

ANALYZING THE AMES HOUSING MARKET

Problem Statement

Create and curate a set of synergistic features that gives insight in how to effectively increase a house's value in Ames, Iowa

Linear Regression Modeling

Optimize for R^2 Score, but watch squared errors to find best fitting variables

Throw out anything that negatively impacts score no matter traditional advice

One Percent = Thousands of dollars

Method One

Bare Bones

No data transformation

Only use features ready for modeling in
original data

Look at multiple modeling methods
(OLS, Ridge, LASSO)

Method One —————> Method Two

Bare Bones

Kitchen Sink

No data transformation

Transform Sale Price to normal distribution
and clean up outliers

Only use features ready for modeling in
original data

Create as many relevant features possible
and use all with relevant correlation

Look at multiple modeling methods
(OLS, Ridge, LASSO)

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Method One → Method Two → Method ∞

Bare Bones

Kitchen Sink

Iterate, Iterate, Iterate

No data transformation

Transform Sale Price to normal distribution
and clean up outliers

Compare untouched vs transformed
and clean up outliers

Only use features ready for modeling in
original data

Create as many relevant features possible
and use all with relevant correlation

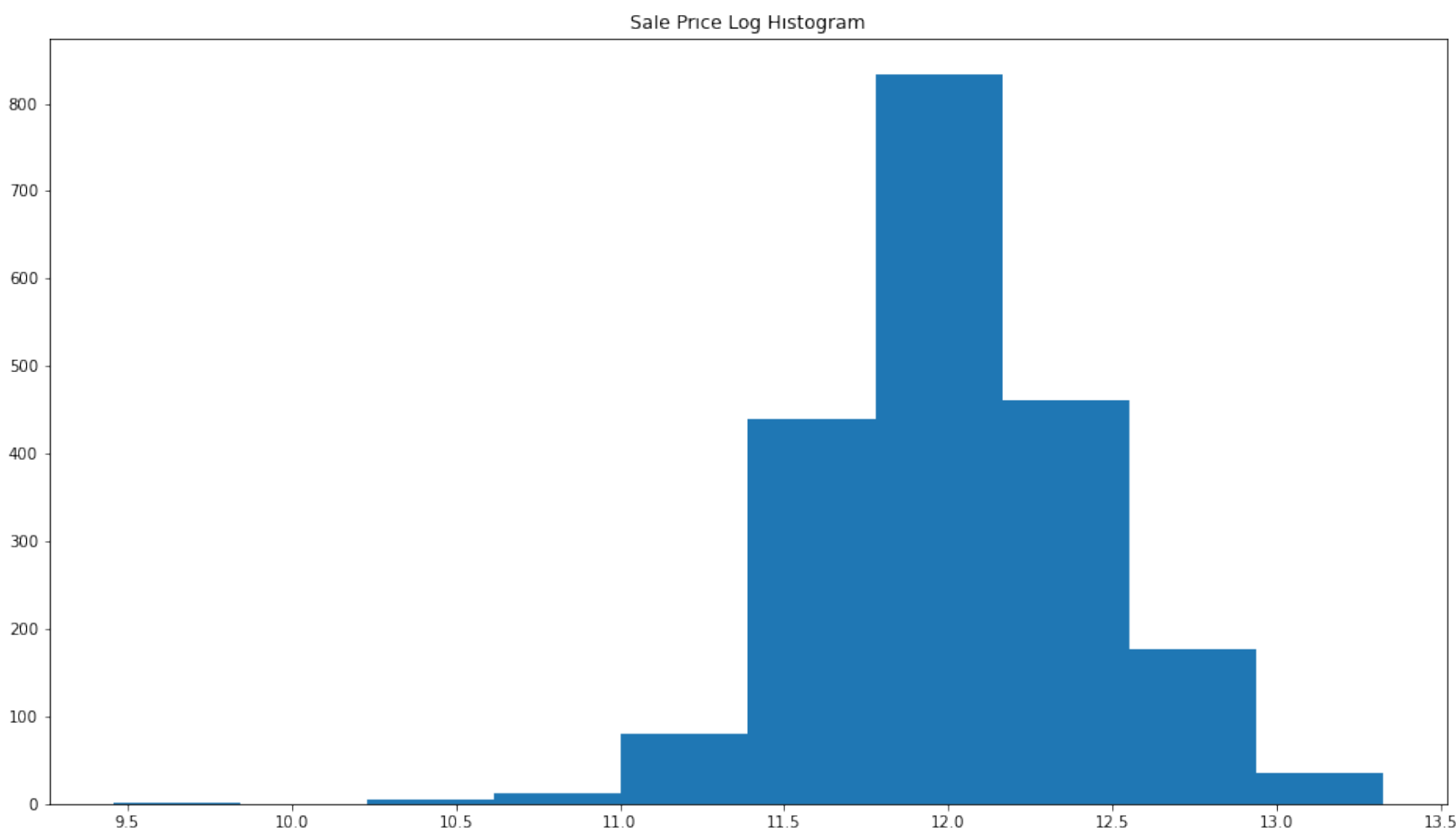
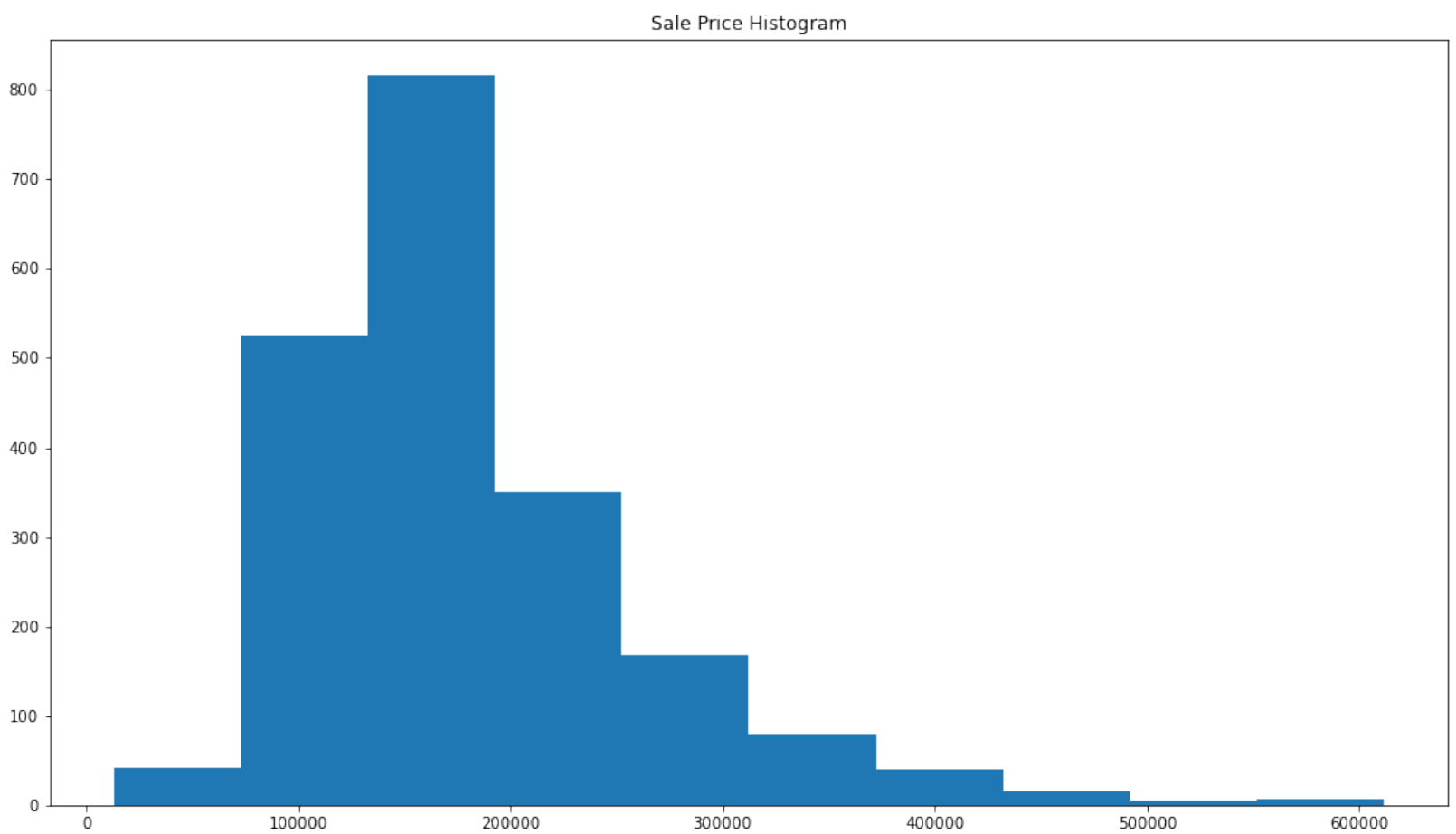
Create a feature list that optimizes R^2
Score

Look at multiple modeling methods
(OLS, Ridge, LASSO)

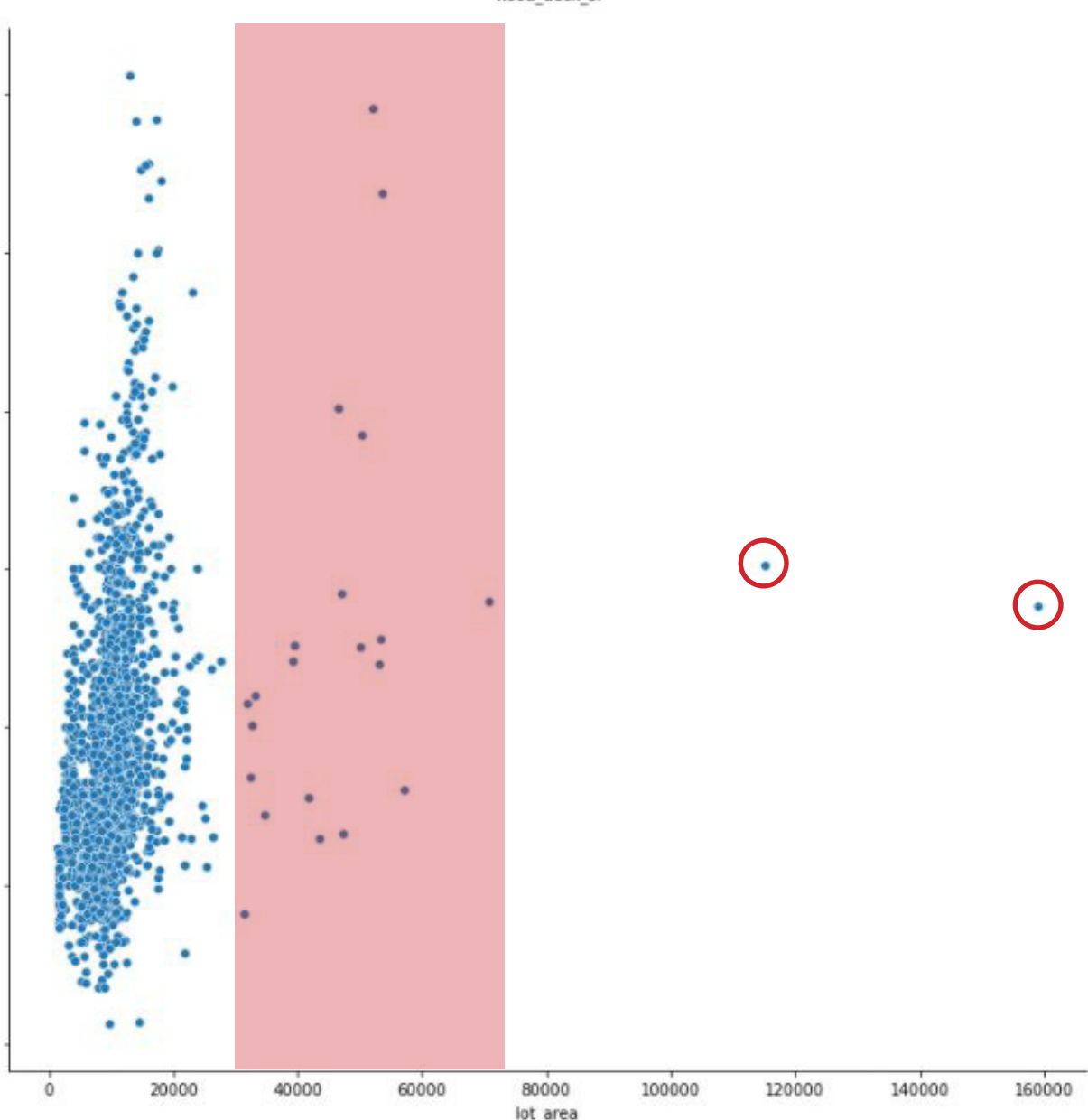
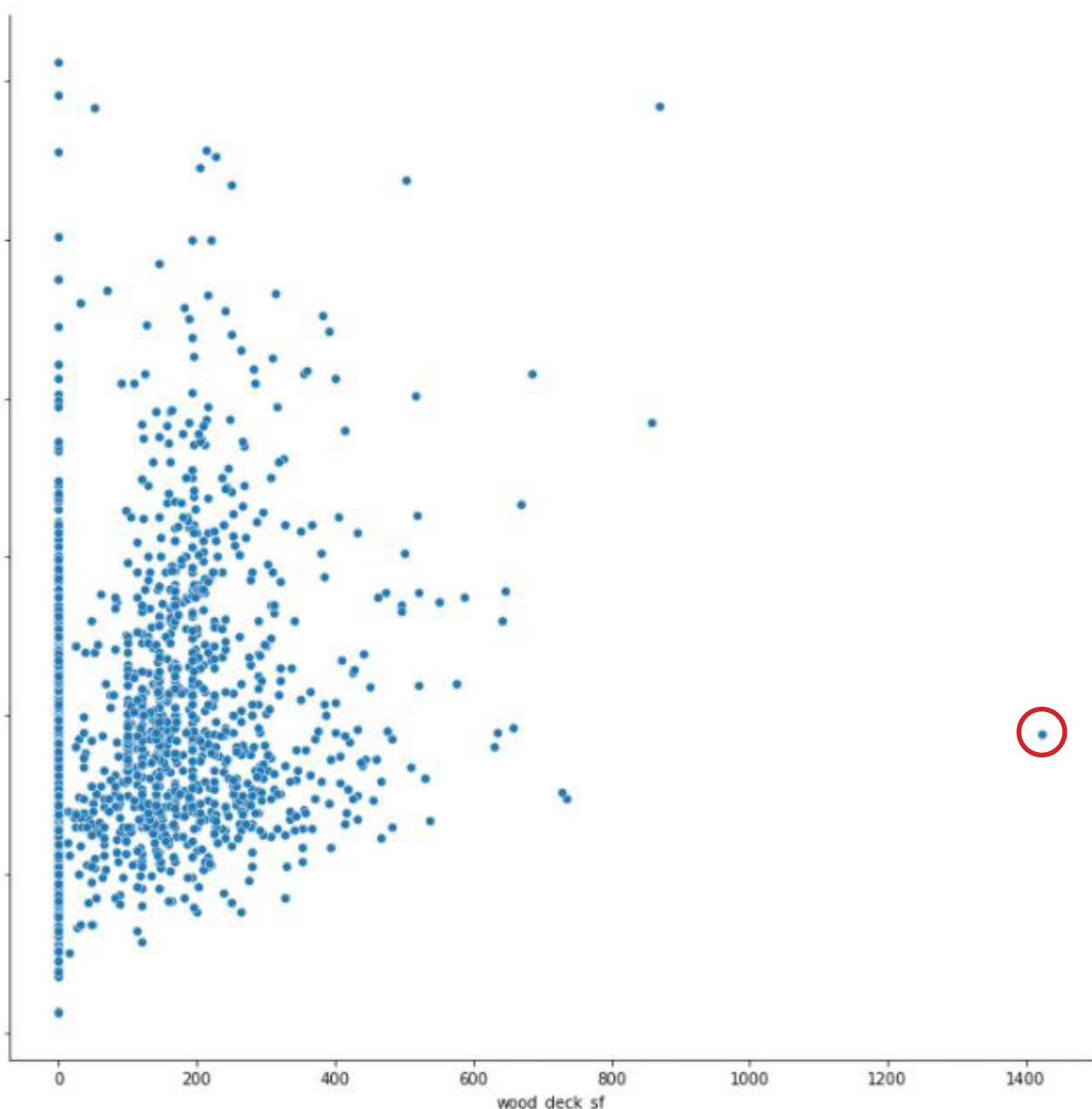
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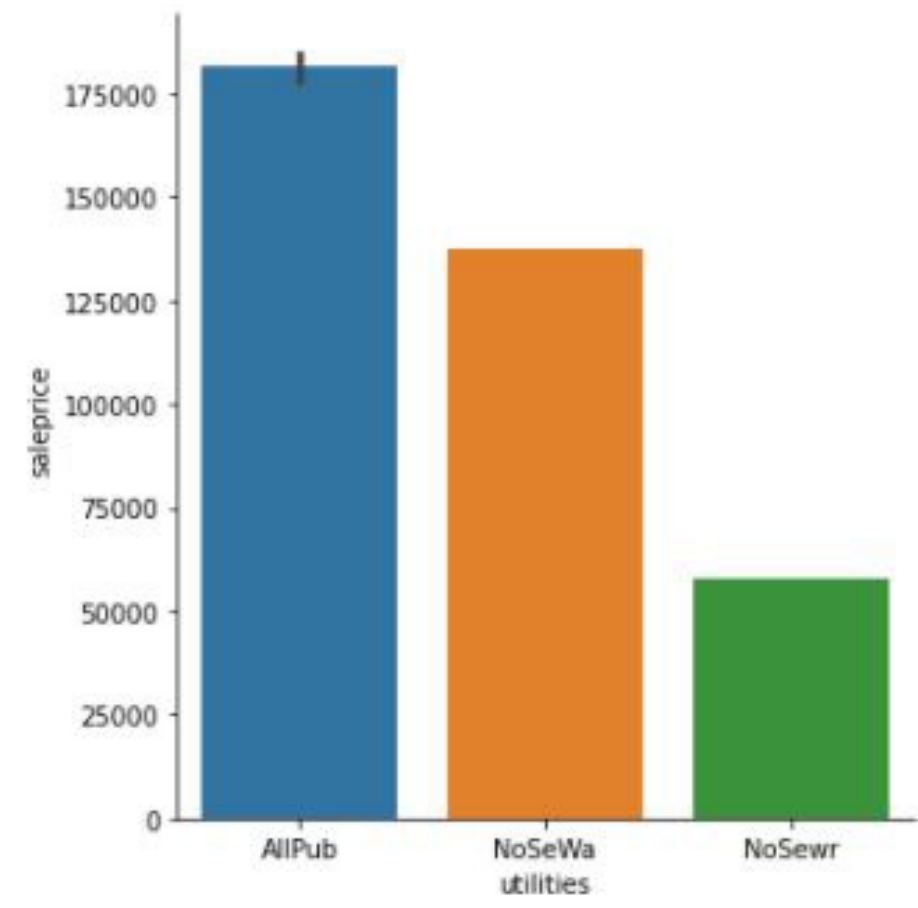
Log Transform



Scrub Outliers

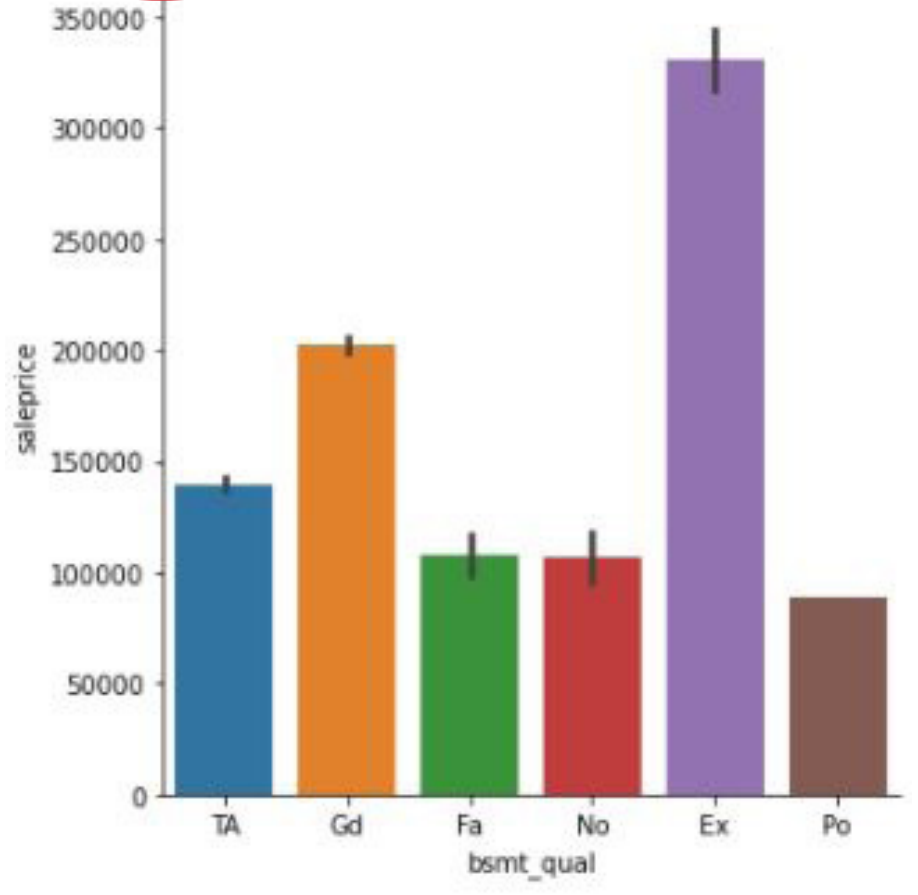


New Features



```
df_train['utilities'].value_counts()
```

AllPub	2047
NoSewr	1
NoSeWa	1



```
df_train['bsmt_qual'].value_counts()
```

TA	887
Gd	864
Ex	182
Fa	60
No	55
Po	1

Feature Methodology

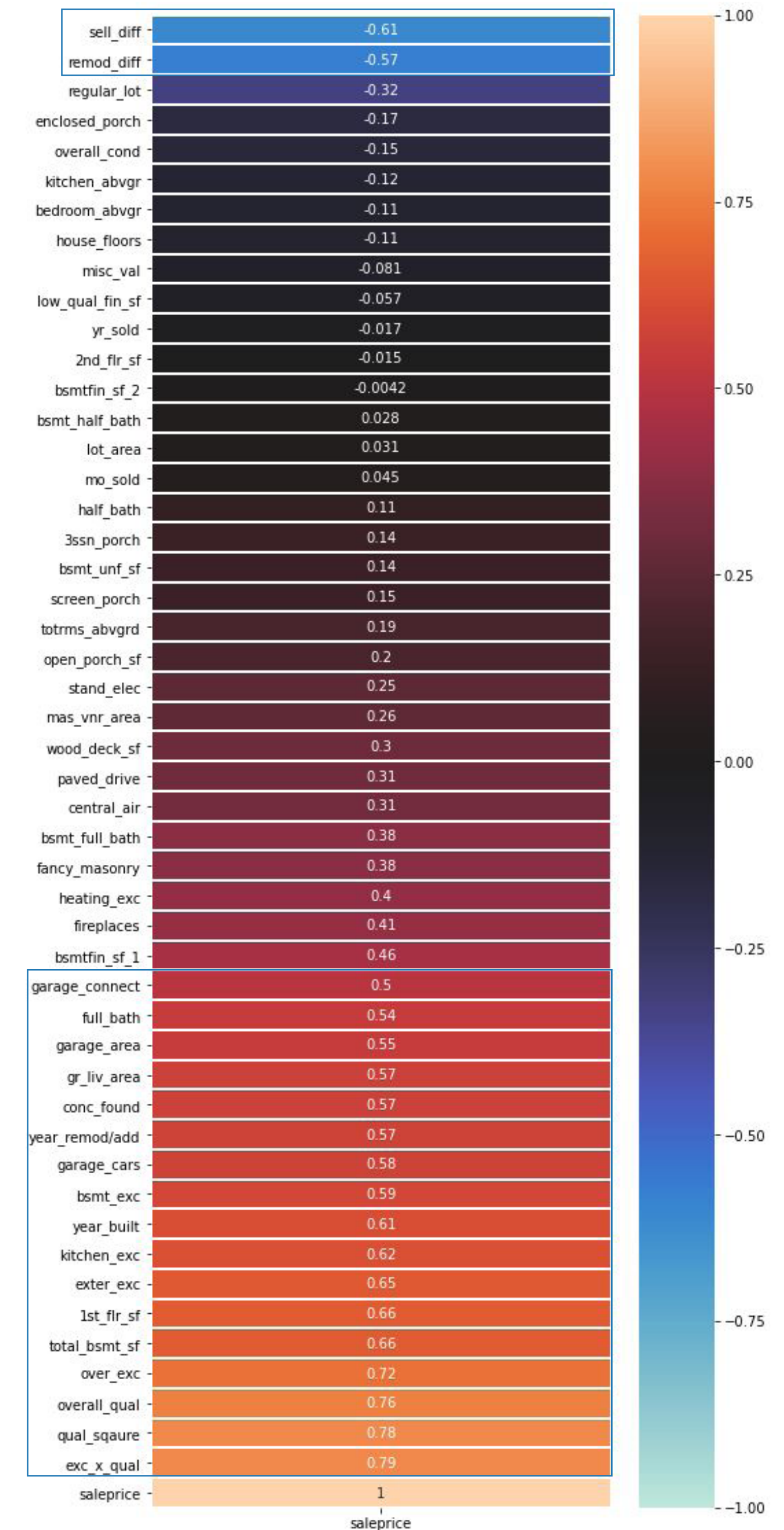
Absolute correlation value of 0.5 or higher

At least \$5,000 differential between values

No value greater than 90% of total value

Forgettable Features: Regular lot?, Fancy masonry?, Paved driveway?,
Heating exceed average?, Central air? Standard heating?

Bad Feature: Amount of floors



Method One

- Year Built
- Year Last Remodeled/Rebuilt
- Overall Quality
- Above Grade SF
- Above Grade Rooms
- Garage SF
- Garage Capacity
- Basement SF
- First Floor SF
- Number of Bathrooms
- Masonry Veneer SF

Method One

————→

Method Two

Year Built
Year Last Remodeled/Rebuilt
Overall Quality
Above Grade SF
Above Grade Rooms
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Garage Capacity
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House Age at Sell
Time Since Last Remodel
Neighborhood
Garage Connected to House?
Concrete Foundation?
Basement Exceed Average?
Kitchen Exceed Average?
Exterior Exceed Average?
Features over Average
(Features over Average)*(Overall Quality)













Method One

Method Two

Method ∞

Ridge Scores	
Cross Variable R ²	0.824
Mean Squared Error	546,945,723
Root Mean Squared Error	23,386

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Cross Variable R ²	0.824
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Method Two

Ridge Scores	
Cross Variable R ²	0.860
Mean Squared Error	458,006,241
Root Mean Squared Error	21,401

Method ∞

Method One

Ridge Scores	
Cross Variable R ²	0.824
Mean Squared Error	546,945,723
Root Mean Squared Error	23,386

Method Two

Ridge Scores	
Cross Variable R ²	0.860
Mean Squared Error	458,006,241
Root Mean Squared Error	21,401

Method ∞

OLS Scores	
Cross Variable R ²	0.873
Mean Squared Error	333,578,729
Root Mean Squared Error	18,264

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	Coefficient	Effectiveness
Kitchen	8,111	11.73x
Basement	691	1.00x
Exterior	-950	-1.37x

Method Two

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In Conclusion

Marked improvement when transforming the data

Binary variables with skewed splits (90/10) less effective for modeling

Remodeling efforts have great impact

Still heavily reliant on experts' opinion

Who sold the house?, Staging?, Time on the market?

Find out subjective scoring methodology

Are there better evaluators out there?

Fit more house area improvements to find out best areas to improve