# Real Estate Pricing Model

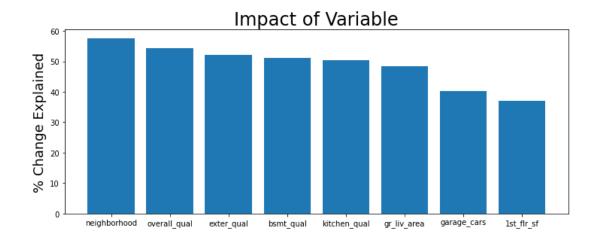


	feature	original_feature	engineering	type	r2_score
0	neighborhood_mode_imputed	neighborhood	mode_imputed	categorical	0.575692
1	overall_qual_mean_imputed^2	overall_qual	mean_imputed^2	numeric	0.54345
2	exter_qual_mode_imputed	exter_qual	mode_imputed	categorical	0.522003
3	bsmt_qual_na_val_imputed	bsmt_qual	na_val_imputed	categorical	0.510579
4	kitchen_qual_mode_imputed	kitchen_qual	mode_imputed	categorical	0.504355
5	gr_liv_area_mean_imputed	gr_liv_area	mean_imputed	numeric	0.483091
6	garage_cars_mean_imputed^2	garage_cars	mean_imputed^2	numeric	0.403627
7	1st_flr_sf_mean_imputed	1st_flr_sf	mean_imputed	numeric	0.371196

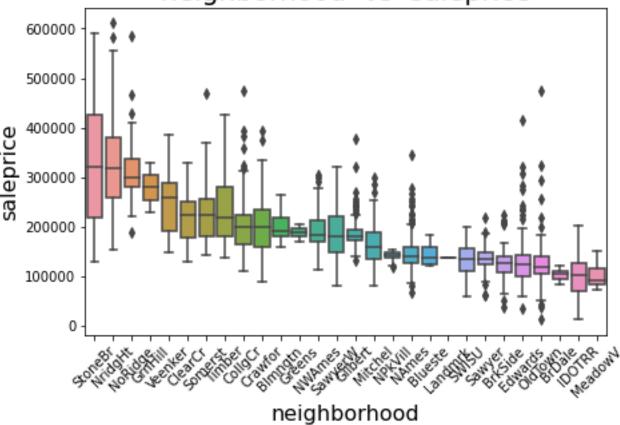
### Main Model Summary

						==
Dep. Variable:	saleprice	R-squared	(uncentered)	:	0.9	69
Model:	0LS		ared (uncente	ered):	0.9	
Method: L					803	
	14 Oct 2022					.00
Time:	00:39:14		hood:		-2435	
No. Observations:	2051				4.873e+	
Df Residuals:	2043	BIC:			4.877e+	-64
Df Model:	8					
Covariance Type:	nonrobust					
	coef	std err	t	P> t	[0.025	0.975]
neighborhood_mode_imputed	0.1845	0.020	9.103	0.000	0.145	0.224
overall qual mean imputed					1515.478	1814.318
exter_qual_mode_imputed			-0.072		-0.049	
bsmt_qual_na_val_imputed					-0.029	
kitchen_qual_mode_imputed					0.031	
gr_liv_area_mean_imputed						
garage_cars_mean_imputed^	2 5078.6186	364.248			4364.283	
lst_flr_sf_mean_imputed	13.9753	2.418	5.780	0.000	9.234	18.717
Omnibus:	654.421	Durbin-Wat	======== son:		1.990	
Prob(Omnibus):	0.000		a (JB):			
Skew:	1.042				0.00	
Kurtosis:	14.286			1.7	2e+05	
					=====	

- R<sup>2</sup> is computed without centering (uncentered) since the model does not contain a constant.
   Standard Errors assume that the covariance matrix of the errors is correctly specified.
   The condition number is large, 1.72e+05. This might indicate that there are strong multicollinearity or other numerical problems.



## neighborhood vs saleprice



Variable Type: categorical

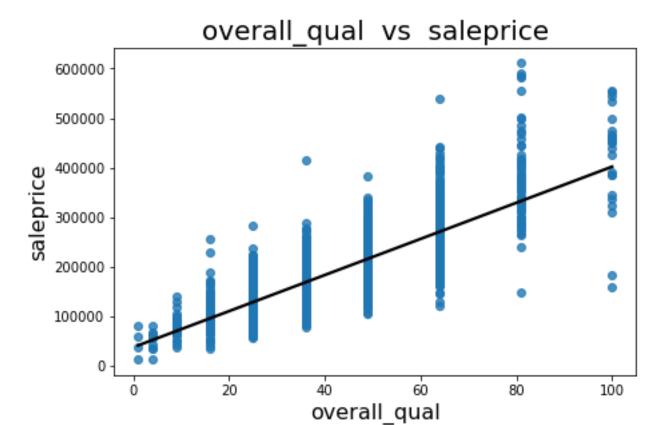
Engineering: mode\_imputed

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observati Df Residuals: Df Model: Covariance Ty	ons:		st Sc 4 Oct 00:	OLS juares 2022	R-squared Adj. R-squa F-statistic Prob (F-sta Log-Likelik AIC: BIC:	ired (uncen :: atistic):	,
	coef	std err		t	P> t	[0.025	0.975]
saleprice	0.9071	0.005	167	.685	0.000	0.896	0.918
Omnibus: Prob(Omnibus) Skew: Kurtosis:	:	0. -0.	000 962				1.840 2479.441 0.00 1.00

#### Notes:

[1] R $^2$  is computed without centering (uncentered) since the model does not contain [2] Standard Errors assume that the covariance matrix of the errors is correctly s



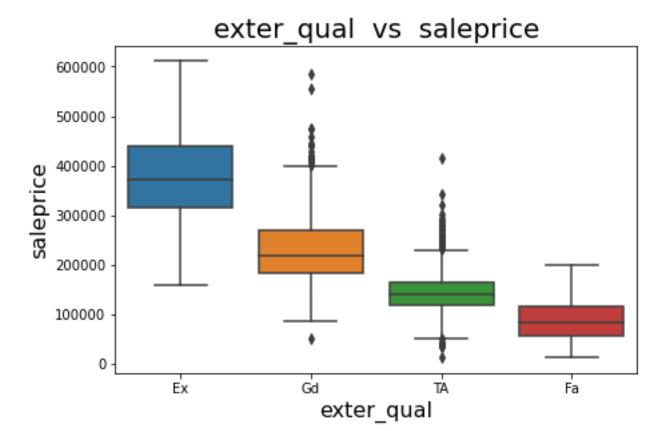
Engineering: mean\_imputed^2

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observati Df Residuals: Df Model: Covariance Ty	ions:		- .east 14 (	OLS Squares		tatistic):	centered):
	coef	std err		t	P> t	[0.025	0.975]
saleprice	0.0002	1.16e-06	183	3.527	0.000	0.000	0.000
Omnibus: Prob(Omnibus) Skew: Kurtosis:	):	θ.	248 000 094 827	Durbin- Jarque- Prob(JB Cond. N	Bera (JB): ):		2.017 685.946 1.12e-149 1.00

#### Notes:

[1] R $^2$  is computed without centering (uncentered) since the model does not contain [2] Standard Errors assume that the covariance matrix of the errors is correctly s



Variable Type: categorical

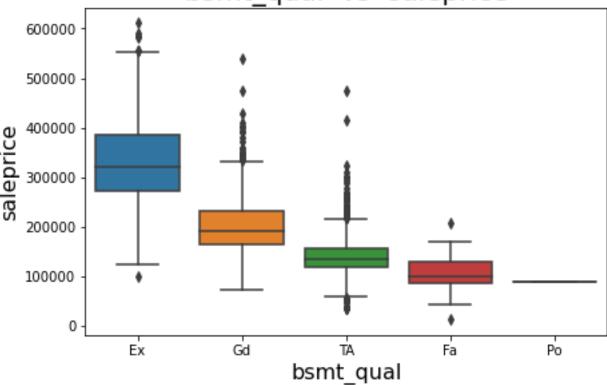
Engineering: mode\_imputed

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observati Df Residuals: Df Model:	ions:	Fri, 14	— 0 t Squar Oct 20 00:39: 20	LS es 22 16 51 50	R-squared (und Adj. R-squared F-statistic: Prob (F-statistic) Log-Likelihood AIC: BIC:	d (uncente stic):	
Covariance Ty		std err	nonrobu		P> t	[0.025	0.9751
saleprice		0.006				0.876	0.898
Omnibus: Prob(Omnibus) Skew: Kurtosis:	):	θ - θ	.000 .815	Jarq Prob	in-Watson: ue-Bera (JB): (JB): . No.		1.851 1262.876 5.89e-275 1.00

- [1]  $R^{2}$  is computed without centering (uncentered) since the model does not contain
- [2] Standard Errors assume that the covariance matrix of the errors is correctly s

### bsmt\_qual vs saleprice



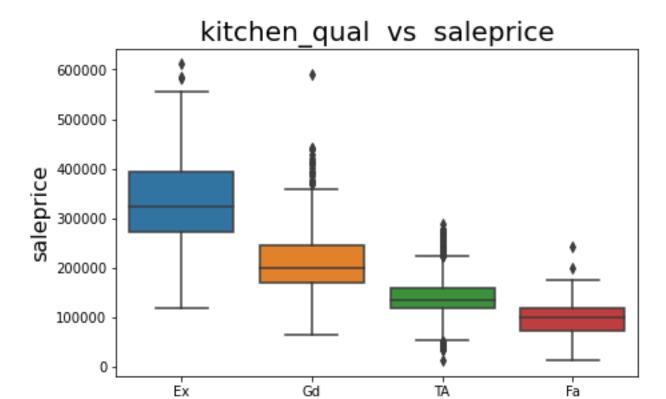
Variable Type: categorical

Engineering: na\_val\_imputed

#### OLS Regression Results

Dep. Variable: Model: Method: Date: Time: No. Observatio Df Residuals: Df Model: Covariance Typ	ons:	_qual_na_va Leas Fri, l <sup>4</sup>	st Squ 4 Oct 00:3	0LS ares 2022 9:17 2051 2050	R-squared (u Adj. R-squar F-statistic: Prob (F-stat Log-Likeliho AIC: BIC:	ed (unceni istic):	
	coef	std err		t	P> t	[0.025	0.975]
saleprice	0.8893	0.006	151	.150	0.000	0.878	0.901
Omnibus: Prob(Omnibus): Skew: Kurtosis:		θ. - θ.	. 233 . 000 . 714 . 575		, ,		1.796 1266.431 9.95e-276 1.00

- [1]  $R^{2}$  is computed without centering (uncentered) since the model does not contain
- [2] Standard Errors assume that the covariance matrix of the errors is correctly s



Variable Type: categorical

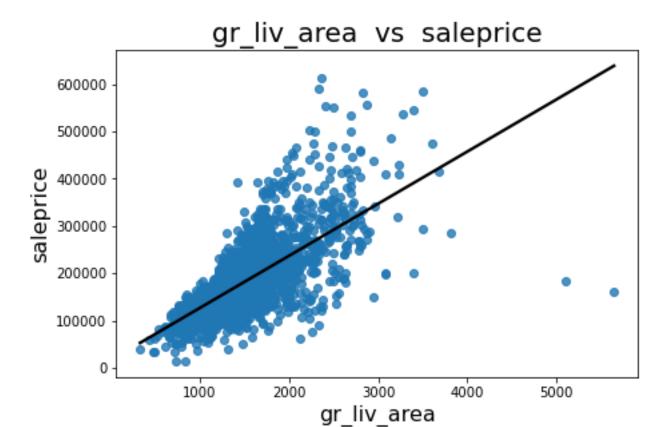
kitchen\_qual

Engineering: mode\_imputed

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observat: Df Residuals: Df Model: Covariance Ty	ions:		- ast Sq 14 Oct 00:	puted 0LS uares 2022 39:17 2051 2050 1 obust		red (uncer : tistic):	
	coef	std err		t	P> t	[0.025	0.975]
saleprice	0.8853	0.006	154	.161	0.000	0.874	0.897
Omnibus: Prob(Omnibus) Skew: Kurtosis:	):	0 -0	. 940 . 000 . 659 . 256		,		1.856 1054.602 9.91e-230 1.00

- [1]  $R^2$  is computed without centering (uncentered) since the model does not contain
- [2] Standard Errors assume that the covariance matrix of the errors is correctly s

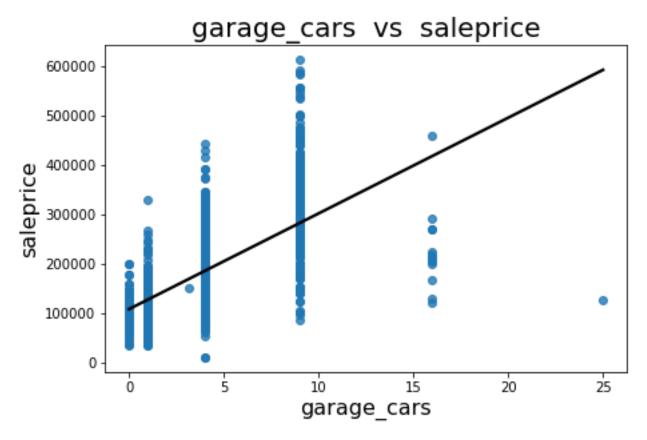


Engineering: mean\_imputed

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observati Df Residuals: Df Model: Covariance Ty	ions:	iv_area_mea Leas Fri, 14	st Squ 4 Oct 2 00:3	0LS ares 2022 9:18 2051 2050	R-squared (u Adj. R-squar F-statistic: Prob (F-stat Log-Likeliho AIC: BIC:	red (uncent :istic):	
	coef	std err		t	P> t	[0.025	0.975]
saleprice	0.0076	5.08e-05	150	.514	0.000	0.008	0.008
Omnibus: Prob(Omnibus) Skew: Kurtosis:	:	0	. 964 . 000 . 303 . 782				1.917 6622.773 0.00 1.00

- [1]  $R^2$  is computed without centering (uncentered) since the model does not contain
- [2] Standard Errors assume that the covariance matrix of the errors is correctly s

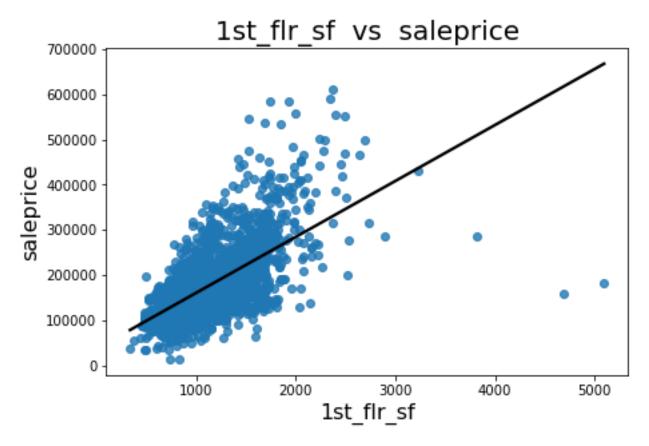


Engineering: mean\_imputed^2

#### OLS Regression Results

Dep. Variate Model: Method: Date: Time: No. Observa Df Residual Df Model: Covariance	ations: ls:		imputed^2 OLS st Squares 4 Oct 2022 00:39:18 2051 2050 1	Adj. R-s F-statis	statistic):	,
	coef	std err	t	P> t	[0.025	0.975]
saleprice	2.098e-05	2.28e-07	91.924	0.000	2.05e-05	2.14e-05
Omnibus: Prob(Omnibu Skew: Kurtosis:	15):	1077.33 0.00 2.13 15.76	33 Prob(Ji	-Bera (JB) B):	:	1.916 15473.080 0.00 1.00

Notes: [1]  $R^2$  is computed without centering (uncentered) since the model does not contair [2] Standard Errors assume that the covariance matrix of the errors is correctly  $\epsilon$ 



Engineering: mean\_imputed

#### OLS Regression Results

Dep. Variable Model: Method: Date: Time: No. Observati Df Residuals: Df Model: Covariance Ty	ons:	Fri, 14	n_imputed OLS t Squares Oct 2022 00:39:19 2051 2050 1		ed (uncente istic):	
	coef	std err	t	P> t	[0.025	0.975]
saleprice	0.0059	4.39e-05	133.906	0.000	0.006	0.006
Omnibus: Prob(Omnibus) Skew: Kurtosis:	:	0	.000 Jaro .557 Prob	pin-Watson: que-Bera (JB): b(JB): I. No.		1.935 9453.482 0.00 1.00

#### Notes:

[1] R $^2$  is computed without centering (uncentered) since the model does not contain [2] Standard Errors assume that the covariance matrix of the errors is correctly s