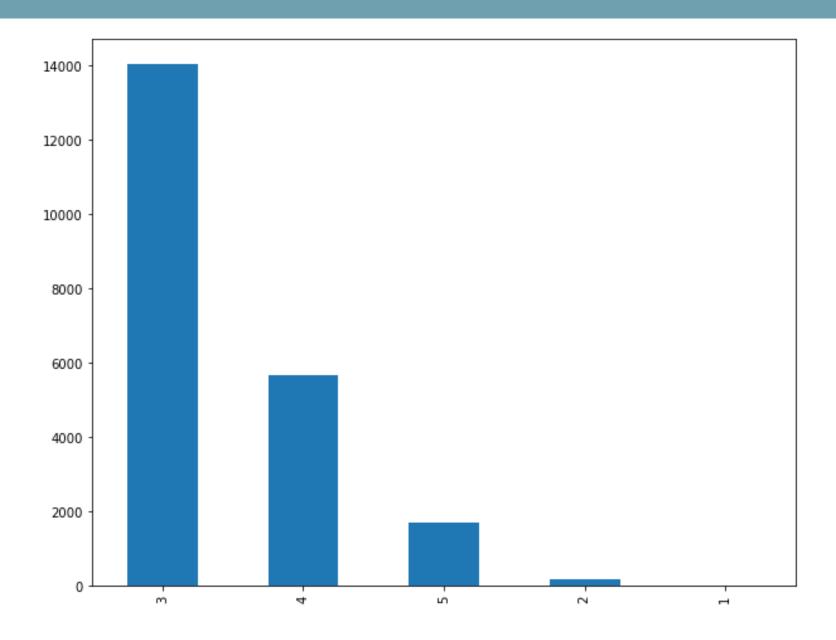
# Assignment

### Assignment 03

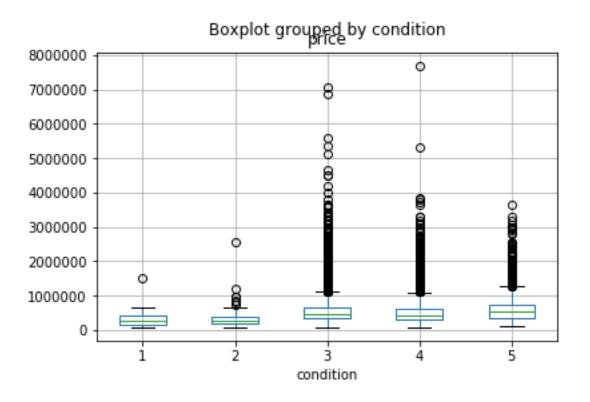
- Add categorical variables to variable set
  - 'view', 'condition', 'grade'
  - Interpret the results
- Ideas to utilize zipcode, lat, and long
  - Without other resources
    - Data manipulation approach based on these three variables
  - With other resources
    - Additional useful information for these three variables

- Illustrate your ideas using Power Point
  - Some students have to create a video clip to explain their results and ideas

# **Condition**

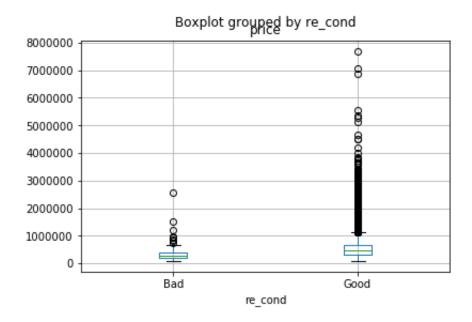


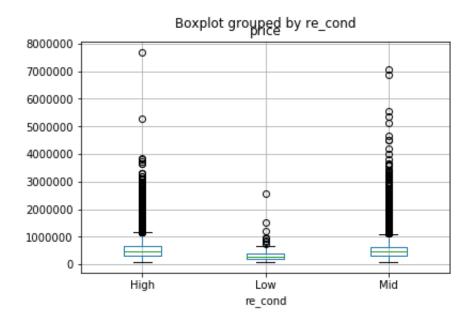
# **Condition**



### **Re-Binning**

- Good and Bad
  - □ Good if condition> 2
- High, Mid, Low
  - High if condition > 3
  - Mid if condition= 3





# **OLS Model 1**

#### OLS Regression Results

==========			========			
Dep. Variable:		price R-squared:		0.598		
Model:		OLS Adj. R-squared:		0.598		
Method: Least:		east Squares	F-statistic:		1/16.	
Date: Tue, 2		22 Sep 2020 Prob (F-statistic):		0.00		
Time:		10:26:26	Log-Likelihood:		-2.3834e+05	
No. Observations:		17290	17290 AIC:		4.767e+05	
Df Residuals:		17274	BIC:		4.768e+05	
Df Model:		15				
Covariance Typ		nonrobust				
	coef	std err	t	P> t	[0.025	0.975]
const	6.124e+06	1.7e+05	36.029	0.000	5.79e+06	6.46e+06
bedrooms	-5.814e+04	2395.751	-24.269	0.000	-6.28e+04	-5.34e+04
bathrooms	6.384e+04	4234.975	15.075	0.000	5.55e+04	7.21e+04
sqft_lot	-0.0092	0.061	-0.150	0.881	-0.129	0.110
floors	5.747e+04	4563.913	12.592	0.000	4.85e+04	6.64e+04
waterfront	6.74e+05	2.12e+04	31.745	0.000	6.32e+05	7.16e+05
sqft_above	240.1351	4.352	55.183	0.000	231.605	248.665
sqft_basement	243.6856	5.424	44.931	0.000	233.055	254.316
yr_built	-3256.4363	85.181	-38.230	0.000	-3423.400	-3089.472
yr_renovated	17.5940	4.695	3.747	0.000	8.391	26.797
sqft_living15	94.8250	4.129	22.964	0.000	86.731	102.919
sqft_lot15	-0.7240	0.091	-7.967	0.000	-0.902	-0.546
cond_2	7.765e+04	5.4e+04	1.437	0.151	-2.83e+04	1.84e+05
cond_3	1.182e+05	5.02e+04	2.353	0.019	1.97e+04	2.17e+05
cond_4	1.277e+05	5.03e+04	2.541	0.011	2.92e+04	2.26e+05
cond_5	1.682e+05	5.05e+04	3.328	0.001	6.91e+04	2.67e+05
Omnibus: 11985.		11985.490	Durbin-Watson:			2.013
Prob(Omnibus):		0.000	Jarque-Bera (JB):		607402.206	
Skew:		2.753	Prob(JB): 0.00		0.00	
Kurtosis:		31.510	Cond. No.			4.94e+06

# **OLS Model 2**

#### OLS Regression Results

Dep. Variable: price			R-squared:		0.598	
Model:		0LS	Adi. R-squared:		0.597	
Method: Le		east Squares.	F-statistic:		1975.	
Date: Tue,		22 Sep 2020	Prob (F-statistic):		0.00	
Time:		10:27:21	Log-Likelihood:		-2.3835e+05	
No. Observations:		17290 AIC:		4.767e+05		
Df Residuals:		17276	BIC:		4.768e+05	
Df Model:		13				
Covariance Type:		nonrobust				
	coef	std err	t	P>¦t¦	[0.025	0.975]
const	6.386e+06	1.62e+05	39.402	0.000	6.07e+06	6.7e+06
bedrooms	-5.814e+04	2397.825	-24.247	0.000	-6.28e+04	-5.34e+04
bathrooms	6.571e+04	4225.364	15.552	0.000	5.74e+04	7.4e+04
sqft_lot	-0.0078	0.061	-0.127	0.899	-0.127	0.112
floors	5.836e+04	4564.724	12.786	0.000	4.94e+04	6.73e+04
waterfront	6.751e+05	2.12e+04	31.777	0.000	6.33e+05	7.17e+05
sqft_above	239.7818	4.355	55.062	0.000	231.246	248.318
sqft_basement	244.5173	5.426	45.061	0.000	233.881	255.153
yr_built	-3321.7808	84.351	-39.381	0.000	-3487.117	-3156.445
yr_renovated	16.1085	4.691	3.434	0.001	6.913	25.304
sqft_living15	94.1772	4.131	22.798	0.000	86.080	102.274
sqft lot15	-0.7268	0.091	-7.991	0.000	-0.905	-0.549
re_cond_Low	-6.997e+04	1.91e+04	-3.656	0.000	-1.07e+05	-3.25e+04
re_cond_Mid	-1.767e+04	4233.380	-4.174	0.000	-2.6e+04	-9373.403
					=======	
Omnibus:		11925.661	Durbin-Watson:		2.014	
Prob(Omnibus):		0.000	Jarque-Bera (JB): 596060.7			
Skew:		2.736	Prob(JB):		0.00	
Kurtosis:		31.239	Cond. No	· 		4.62e+06

### Summary

train,test =train\_test\_split(house, test\_size=0.2, random\_state=30)

- Model 1
  - MSE for test set = 49845975570.53953
- Model 2
  - MSE for test set = 49791693775.603615

### **Dummy Variables**

- If there are too many dummy variables compared to the number of samples, training a model may not work well
  - If there are other variables that can describe the characteristics of each category of the specific categorical variable
  - If there are too many categories, reduce the number of categories by checking the relationship between the variable and the target