-Likelihood:	-2.2197e+05
No. Observations:	17196	AIC:	4.441e+05
Df Residuals:	17097	BIC:	4.449e+05
Df Model:	98		
Covariance Type:	nonrobust		

	coef	std err	t	P>|t|	[0.025	0.975]
Intercept	-2.614e+07	4.09e+06	-6.388	0.000	-3.42e+07	-1.81e+07
bedrooms[T.2]	-3.75e+04	1.12e+04	-3.346	0.001	-5.95e+04	-1.55e+04
bedrooms[T.3]	-3.914e+04	1.11e+04	-3.529	0.000	-6.09e+04	-1.74e+04
bedrooms[T.4]	-4.583e+04	1.12e+04	-4.079	0.000	-6.79e+04	-2.38e+04
bedrooms[T.5]	-5.996e+04	1.16e+04	-5.166	0.000	-8.27e+04	-3.72e+04
bedrooms[T.6]	-8.288e+04	1.34e+04	-6.195	0.000	-1.09e+05	-5.67e+04
bedrooms[T.7]	-1.569e+05	2.26e+04	-6.931	0.000	-2.01e+05	-1.12e+05
bedrooms[T.8]	-1.713e+05	3.9e+04	-4.386	0.000	-2.48e+05	-9.47e+04
bedrooms[T.9]	-1.702e+05	6.05e+04	-2.815	0.005	-2.89e+05	-5.17e+04
bedrooms[T.10]	-2.046e+05	7.11e+04	-2.879	0.004	-3.44e+05	-6.53e+04
bedrooms[T.33]	5.447e+04	9.88e+04	0.551	0.582	-1.39e+05	2.48e+05
bathrooms[T.1]	9906.0948	6.94e+04	0.143	0.887	-1.26e+05	1.46e+05
bathrooms[T.2]	-6224.8855	6.94e+04	-0.090	0.929	-1.42e+05	1.3e+05
bathrooms[T.3]	1.342e+04	6.94e+04	0.193	0.847	-1.23e+05	1.49e+05
bathrooms[T.4]	5.608e+04	6.95e+04	0.807	0.420	-8.01e+04	1.92e+05
bathrooms[T.5]	4.265e+04	7.23e+04	0.590	0.555	-9.9e+04	1.84e+05
bathrooms[T.6]	-5.67e+04	7.9e+04	-0.717	0.473	-2.12e+05	9.82e+04
bathrooms[T.7]	-3.736e+05	1.22e+05	-3.060	0.002	-6.13e+05	-1.34e+05
bathrooms[T.8]	-1.702e+05	6.05e+04	-2.815	0.005	-2.89e+05	-5.17e+04