

Hou\$e Price\$

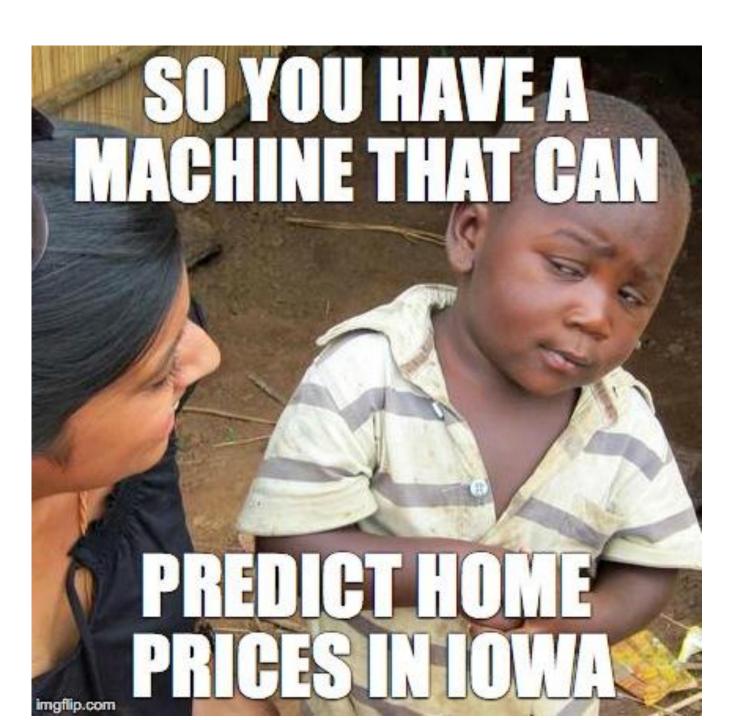
A MACHINE LEARNING PROJECT

Gradient De\$cendant\$

Agenda

Project aims at predicting house prices (residential) in Ames, Iowa, USA based on data set provided by Kaggle between 2006 and 2010.

- Data Exploration
- Data Pre-processing
- Feature Engineering
- Model Training and Evaluation
- Analysis, Conclusions and Considerations

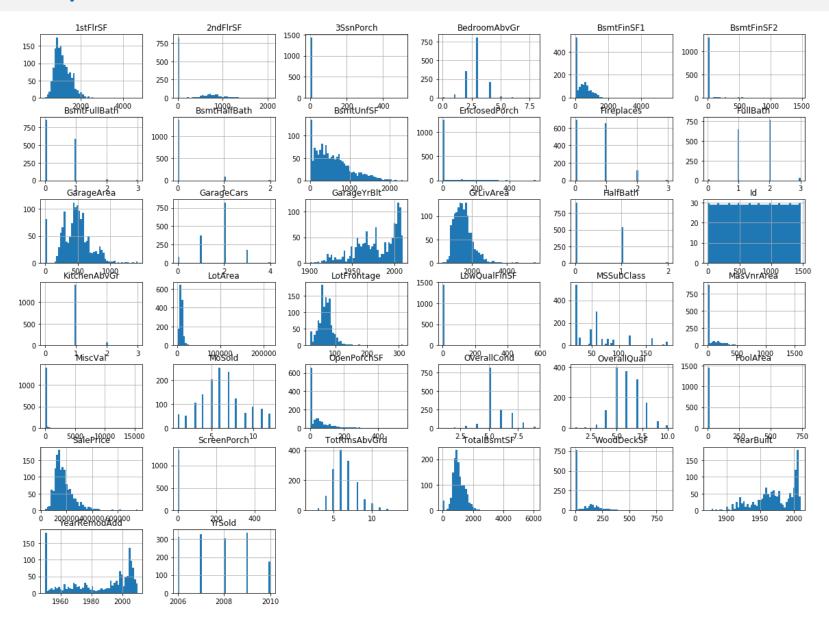


Data Exploration

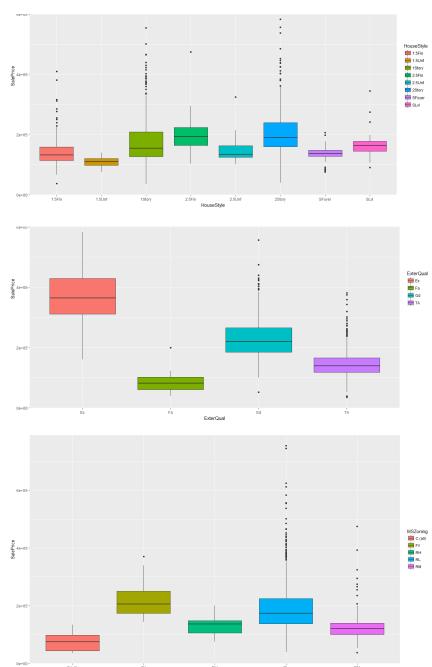
What factors do we believe to influence house prices?

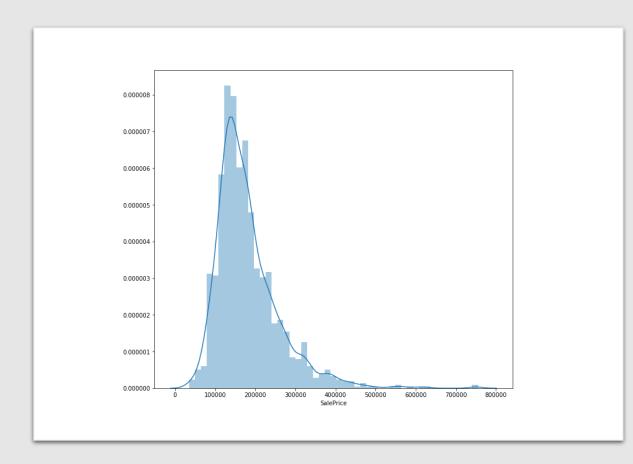
- The dataset contains 1460 observations in the training set and 1459 observations in the test set.
- There are 46 categorical variables including 23 nominal and 23 ordinal ones, and 33 numeric variables in the dataset.
- The training set has the sale price as response while the test set doesn't.

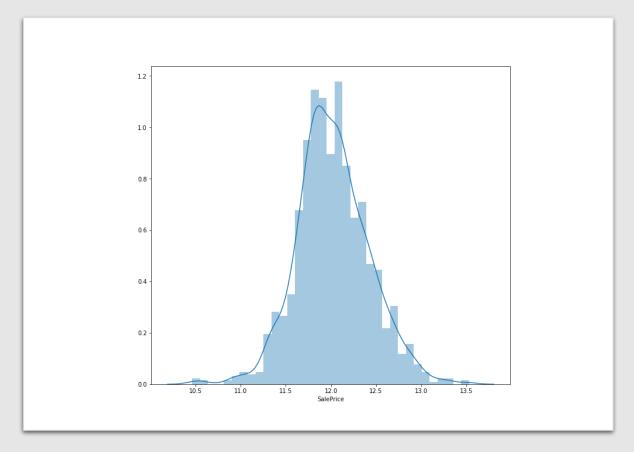
Univariate Analysis





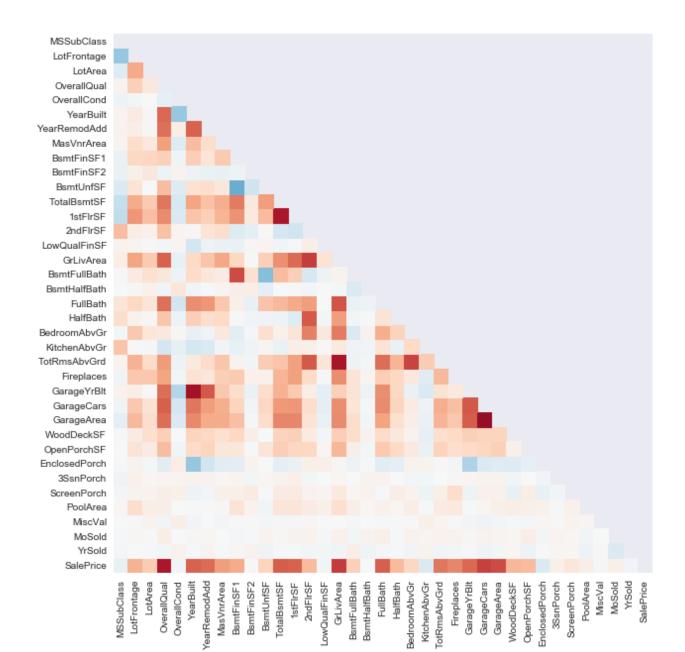






Log(Sale Price)

Correlation



Variables	Correlation
overallqual	0.79
Totalbsmtsf	0.61
1stflrsf	0.61
Grlivearea	0.71
Garagecars	0.64
Garagearea	0.62

0.8

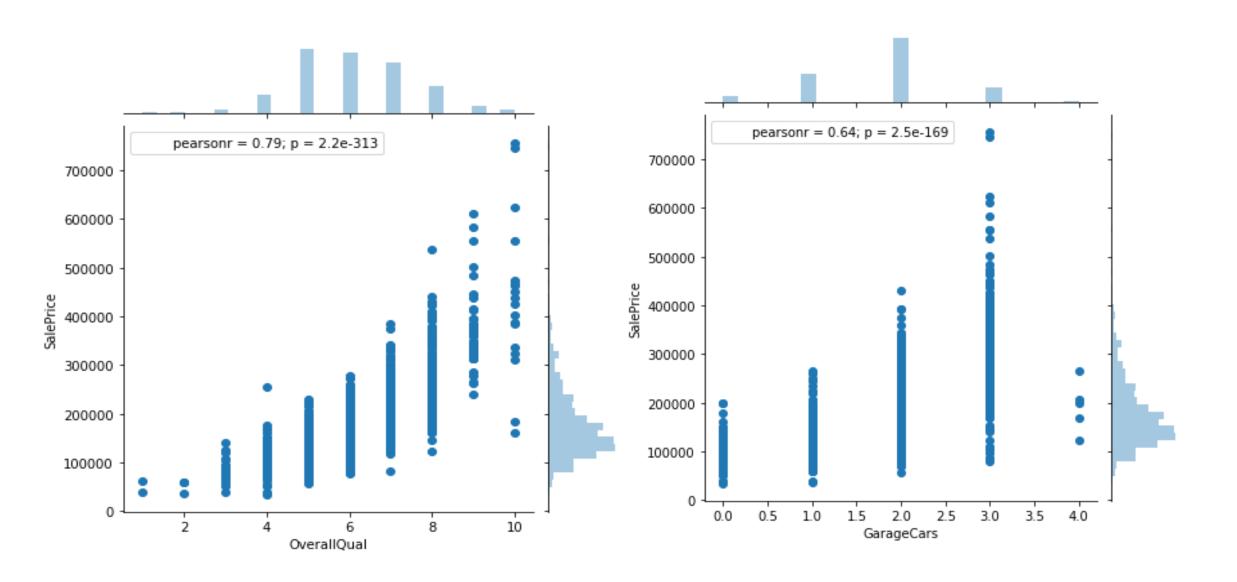
0.4

0.0

-0.4

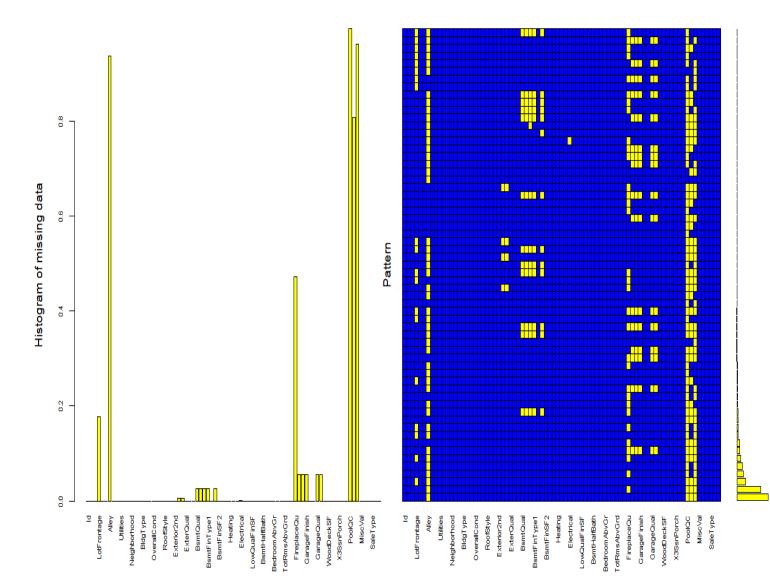
-0.8

Overall Quality & Garage



Data Pre-processing I thought I told you to clean the basement!

Missing Values



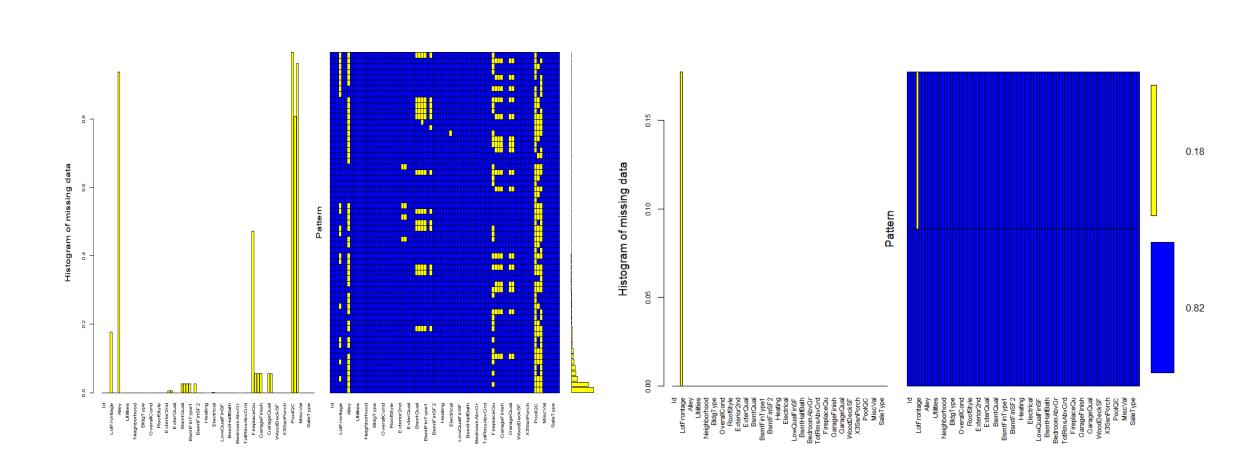
Variable	Missing
PoolQC	1453
MiscFeature	1406
Alley	1369
Fence	1179
FireplaceQu	1179
LotFrontage	259
GargeType	81

- 1. Categorical values to "None"
- 2. Quantitative values to 0

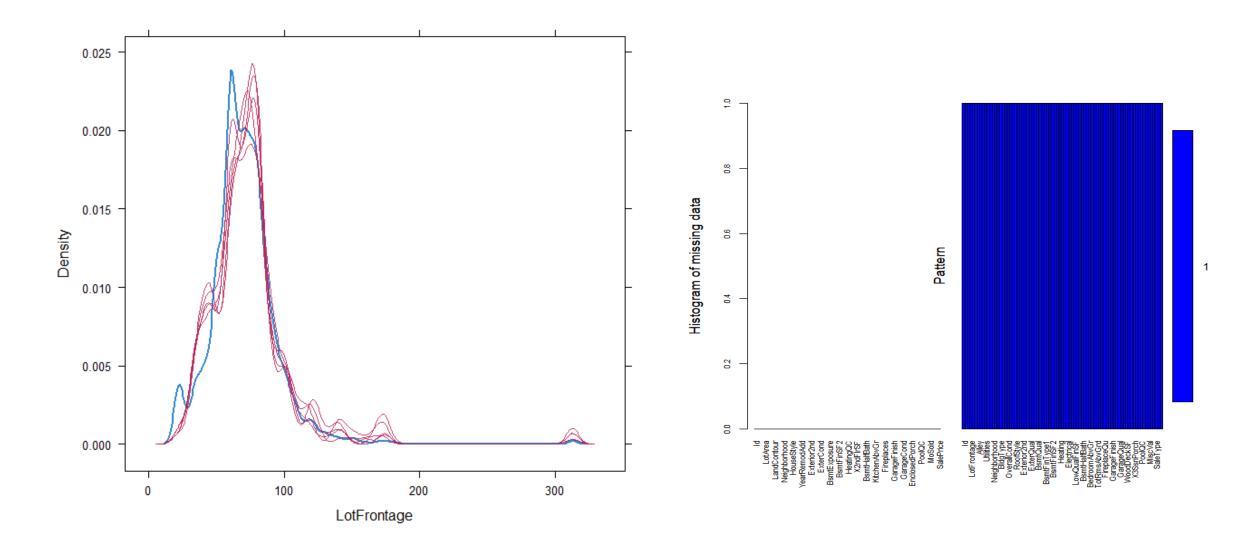
Missing Values

Before

After



Mice Imputations



Pre-Processing

- SalePrice transformation LOG value
- Skewness
 - > Scale all the variables with skewness above 0.75
- New features
 - ➤ IsRemodeled (0 or 1)
 - > QtrSold = MoSold / 3
 - > TotalSF = TotalBsmtSF + GrLivArea + GarageArea
- Sale Condition Normal
- Data correction
 - ➤ GarageYrBlt in Test Data

Pre-Processing

Anova

- > Street was found not to be significant and so dropped from the dataset
- > GarageType, GarageFinish and Garage Condition are more significant in predicting sale price
 - ➤ Interaction between GarageType and Garage Finish had a significant p-value

Correlation plot

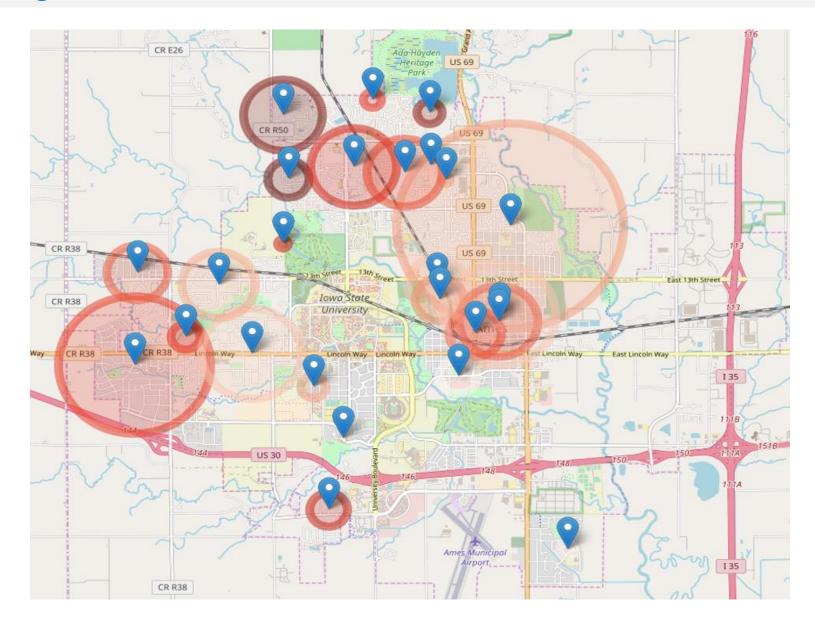
- ➤ Sales price has strong correlation with Overall quality, TotalBsmtArea, x1stFlrSE, GrliviArea, Fullbath, TotalRoom, CarGarage, GarageArea
- > Sale price has a second level correlation with year built, year remodeled, MasVnrArea, GarageYrbuilt, Fireplace
- > SalesPrice negatively correlated with overallCondition, kitchenAbvGrd, enclosedPorch



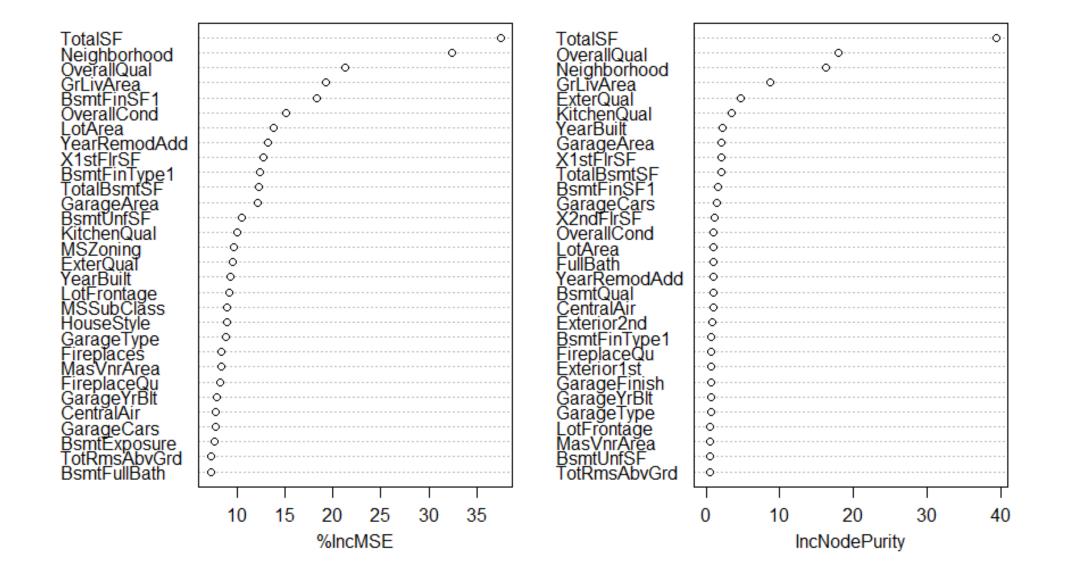
'If you torture the data enough, it will always confess'

- Ronald Coase, British Economist

Ames, IA Neighborhood

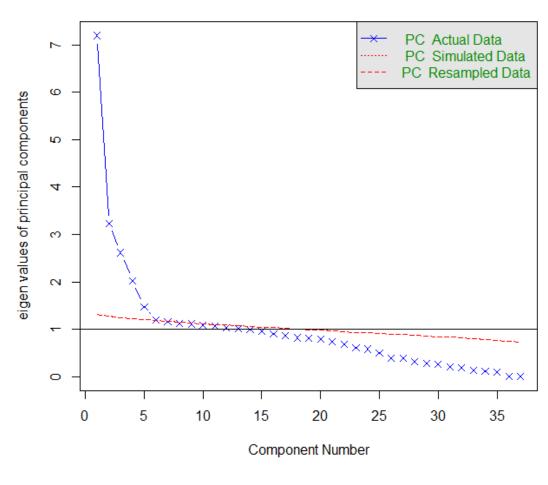


Random Forest For Best Variables



PCA – Component Analysis

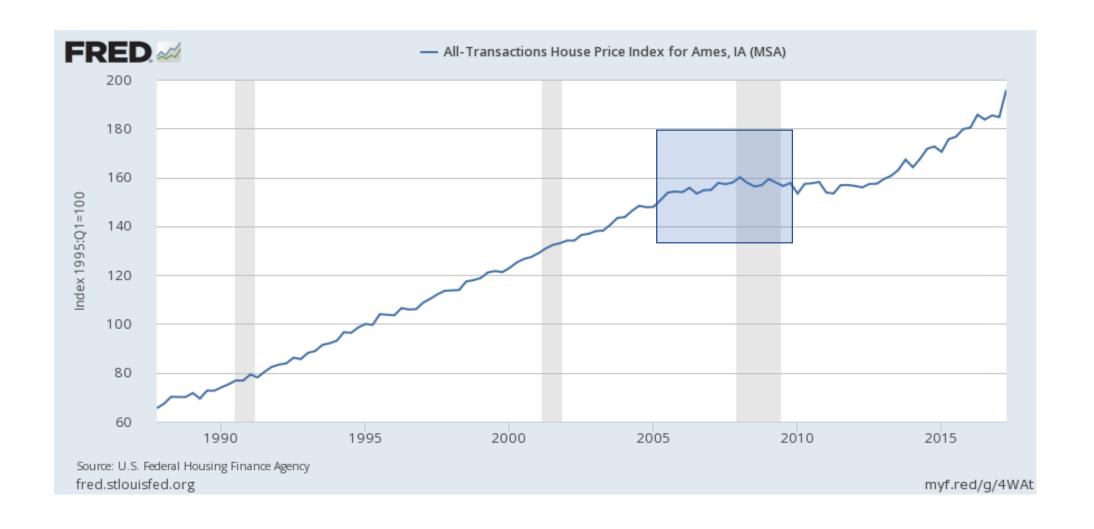
Parallel Analysis Scree Plots



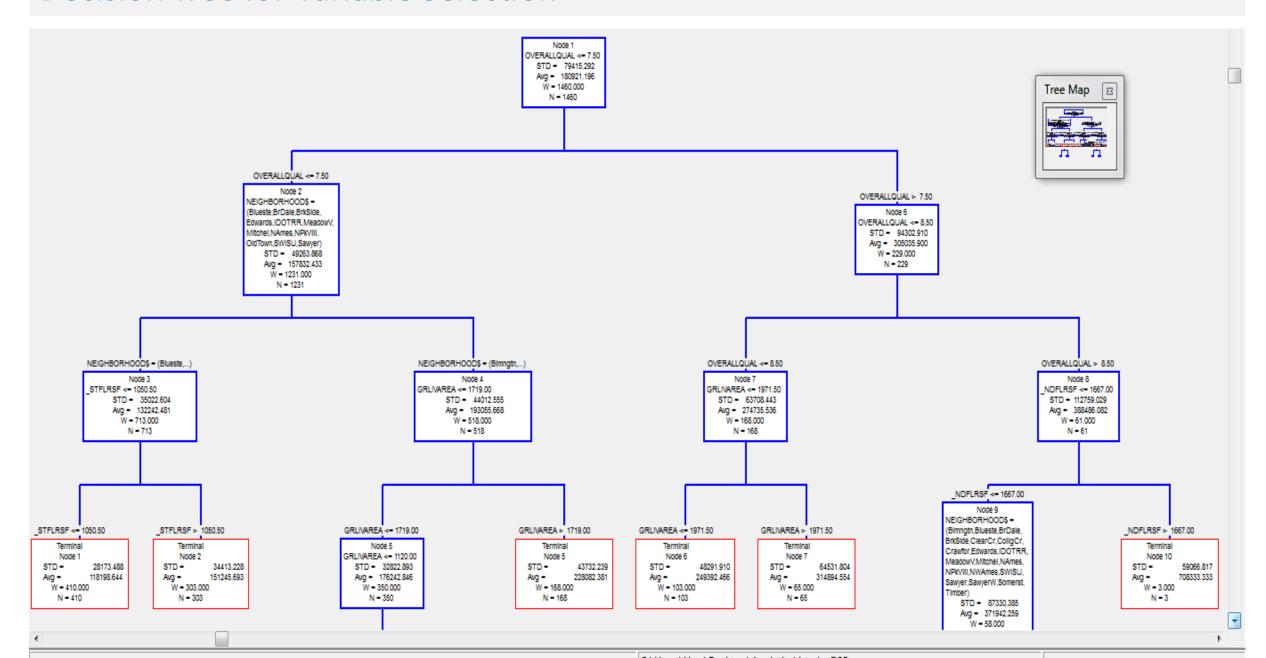
sp\$finalModel\$xNames [1] "PC1" "PC2" "PC3"

```
Principal Components Analysis
Call: principal(r = houseTrain[, numericFeatures[1:37]], nfactors = 5,
    rotate = "none")
Standardized loadings (pattern matrix) based upon correlation matrix
                                             h2
                                                   u2 com
                   0.03
                         0.00 0.01 -0.06 0.0049 0.995 1.5
MSSubClass
                    LotFrontage
                         0.52 -0.18 0.12 0.4945 0.505 2.3
LotArea
                              0.04 0.08 0.3272 0.673 1.7
OverallQual
              0.80 -0.03 -0.17 -0.01 0.14 0.7000 0.300 1.2
OverallCond
              -0.22 0.11 0.20 0.13 0.48 0.3448 0.655 2.1
YearBuilt
              0.64 -0.38 -0.50 0.03 -0.03 0.8004 0.200 2.6
YearRemodAdd
              0.56 -0.17 -0.40
                               0.02 0.19 0.5352 0.465 2.3
MasVnrArea
                               0.07 -0.03 0.2765 0.723 1.1
BsmtFinSF1
              0.38 -0.50 0.37 0.49 -0.18 0.7969 0.203 4.0
BsmtFinSF2
              -0.03 -0.11 0.27 0.14 0.11 0.1192 0.881 2.3
BsmtUnfSF
                   0.20 -0.15 -0.81 0.05 0.8168 0.183 1.5
TotalBsmtSF
                         0.33 -0.25 -0.10 0.7842 0.216 2.4
X1stFlrSF
              X2ndFlrSF
              0.35  0.76 -0.17  0.37  0.12  0.8852  0.115  2.1
LowQualFinSF
                         0.15 -0.03 -0.02 0.0758 0.924 1.9
GrLivArea
                         0.20 0.10 -0.03 0.9077 0.092 1.8
BsmtFullBath
                         0.27 0.52 -0.26 0.7007 0.299 3.4
BsmtHalfBath
                         0.11 0.06 0.29 0.1003 0.900 1.4
FullBath
              0.70 0.27 -0.19 -0.10 -0.18 0.6359 0.364 1.7
HalfBath
                   0.39 -0.26 0.44 0.25 0.5791 0.421 4.2
BedroomAbvGr
                   0.66 0.21 -0.04 -0.08 0.5815 0.418 1.7
KitchenAbvGr
                         0.08 -0.05 -0.70 0.6118 0.388 1.5
TotRmsAbvGrd
                         0.18 0.01 -0.13 0.8227 0.177 2.3
Fireplaces
              0.49 0.04 0.30 0.15 0.21 0.4016 0.598 2.3
GarageYrBlt
              0.65 -0.32 -0.52 -0.01 -0.01 0.7964 0.204 2.4
GarageCars
              0.77 -0.13 -0.17 -0.04 0.01 0.6361 0.364 1.2
GarageArea
              0.75 -0.19 -0.07 -0.05 0.01 0.6096 0.390 1.1
WoodDeckSF
              0.36 -0.12 0.05 0.19 0.07 0.1864 0.814 1.9
OpenPorchSF
              0.39 0.08 -0.02 0.04 0.17 0.1943 0.806 1.5
EnclosedPorch
                         0.29 -0.08 -0.05 0.1798 0.820 3.0
X3SsnPorch
                         0.01 -0.07 0.09 0.0178 0.982 3.1
ScreenPorch
                         0.18 0.10 0.24 0.1079 0.892 2.5
PoolArea
                         0.24 0.12 0.00 0.0917 0.908 2.2
MiscVal
                         0.08 0.03 0.02 0.0103 0.990 2.4
MoSold
                    0.05 -0.01 -0.06 0.08 0.0159 0.984 3.8
YrSold
              -0.04 -0.06 0.01 0.07 -0.07 0.0144 0.986 3.6
```

Case-Shiller Ames Housing Index



Decision Tree for Variable Selection



Model Training and Evaluation

Multiple Linear Regression

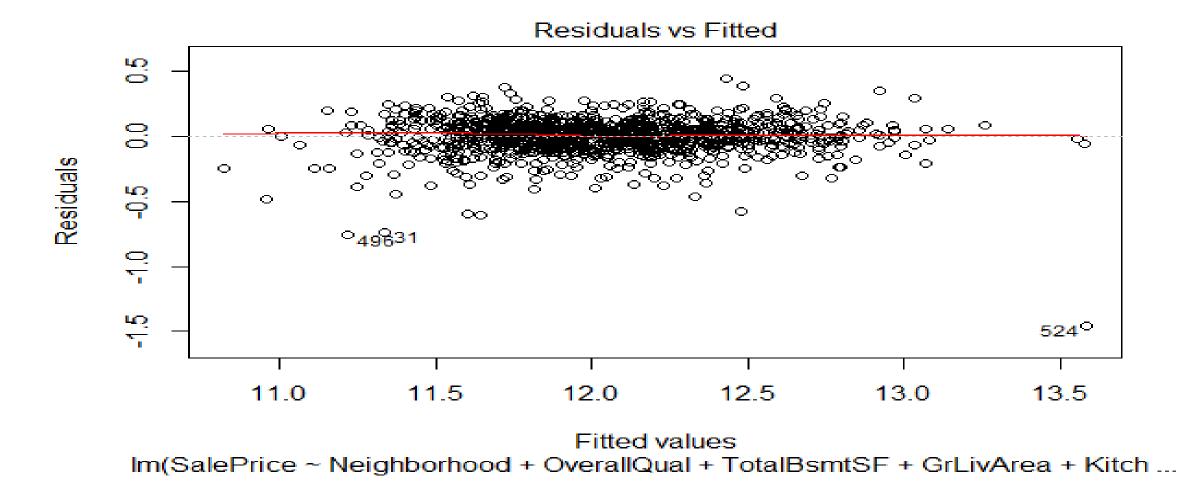
Summary Statistics

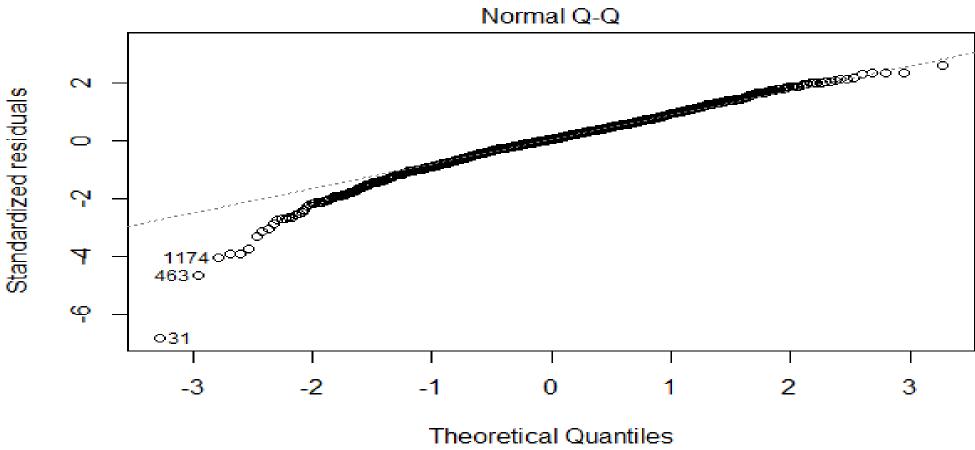
Coefficients:				
	Estimate	Std. Error	t value	
(Intercept)	1.11e+01	6.14e-02	180.06	< 2e-16 ***
NeighborhoodBlueste	-8.99e-02	9.54e-02	-0.94	0.34659
NeighborhoodBrDale	-1.75e-01	6.30e-02	-2.77	0.00568 **
NeighborhoodBrkSide	-5.13e-02	4.66e-02	-1.10	0.27199
NeighborhoodClearCr	8.50e-02	5.19e-02	1.64	0.10203
NeighborhoodCollgCr	4.38e-03	4.26e-02	0.10	0.91818
NeighborhoodCrawfor	1.11e-01	4.65e-02	2.39	0.01727 *
NeighborhoodEdwards	-8.14e-02	4.51e-02	-1.80	0.07153 .
NeighborhoodGilbert	6.99e-02	4.55e-02	1.54	0.12424
NeighborhoodIDOTRR	-1.92e-01	5.12e-02	-3.76	0.00018 ***
NeighborhoodMeadowV	-1.94e-01	5.75e-02	-3.37	0.00079 ***
NeighborhoodMitchel	-4.10e-02	4.69e-02	-0.87	0.38250
NeighborhoodNAmes	-3.85e-02	4.34e-02	-0.89	0.37490
NeighborhoodNoRidge	5.02e-02	4.80e-02	1.05	0.29552
	-9.07e-02	6.23e-02	-1.46	0.14587
NeighborhoodNridgHt	2.13e-02	4.68e-02	0.46	0.64841
NeighborhoodNWAmes	-2.48e-03	4.50e-02	-0.06	0.95600
NeighborhoodOldTown	-1.56e-01	4.49e-02	-3.47	0.00054 ***
NeighborhoodSawyer	-3.44e-02	4.55e-02	-0.76	0.45004
NeighborhoodSawyerW		4.56e-02	-0.79	0.43262
NeighborhoodSomerst	3.48e-02	4.54e-02	0.77	0.44347
NeighborhoodStoneBr	1.26e-02	5.52e-02	0.23	0.81973
NeighborhoodSWISU	-1.11e-01	5.06e-02	-2.20	0.02823 *
NeighborhoodTimber	-2.55e-02	5.21e-02	-0.49	0.62547
NeighborhoodVeenker	1.27e-01	5.80e-02	2.20	0.02838 *
TotalBsmtSF	1.08e-04	2.33e-05	4.66	3.7e-06 ***
GrLivArea	2.41e-04	1.21e-05	19.97	< 2e-16 ***
KitchenQualFa	-2.37e-01	3.43e-02	-6.92	8.8e-12 ***
KitchenQualGd	-7.35e-02	2.16e-02	-3.41	0.00068 ***
KitchenQualTA	-1.45e-01	2.36e-02	-6.13	1.3e-09 ***
GarageArea	2.37e-04	2.63e-05	9.01	< 2e-16 ***
BsmtFinSF1	6.66e-05	1.63e-05	4.08	4.8e-05 ***
BsmtFinType1BLQ	-3.35e-02	1.56e-02	-2.14	0.03241 *
BsmtFinType1GLQ	-1.79e-02	1.41e-02	-1.27	0.20607
BsmtFinType1LwQ	-5.24e-02	2.03e-02	-2.58	0.00997 **
BsmtFinType1NB	-9.68e-02	3.54e-02	-2.73	0.00637 **
BsmtFinType1Rec	-5.75e-02	1.63e-02	-3.52	0.00037
BsmtFinType1Unf	-7.64e-02	1.67e-02	-4.57	5.4e-06 ***
OverallQual	8.41e-02	5.16e-03	16.30	< 2e-16 ***
LotArea	2.31e-06			
X1stFlrSF	-1.25e-05			
	1.230 03	2. 100 03	0.31	0.01507
	***' 0 001	'**' 0 01 '	* ' 0 05	'.' 0.1 ' ' 1
J.g.,,,,,	3.001	0.01	0.03	1
ocidual standard o	mon. 0 110	0 00 016 40	mans of	fraadom
Residual standard en				
Multiple R-squared:	0.896,	Adjusted	κ-square	ed: 0.892

F-statistic: 198 on 40 and 916 DF, p-value: <2e-16

Assumptions

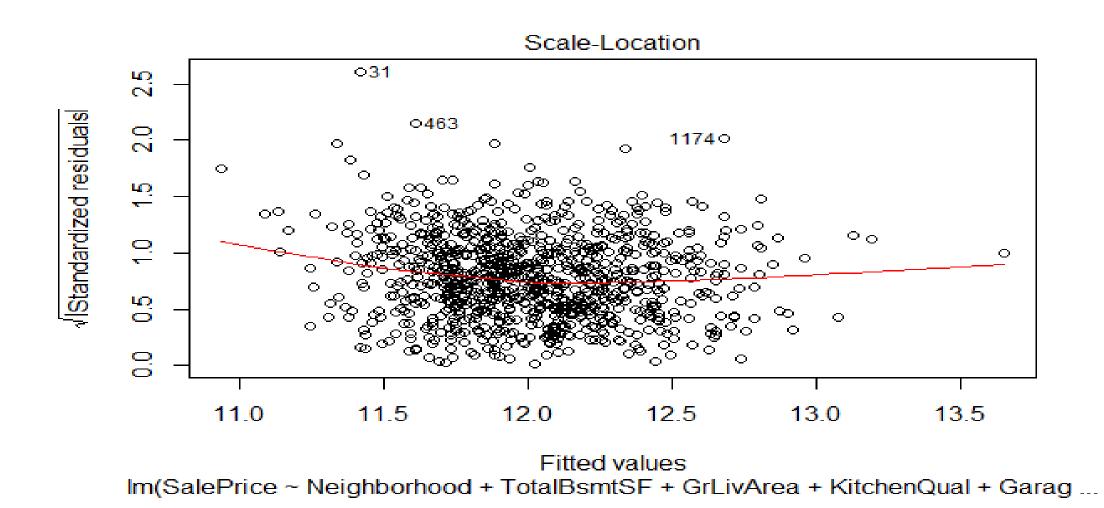
Constant Variance



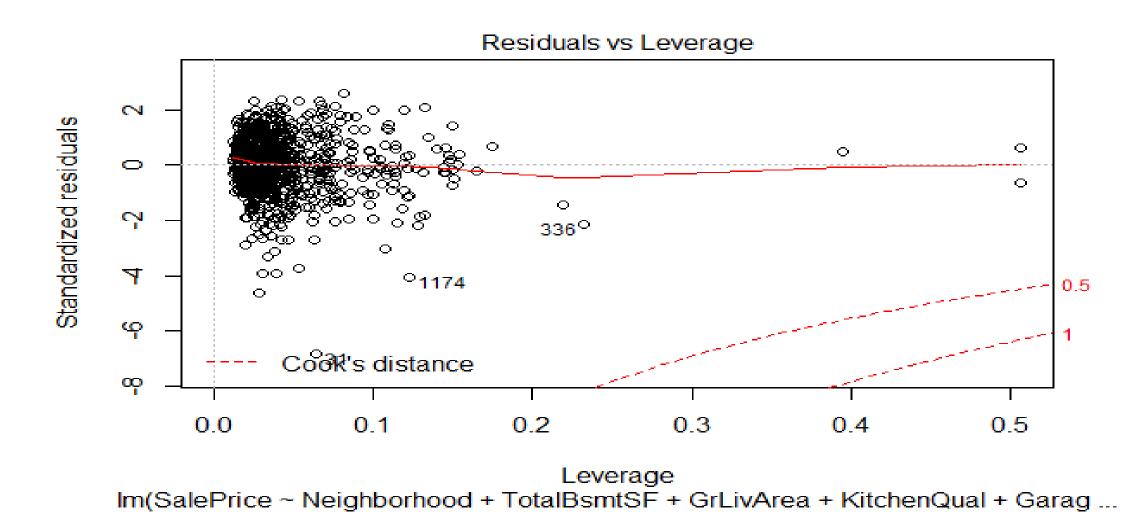


Im(SalePrice ~ Neighborhood + TotalBsmtSF + GrLivArea + KitchenQual + Garag ...

Independent Error



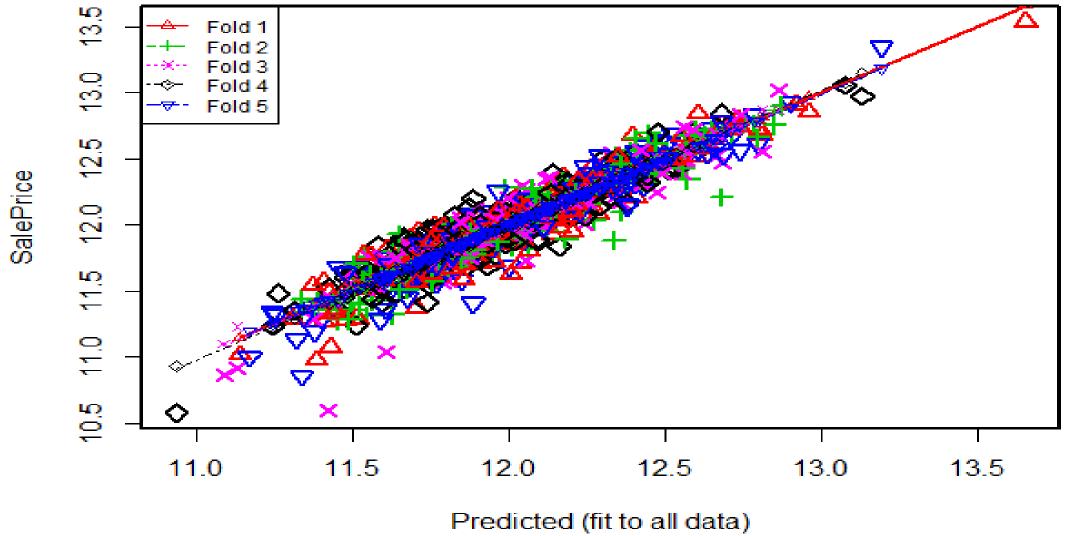
Residuals and Leverage



Cross Validation

Predicted vs Actuals

Small symbols show cross-validation predicted values



Analysis, Conclusions and Considerations

Extracting Additional Value from the Models

- Key Stakeholders: individuals (home seekers, investors), real estate developers, asset managers, realtors, contractors, creditors/banks
- Discovering investment opportunities by comparing fixed features vs changeable features
- Considering macro/micro economic factors
- Quantifying the opportunity and deriving actionable insights and conclusions

Statistically supported conclusions

1. Go Big or Go Home (pun intended)

- Total Square Footage(using basement + living + garage space) average (mean) of homes in Ames cost 68 sq/ft.
- 73% of sales price is implied by attributes related to size (R-sq)
- Sorry Guys, Size Matters!

2. Be Selective but not obsessive

- 6 variables including Size, Age (year built and year remodeled), Location and Fireplaces account for 81% for price.
- 12 Variables that include (Lot Area, Year Built, YearRemodAdd, TotalBsmtSf, BsmtFinSFOne, GrLivArea, BedroomAbvGr, KitchenAbvGr, Fireplaces, Garage Area, Good Neighborhood, Troubled Neighborhood) tells 88% of the story (based on R-sq).

Flip or Flop – Investment opportunities by renovating?

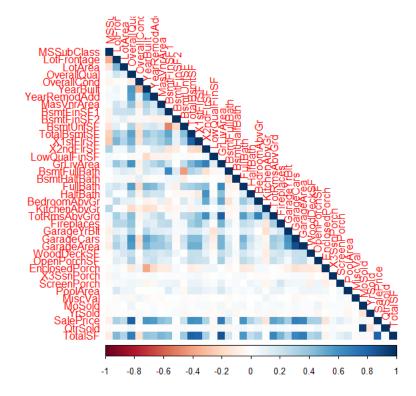
- If 88% of price is accounted for by fixed features, only 12% (or \$19k of house value) remains to other features which limits upside potential to add value via renovation and repairs using "Quality" (ie kitchen quality) and "Condition" (ie garage condition) metrics, including roof, exterior features, heating, electric components, etc.
- If the goal is to add value via renovation, the most promising upside will come from:
 - 1. Adding a garage if land and zoning eligibility permits (worth approximately \$6k per car)
 - 2. Complete and unfinished basement (amounts to \$35/sq. ft.)
 - 3. Remodel/Upgrade the kitchen.
- Profit model relies on ability to source low cost contractors.
- Analysis suggests that the most prudent investment strategy would be investing in underpriced listings based on the criteria in the linear model vs "flipping" houses by renovating features and making improvements.

Thank You!

Appendix-1

File	Edit	Format	View	He
Rsm	tFir	nSF1		1

	HUI.	Ӱ́_Т					· ·	VICW II				
Df Sum	Sq Me	an Sq	F value	Pr(>F)			BsmtFinSF1	1		1.601	147.692 < 2e-1	5 ***
Id	1	0.07	0.075	6.913	0.008665	**	BsmtFinType2	6	0.27	0.045	4.116 0.00042	
MSSubClass	1	1.27	1.267	116.842	< 2e-16	***	BsmtFinSF2	1		0.156	14.381 0.00015	
MSZoning	4	39.87	9.967	919.343	< 2e-16	***	BsmtUnfSF	1		4.487	413.862 < 2e-16	
LotFrontage	1	14.94	14.936	1377.695	< 2e-16	***	Heating	5	0.60	0.119	10.996 2.27e-10	
LotArea	1	2.71			< 2e-16		HeatingQC	4	0.27	0.068	6.271 5.40e-0	
Street	1	0.55			1.48e-12		CentralAir	1	0.43	0.433	39.919 3.72e-10	
Alley	2	1.12			< 2e-16		Electrical	5	0.21	0.042	3.837 0.00187	7 **
LotShape	3	6.75			< 2e-16		X1stFlrSF	1	2.71	2.708	249.784 < 2e-16	5 ***
LandContour	3	3.08		94.771			X2ndFlrSF	1	2.24	2.240	206.651 < 2e-16	5 ***
Utilities	1	0.17			9.86e-05		LowQualFinSF	1	0.05	0.049	4.550 0.03311	3 *
LotConfig	4	1.04		23.945			BsmtFullBath	1	0.13	0.127	11.750 0.000629	9 ***
LandSlope	2	1.75		80.769			BsmtHalfBath	1	0.00	0.001	0.075 0.78436	9
Neighborhood	24	75.42					FullBath	1	0.00	0.000	0.042 0.83679	4
Condition1	8	1.66		19.099			HalfBath	1	0.13	0.132	12.173 0.00050	2 ***
Condition2	7	1.21		15.912			BedroomAbvGr	1	0.00	0.002	0.198 0.65618	1
BldgType	4	3.97		91.565			KitchenAbvGr	1	0.04	0.041	3.805 0.05133	5.
HouseStyle	7	3.28		43.229			KitchenQual	3	0.18	0.061	5.644 0.000769	9 ***
OverallOual	1	28.30		2610.116			TotRmsAbvGrd	1	0.03	0.030	2.786 0.095348	
OverallCond	1	1.29		118.538			Functional	6	0.68	0.114	10.511 2.16e-1	1 ***
YearBuilt	1	2.62		242.043			Fireplaces	1	0.24	0.235	21.711 3.52e-0	5 ***
YearRemodAdd	1	0.77		71.458			FireplaceQu	5	0.09	0.017	1.595 0.158518	
3							GarageType	6	0.70	0.117	10.824 9.33e-1	2 ***
RoofStyle	5 7	0.54			1.93e-09		GarageYrBlt	1		0.105	9.656 0.00193	
RoofMatl		1.44		18.985			GarageFinish	2	0.03	0.013	1.174 0.30940	
Exterior1st	14	2.03			< 2e-16		GarageCars	1	0.36	0.357	32.945 1.20e-0	
Exterior2nd	14	0.52			1.75e-05		GarageArea	1		0.196	18.114 2.24e-0	
MasVnrType	3	0.79			2.47e-15		GarageQual	4		0.035	3.211 0.012364	
MasVnrArea	1	0.49			3.16e-11		GarageCond	4	0.14	0.035	3.191 0.01279	
ExterQual	3	0.32			2.17e-06		PavedDrive	2		0.006	0.579 0.56040	
ExterCond	4	0.31			1.07e-05		WoodDeckSF	1	0.06	0.059	5.481 0.01938	
Foundation	5	0.88			1.98e-15		OpenPorchSF	1		0.003	0.260 0.61054	
BsmtQual	4	0.90		20.775		***	EnclosedPorch		0.03	0.030	2.776 0.095958	
BsmtCond	3	0.05			0.206600	at at at	X3SsnPorch	. 1		0.016	1.481 0.22382	
BsmtExposure	4	0.92		21.268			ScreenPorch	1		0.187	17.217 3.57e-0	
BsmtFinType1	5	1.35		24.889			PoolArea	1		0.034	3.112 0.07795	
BsmtFinSF1	1	1.60			< 2e-16		PoolQC	3	0.10	0.033	3.064 0.027234	
BsmtFinType2	6	0.27			0.000425		Fence	4	0.10	0.025	2.318 0.05531	
BsmtFinSF2	1	0.16			0.000157		MiscFeature	4	0.01	0.002	0.200 0.93831	
BsmtUnfSF	1	4.49			< 2e-16		MiscVal	1	0.00	0.002	0.044 0.833119	
Heating	5	0.60			2.27e-10		MoSold	1	0.00	0.000	0.022 0.88113	
HeatingQC	4	0.27			5.40e-05		YrSold	1		0.017	1.568 0.21069	
CentralAir	1	0.43			3.72e-10		SaleType	8	0.32	0.040	3.675 0.00030	
Electrical	5	0.21			0.001877		SaleCondition	_	0.27	0.055	5.050 0.00013	
X1stFlrSF	1	2.71			< 2e-16		IsRemodeled	1	0.01	0.006	0.566 0.451824	
X2ndFlrSF	1	2.24			< 2e-16		OtrSold	1	0.00	0.001	0.063 0.80186	
LowQualFinSF	1	0.05			0.033113		Residuals	1204		0.001	0.000 0.00100	•
BsmtFullBath	1	0.13			0.000629	***	VC2TANGT2	1204	13.03	0.011		
BsmtHalfBath	1	0.00			0.784360		Signif. codes	· a	(***) A	001 (**)	0.01 (*, 0.05 (, a 1
FullBath	1	0.00	0.000	0.042	0.836794		Jighin. Codes	,. Đ	٥.	001	0.01	. 0.1
							<					



	Df	Sum Sq	Mean Sq	F value	Pr(>F)
GarageType	6	77.90	12.984	144.375	< 2e-16 ***
GarageFinish	2	20.51	10.254	114.023	< 2e-16 ***
GarageQual	4	1.32	0.330	3.669	0.00558 **
GarageCond	4	2.24	0.561	6.239	5.63e-05 ***
GarageType:GarageFinish	8	1.81	0.226	2.513	0.01032 *
GarageType:GarageQual	6	0.50	0.084	0.934	0.46968
GarageFinish:GarageQual	4	0.49	0.123	1.363	0.24461
GarageType:GarageCond	3	0.25	0.085	0.941	0.42006
GarageFinish:GarageCond	1	0.01	0.006	0.066	0.79777
GarageQual:GarageCond	2	0.22	0.109	1.215	0.29712
GarageType:GarageQual:GarageCond	1	0.02	0.018	0.200	0.65508
Residuals	1418	127.52	0.090		

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Appendix-2

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MSZOning Street Alley LotShape LandContour Utilities
9.909275e-18 1.594118e-03 1.420779e-02 4.495140e-01 3.127842e-08 1.939245e-02
LotConfig LandSlope Neighborhood Condition1 Condition2 BldgType
1.412358e-01 1.668988e-01 5.576778e-43 7.515057e-01 7.173339e-01 3.332217e-23
HouseStyle RoofStyle RoofMatl Exterior1st Exterior2nd MasVnrType
1.626511e-06 6.468740e-08 8.029727e-03 2.028676e-28 5.394584e-24 9.246429e-31
ExterQual ExterCond Foundation BsmtQual BsmtCond BsmtExposure
3.071252e-46 1.813475e-02 2.161004e-30 9.358370e-55 1.223155e-07 2.937449e-11
BsmtFinType1 BsmtFinType2 Heating HeatingQC CentralAir Electrical
1.909141e-15 5.928728e-04 9.990498e-01 8.132604e-22 2.620041e-04 3.363524e-23
KitchenQual Functional FireplaceQu GarageType GarageFinish GarageQual
1.445136e-36 2.120077e-01 3.707982e-16 8.370676e-22 1.116818e-27 3.817652e-07
GarageCond PavedDrive PoolQC Fence MiscFeature SaleType
1.293424e-07 5.899112e-06 5.929498e-09 2.227592e-04 9.927046e-01 2.998978e-321 IsRemodeled 9.788271e-01
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