# **EDA-Project**

House prices of King County by Marek Nowaczewski

## Given data

Variables	Number of entities: 21597		
id	sqft basement		
date	yr built		
price	yr_renovated		
bedrooms	zipcode		
bathrooms	lat		
sqft_living	long		
sqft_lot	sqft_living15		
floors	sqft_lot15		
waterfront	spec_price		
view	lo_li_rat		
condition	waterfront_str		
grade	ziporder		
sqft_above			

## Calculated columns for analysis

### Variables

- specific price
- ratio between lot size and living size

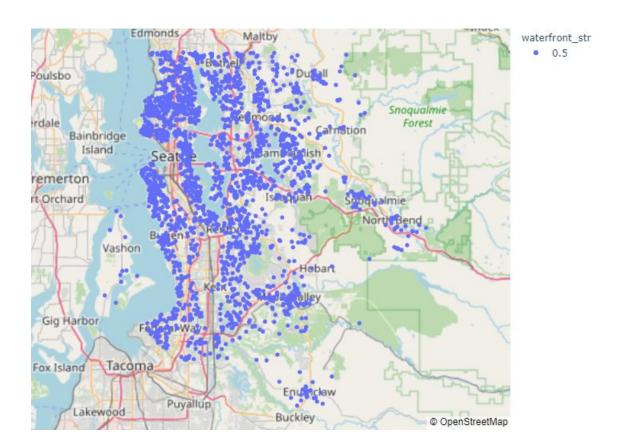
### Derivate / calculated data

```
df['spec_price'] = df.price / df.sqft_living
df['lo_li_rat'] = df.sqft_lot / df.sqft_living
```

## Missing values

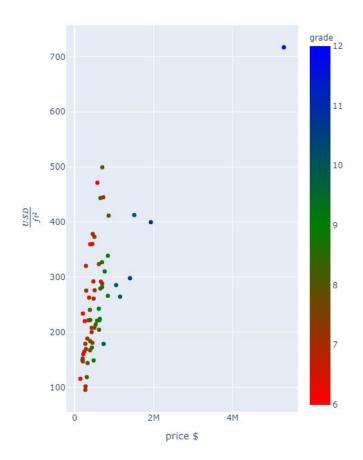
	counts	percentage
waterfront	2376	11.00
view	63	0.29
yr_renovated	3842	17.79

## Missing values waterfront



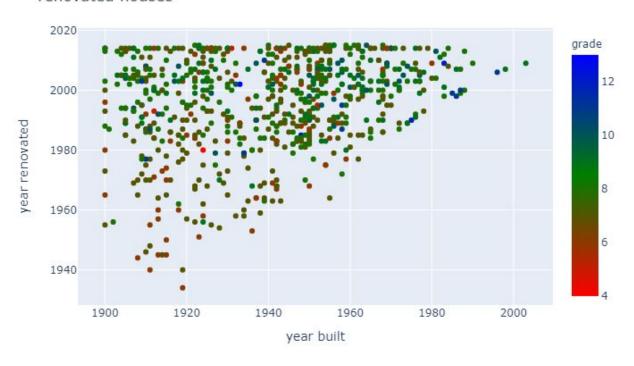
## Missing values view

	price	grade	spec_price	counts
view				
0.0	4.968061e+05	7.566214	256.894200	19422
1.0	8.133733e+05	8.115152	320.076130	330
2.0	7.913904e+05	8.315569	304.420288	957
3.0	9.732852e+05	8.730315	323.027575	508
4.0	1.452466e+06	9.063091	434.540453	317



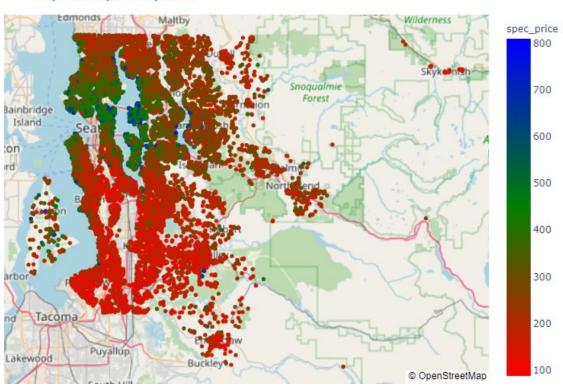
## Missing values yr\_renovated

#### renovated houses



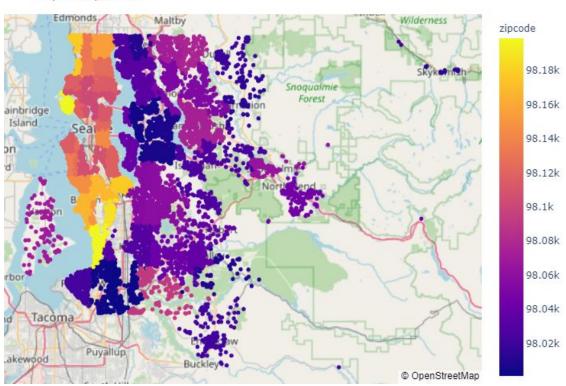
## Location, location and location....

Map with specific prices



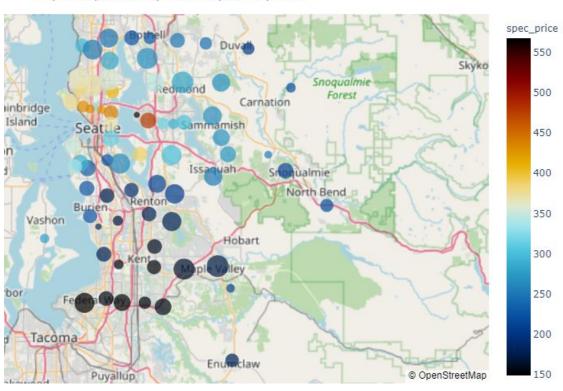
### Location, location and location.... #2

Map of zipcodes



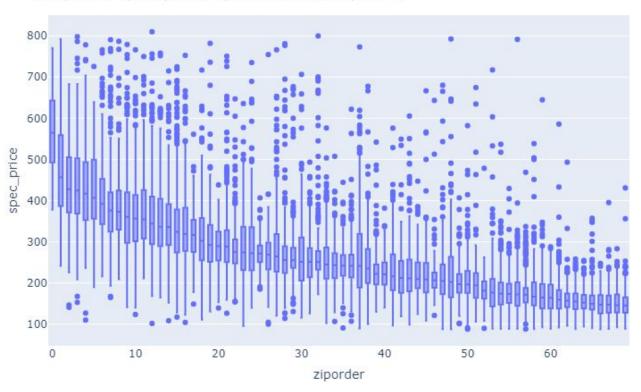
### Location, location and location.... #3

Map of zipcodes by mean specific prices



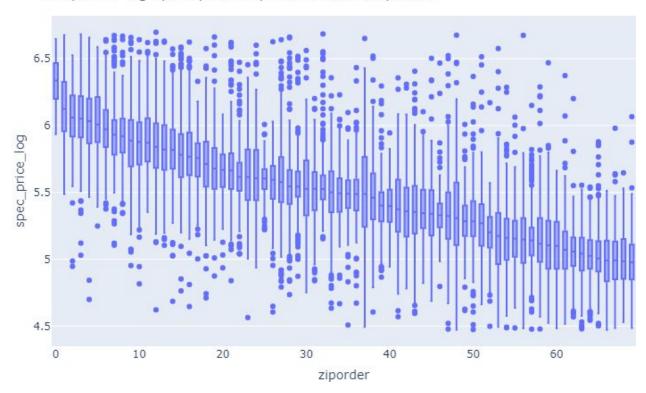
## Order of zipcodes

Boxplot of spec. prices by new ordered zipcodes



## Order of zipcodes

Boxplot of log spec. prices by new ordered zipcodes



### Correlations



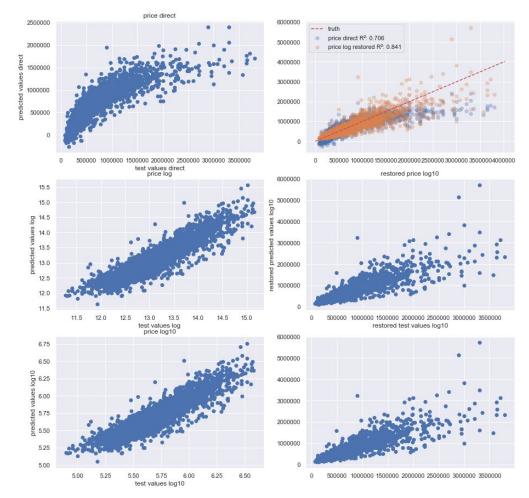
## **Linear Regression**

#### here used

- ziporder
- sqft\_living
- grade
- yr\_built
- long
- condition

### tragets

- price
- In(price)
- log10(price)



### **Predictions**

### price

#### OLS Regression Results

#### Dep. Variable: R-squared: 0.706 Model: OLS Adj. R-squared: 0.705 Least Squares F-statistic: Method: 5775. Wed, 10 Jun 2020 Prob (F-statistic): Date: 0.00 Time: Log-Likelihood: 10:07:57 -1.9710e+05 No. Observations: AIC: 14469 3.942e+05 Df Residuals: 14462 BIC: 3.943e+05 Df Model: Covariance Type: nonrobust

	coef	std err	t	P> t	[0.025	0.975]
const	-3.721e+05	1.73e+06	-0.215	0.830	-3.76e+06	3.02e+06
x1	-6784.3794	96.683	-70.171	0.000	-6973.891	-6594.868
x2	204.0673	2.878	70.902	0.000	198.426	209.709
x3	8.325e+04	2454.583	33.915	0.000	7.84e+04	8.81e+04
x4	-1577.6081	81.014	-19.473	0.00	-1/30.400	011
x5	-2.555e+04	1.36e+04	-1.872	0.061	-5.23e+04	1200.643
x6	2.029e+04	2633.930	7.704	0.000	1.51e+04	2 55-104

### log(price)

OL	S Regression	Results

Dep. Variable:		У	R-squar	ed:		0.841
Model:		OLS	Adj. R-	squared:		0.841
Method:	L	east Squares	F-stati	stic:		1.274e+04
Date:	Wed,	10 Jun 2020	Prob (F	-statistic)		0.00
Time:		10:07:57	Log-Lik	elihood:		2060.4
No. Observations:		14469	AIC:			-4107.
Df Residuals:		14462	BIC:			-4054.
Df Model:		6				
Covariance Type:		nonrobust				
CC	oef	std err	t	P> t	[0.025	0.975]

	coef	std err	t	P> t	[0.025	0.975]
const	32.0037	1.820	17.583	0.000	28.436	35.571
x1	-0.0144	0.000	-142.061	0.000	-0.015	-0.014
x2	0.0002	3.03e-06	78.986	0.000	0.000	0.000
x3	0.1347	0.003	52.171	0.000	0.130	0.140
x4	-0.0010	8.52e-05	-11.179	0.000	0.000	9 991
x5	0.1496	0.014	10.419	0.000	0.121	0.178
х6	0.0497	0.003	17.933	0.000		ورون. درون.

## **Predictions**



### Outlook

#### ToDo

- search / optimization / selection of input data
- filtering outliers
- usage of a squared regression?.
- other handling of zipcodes other categorical data
- usage of dummies
- interactive plots in presentation

# Thank you!