



# Ames Housing Price Analysis

Siddharth Patel  
June 14, 2021



# Overview

Housing prices can be affected by

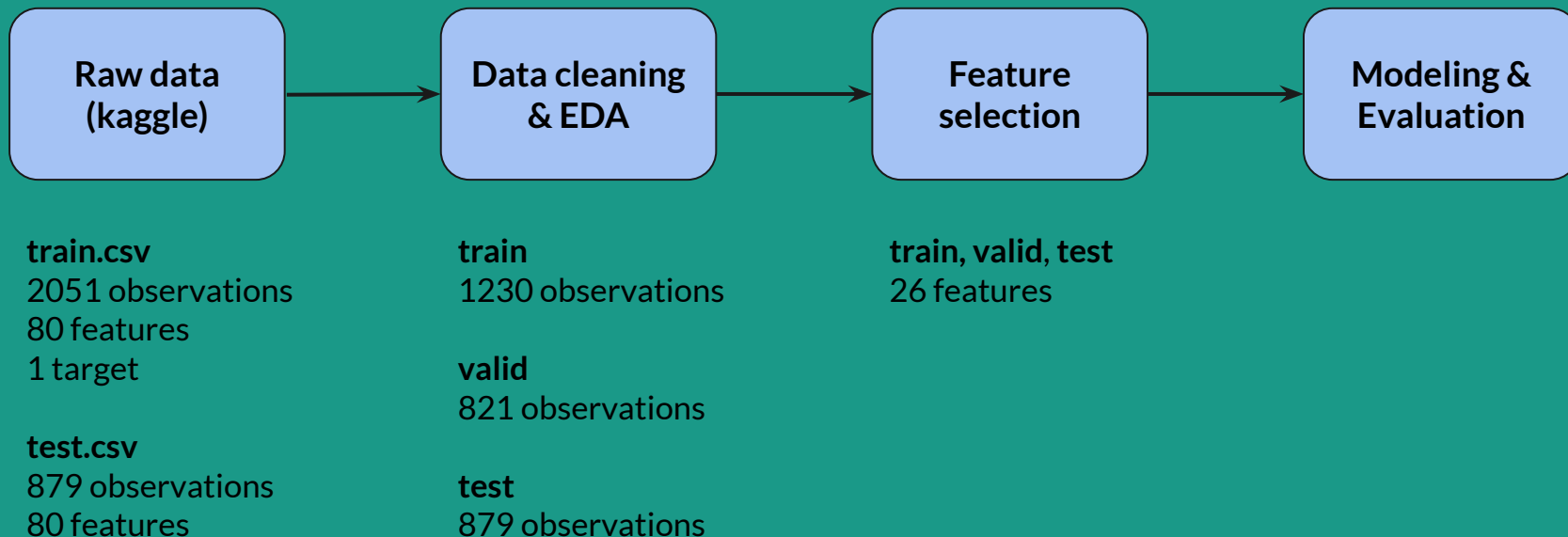
- Extrinsic factors: policies, tax, market [not in control]
- Intrinsic factors: plot area, amenities, beds & baths, quality [can be quantified]

## Problem Statement

As a data analyst we are tasked with the challenge to use this quantified housing data, and make analytical predictions for the property sale prices in the city, using simple machine learning methods.

At our agency, we use this data to predict house prices for any seller that intends to make a sale for their property based on the house's features.

## Data workflow

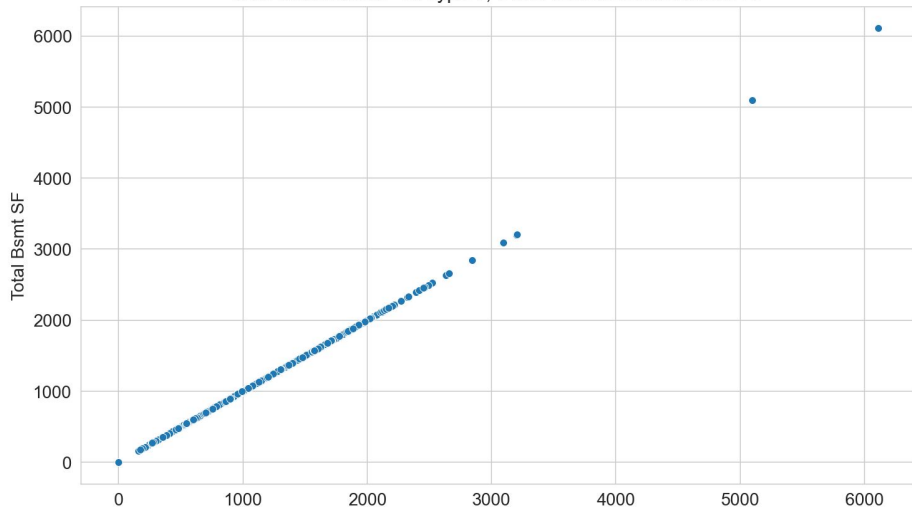


# Data Exploration

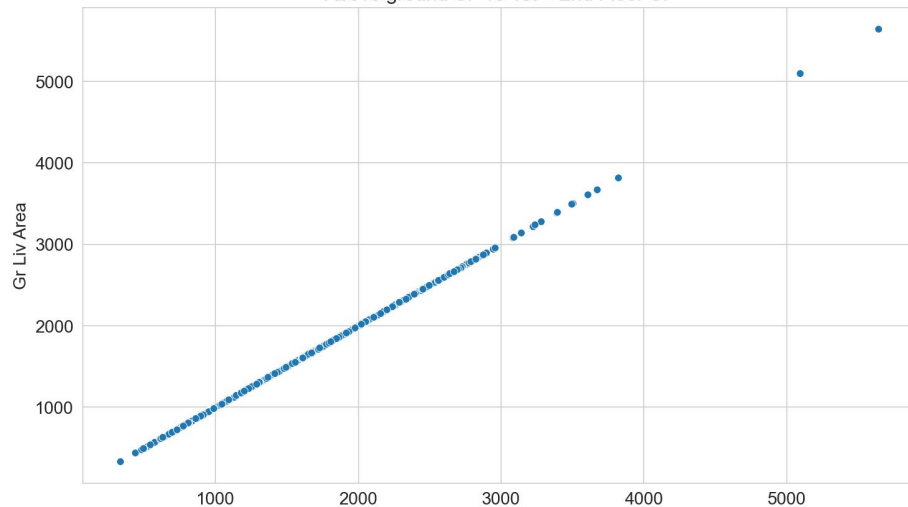
- Dropped columns
- Added interaction columns
- Replace NaNs with NA/mode/mean
- Binarized columns
- One-hot encoded ordinal features
- Outlier threshold

# EDA - columns that add upto another column

Total basement SF vs Type 1, 2 and unfinished basement SF



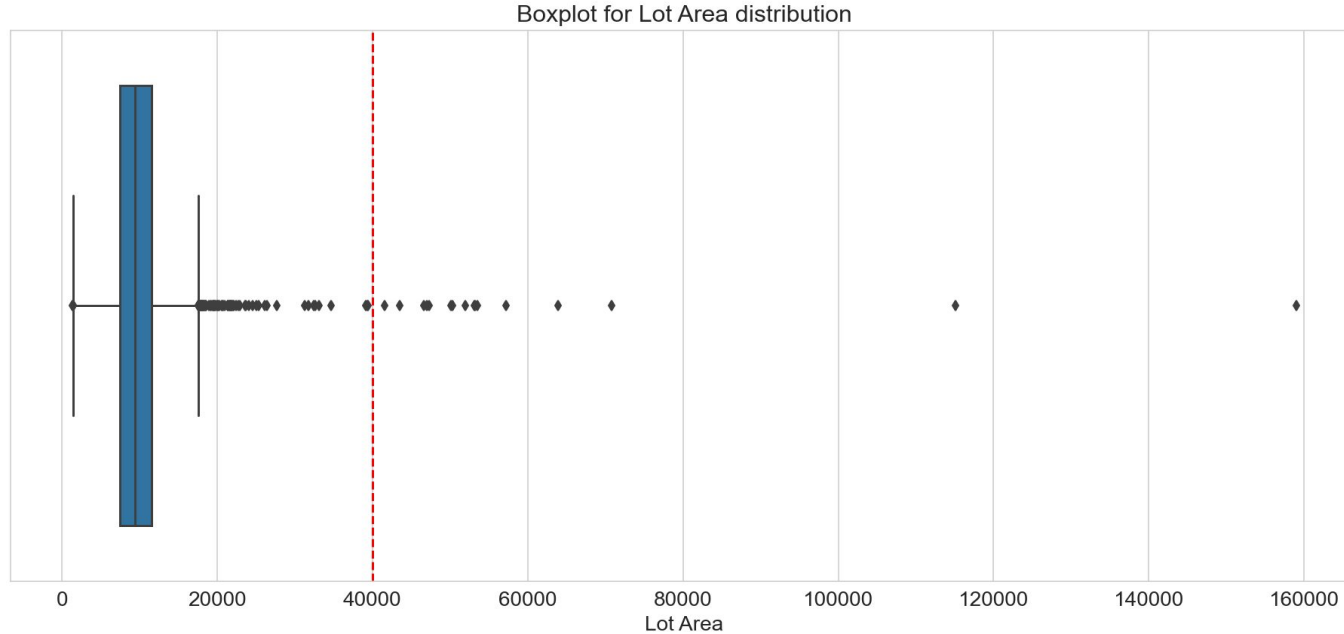
Above ground SF vs 1st + 2nd Floor SF



Total Bsmt SF = BsmtFin SF 1 + BsmtFin SF 2 + Bsmt Unf SF

Gr Liv Area = 1st Flr SF + 2nd Flr SF + Low Qual Fin SF

# EDA - outliers need to go



Lot Area box plot showing outliers > 40000

# Modeling & Feature Selection

Item 1

Item 2

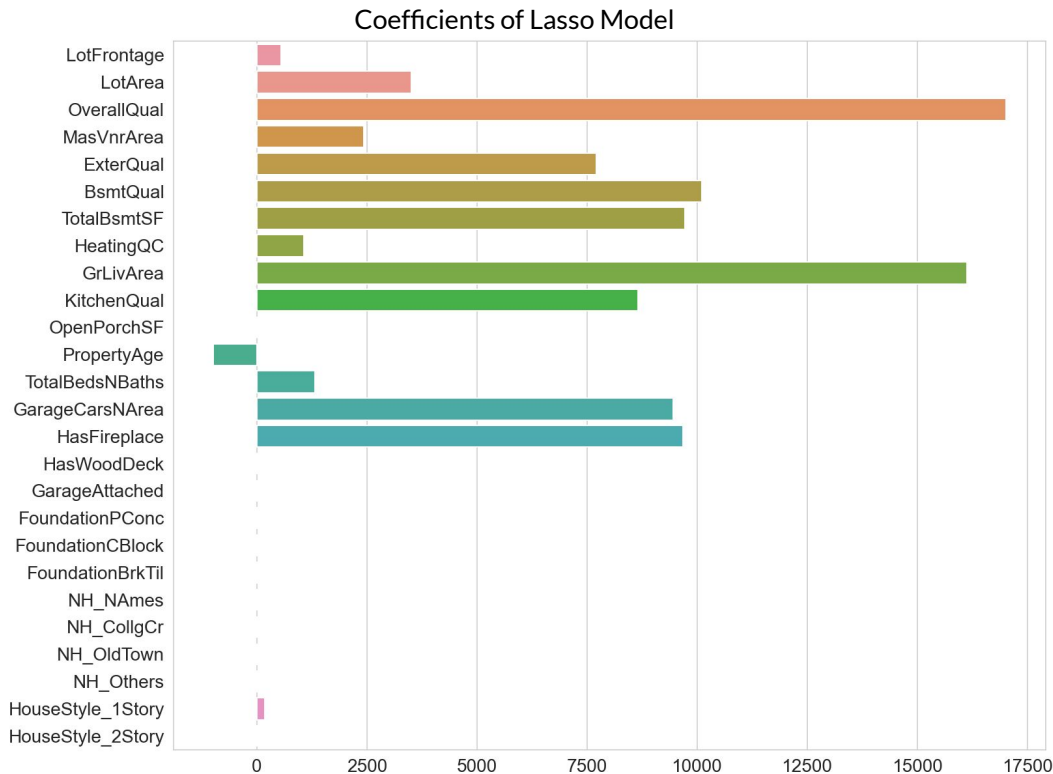


SalePrice increases with Ground Living Area



SalePrice decreases with PropertyAge

# Modeling & Feature Selection



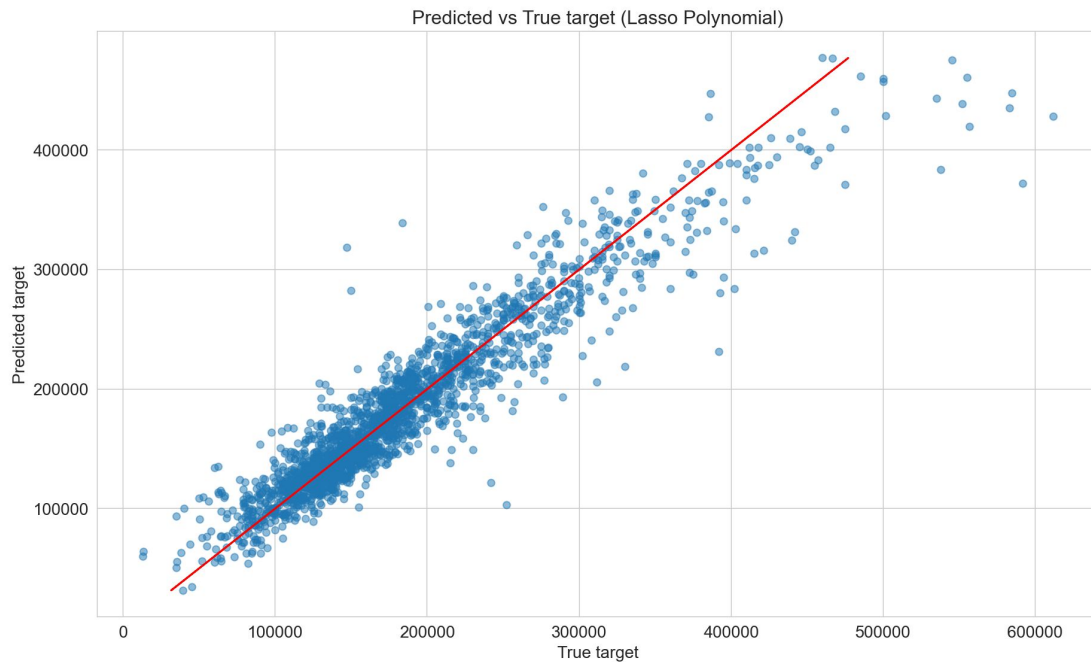
Models built:

1. SLR
2. MLR
3. Ridge
4. Lasso
5. ElasticNet

Lasso has the capability to zero the coefficients that are not significant.



# Results - Lasso



## RMSE summary

Model	Train	Test
Baseline	\$77014.97	\$80259.83
Production	\$24470.15	\$28435.67

# Conclusions

- 80 → 26 features selected
- Scaling and polynomial feature interactions
- Production model: Lasso regularized regression with polynomial features
- 306 / 351 polynomial interaction terms zeroed
- Most positive impact features:
  - Basement Quality
  - Exterior Quality
  - Kitchen Quality
  - Overall Quality
  - Has Fireplace

# Thank You!