

HOUSE SALE PRICES PREDICTION

Nopasorn Kowathanakul





THE PROJECT

From the sales data in King County, I will do the data analysis and select influence factors for making a prediction model. Then, the following questions will be answered:

1. How does the location affect the price?
2. Do built and renovated times affect the price?
3. Does selling time affect the price?
4. Can we predict the price from a condition?

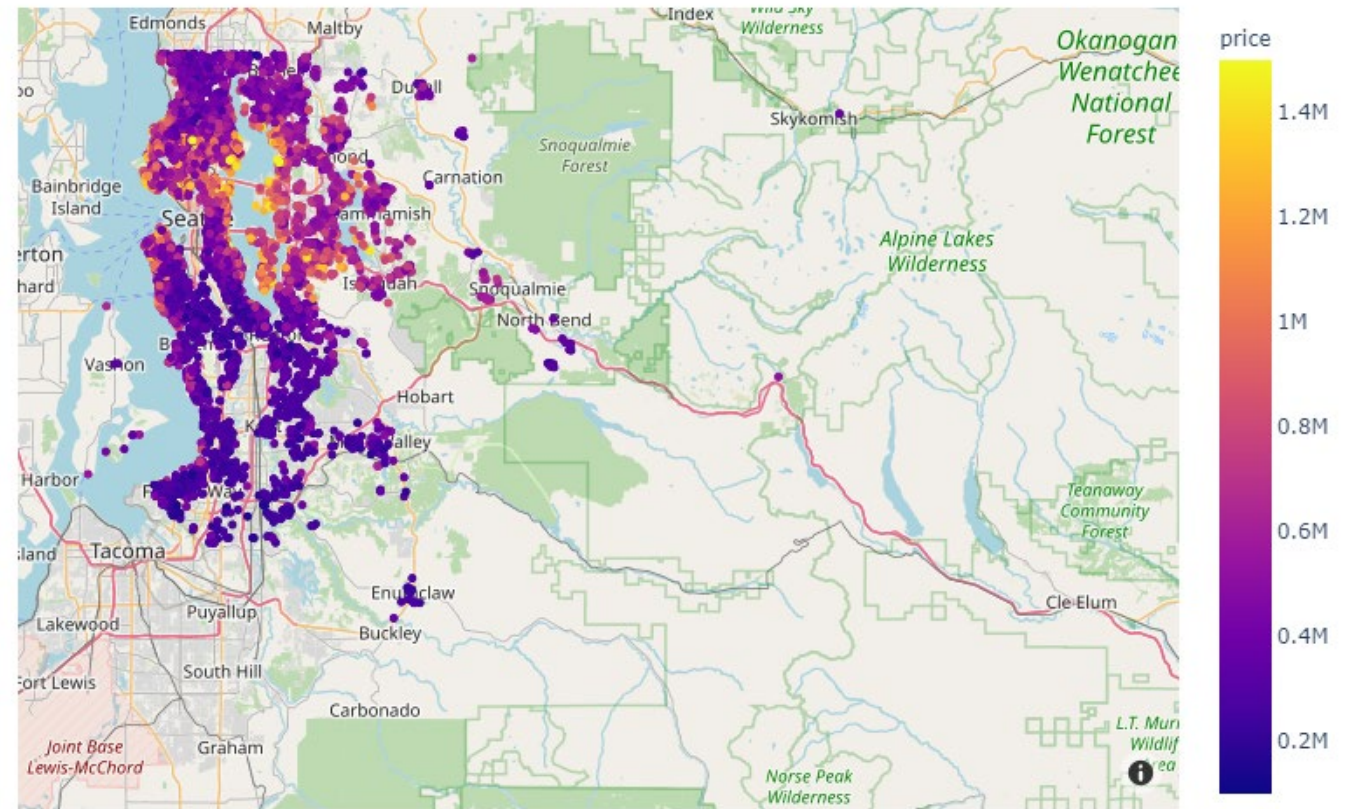
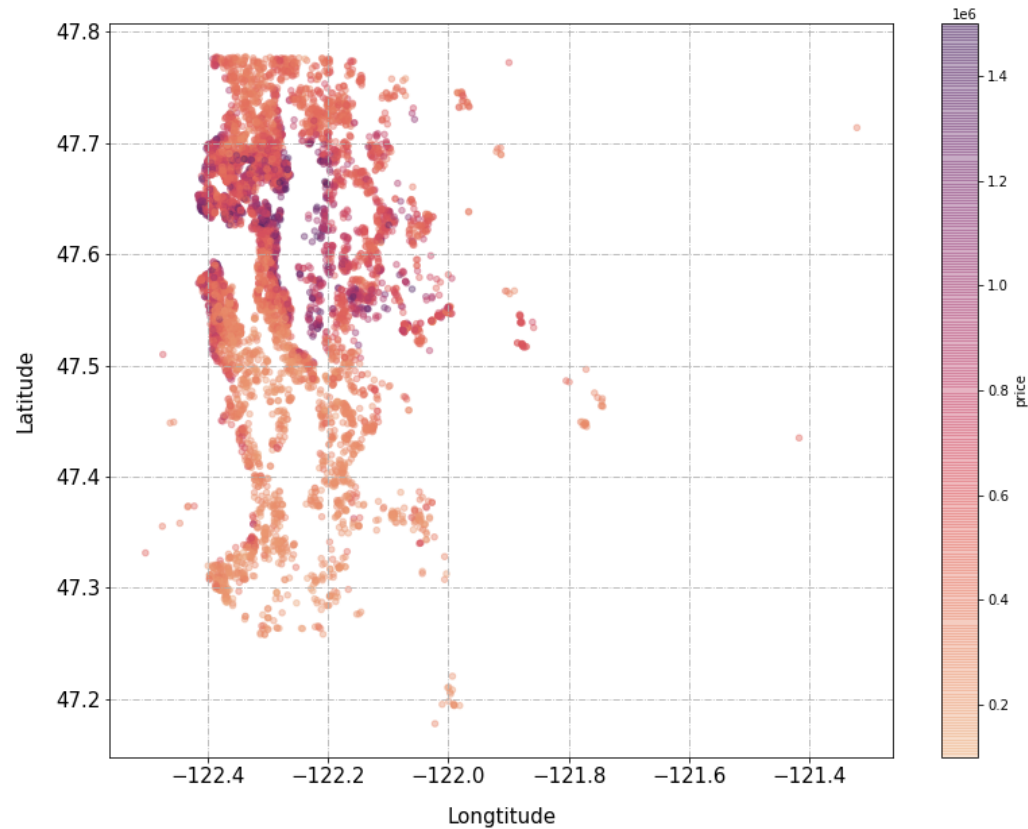


King County House Sales dataset

ID - UNIQUE IDENTIFIED FOR A HOUSE
DATE - DATE HOUSE WAS SOLD
PRICE - PRICE IS PREDICTION TARGET
BEDROOMS - NUMBER OF BEDROOMS/HOUSE
BATHROOMS - NUMBER OF BATHROOMS/BEDROOMS
SQFT_LIVING - SQUARE FOOTAGE OF THE HOME
SQFT_LOT - SQUARE FOOTAGE OF THE LOT
FLOORS - TOTAL FLOORS (LEVELS) IN HOUSE
WATERFRONT - HOUSE WHICH HAS A VIEW TO A WATERFRONT
VIEW - HAS BEEN VIEWED
CONDITION - HOW GOOD THE CONDITION IS (OVERALL)
GRADE - OVERALL GRADE GIVEN TO THE HOUSING UNIT, BASED ON KING COUNTY GRADING SYSTEM
SQFT_ABOVE - SQUARE FOOTAGE OF HOUSE APART FROM BASEMENT
SQFT_BASEMENT - SQUARE FOOTAGE OF THE BASEMENT
YR_BUILT - BUILT YEAR
YR_RENOVATED - YEAR WHEN HOUSE WAS RENOVATED
ZIPCODE - ZIP
LAT - LATITUDE COORDINATE
LONG - LONGITUDE COORDINATE
SQFT_LIVING15 - THE SQUARE FOOTAGE OF INTERIOR HOUSING LIVING SPACE FOR THE NEAREST 15 NEIGHBORS
SQFT_LOT15 - THE SQUARE FOOTAGE OF THE LAND LOTS OF THE NEAREST 15 NEIGHBORS



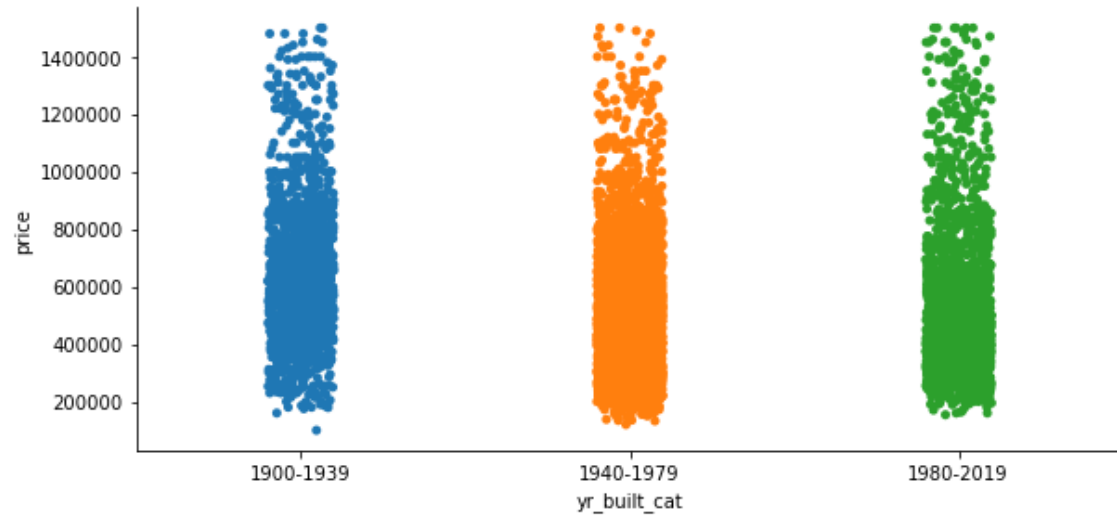
LOCATION EFFECT



BUILT AND RENOVATED TIMES

```
Mean prices for yr_built_cat
1900-1939    639911.185238
1940-1979    503076.052348
1980-2019    538176.536572
Name: price, dtype: float64
```

```
Median prices for yr_built_cat
1900-1939    597000.0
1940-1979    459975.0
1980-2019    475000.0
Name: price, dtype: float64
```



```
Mean prices for renovated
original      532330.073109
renovated     731288.941176
Name: price, dtype: float64
```

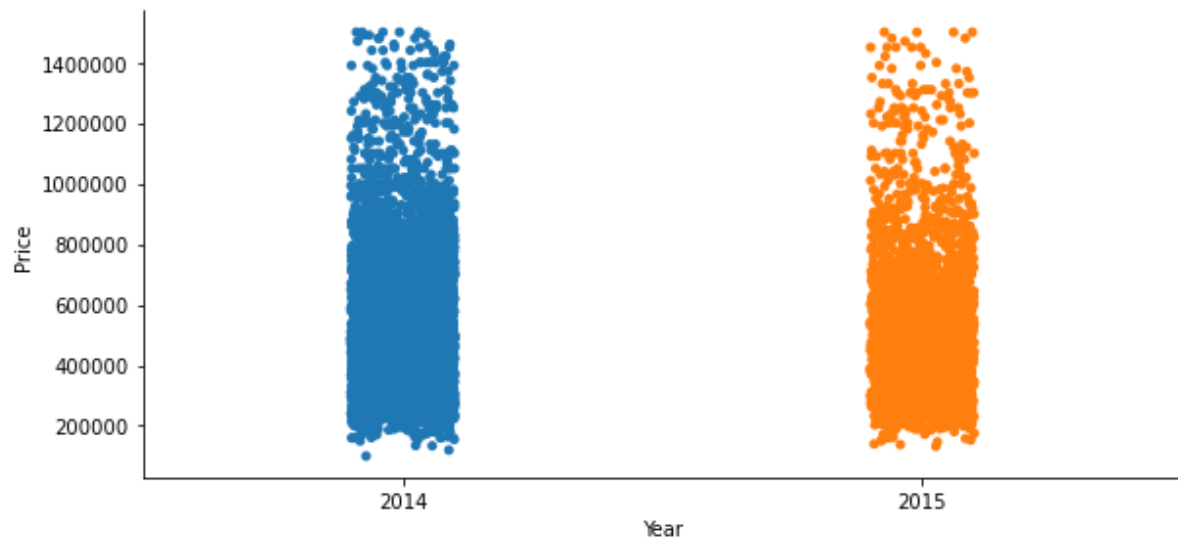
```
Median prices for renovated
original      484000.0
renovated     721000.0
Name: price, dtype: float64
```



SELLING TIME

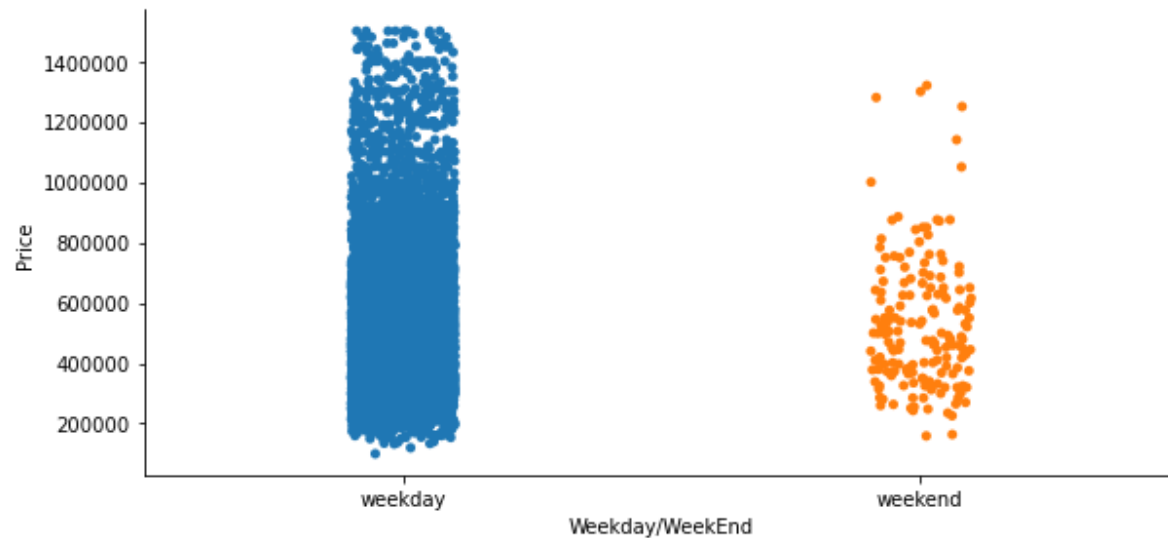
```
Mean prices for year
2014    538130.161706
2015    544165.609507
Name: price, dtype: float64
```

```
Median prices for year
2014    489000.0
2015    499000.0
Name: price, dtype: float64
```



```
Mean prices for day_cat
weekday    540430.406543
weekend    525963.441860
Name: price, dtype: float64
```

```
Median prices for day_cat
weekday    490000.0
weekend    484000.0
Name: price, dtype: float64
```



PRICE PREDICTION

$$\begin{aligned} \text{Est.Price} = & 0.23x \text{ sqft_living} + 0.12x \text{ sqft_above} + 0.12x \text{ sqft_basement} + 0.09x \text{ sqft_living15} \\ & - 0.05x \text{ bedrooms} - 0.08x \text{ bathrooms} + 0.33x \text{ grade} + 0.22x \text{ condition} + 0.11x \text{ floors} \end{aligned}$$

OLS Regression Results						
Dep. Variable:	price	R-squared (uncentered):	0.962			
Model:	OLS	Adj. R-squared (uncentered):	0.962			
Method:	Least Squares	F-statistic:	1.568e+04			
Date:	Mon, 24 May 2021	Prob (F-statistic):	0.00			
Time:	20:31:38	Log-Likelihood:	3965.2			
No. Observations:	5590	AIC:	-7912.			
Df Residuals:	5581	BIC:	-7853.			
Df Model:	9					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
sqft_living	0.2066	0.031	6.655	0.000	0.146	0.268
sqft_above	0.0447	0.031	1.438	0.151	-0.016	0.106
sqft_basement	0.0574	0.023	2.450	0.014	0.011	0.103
sqft_living15	0.1871	0.016	11.872	0.000	0.156	0.218
bedrooms	-0.0697	0.014	-5.154	0.000	-0.096	-0.043
bathrooms	-0.0634	0.013	-4.727	0.000	-0.090	-0.037
grade	0.3986	0.018	22.281	0.000	0.364	0.434
condition	0.2557	0.013	19.667	0.000	0.230	0.281
floors	0.1125	0.008	14.657	0.000	0.097	0.128
Omnibus:	65.643	Durbin-Watson:	2.009			
Prob(Omnibus):	0.000	Jarque-Bera (JB):	62.470			
Skew:	-0.227	Prob(JB):	2.72e-14			
Kurtosis:	2.750	Cond. No.	52.6			

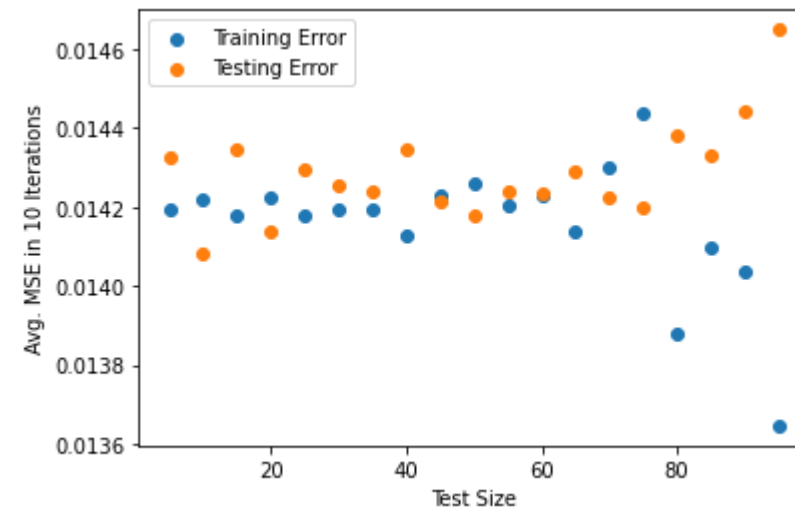
Notes:

[1] R² is computed without centering (uncentered) since the model does not contain a constant.

[2] Standard Errors assume that the covariance matrix of the errors is correctly specified.

Train Mean Squarred Error: 0.01417106149621563

Test Mean Squarred Error: 0.014334953915793956



FUTURE PLAN

For the future, I should include more detail about location, such as downtown, shopping mall, supermarket or public transportation, to increase the accuracy of prediction.





THANK YOU