Ames Housing Price Analysis

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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.

GA
GENERAL ASSEMBLY

Understanding data
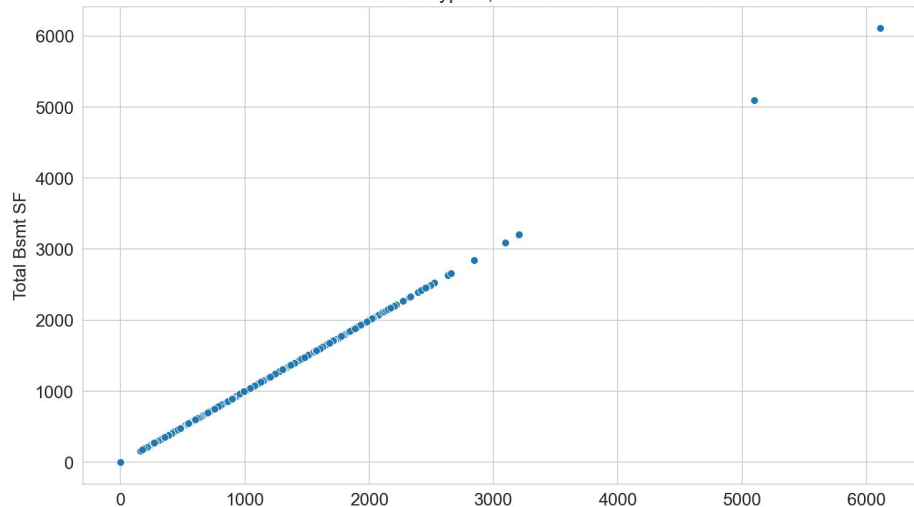
Kaggle

Train.csv

Train

Valid

Test.csv
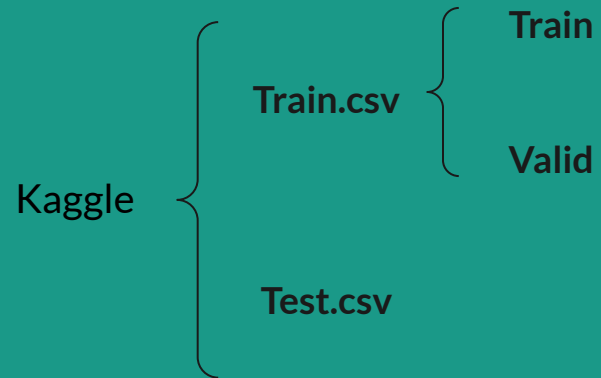
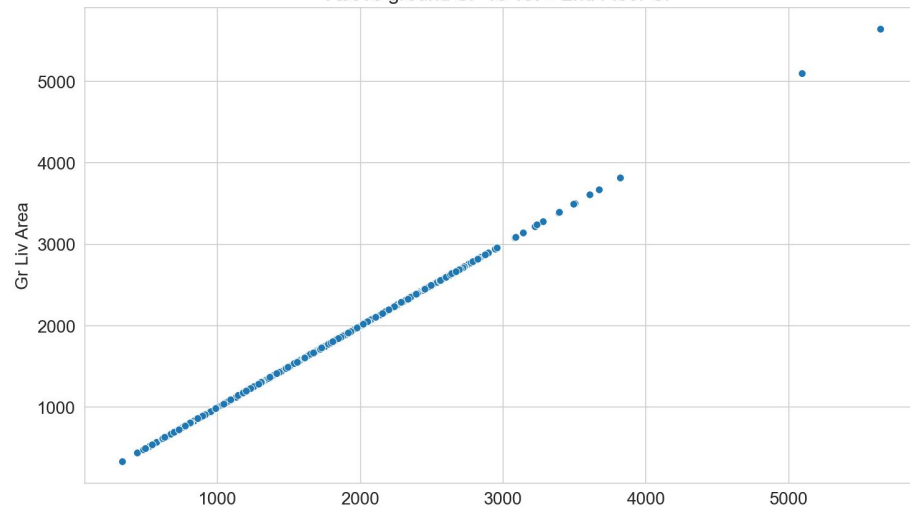# Workflow

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

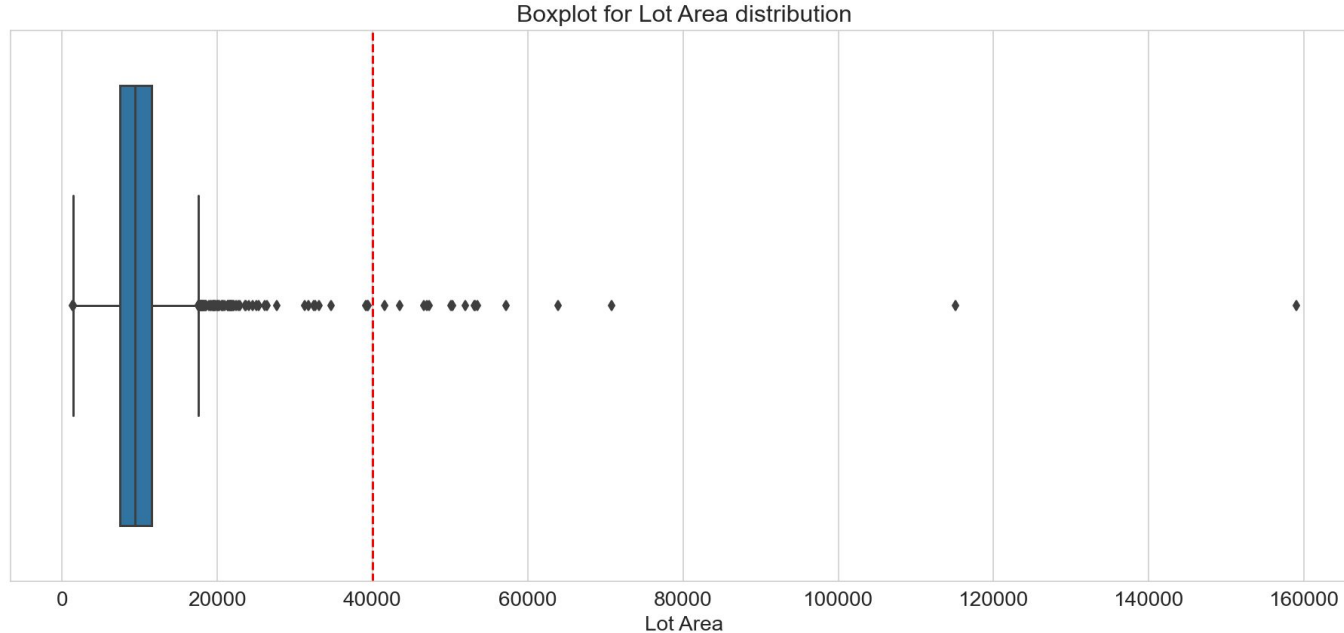Total basement SF vs Type 1, 2 and unfinished basement SF

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

Boxplot for Lot Area distribution

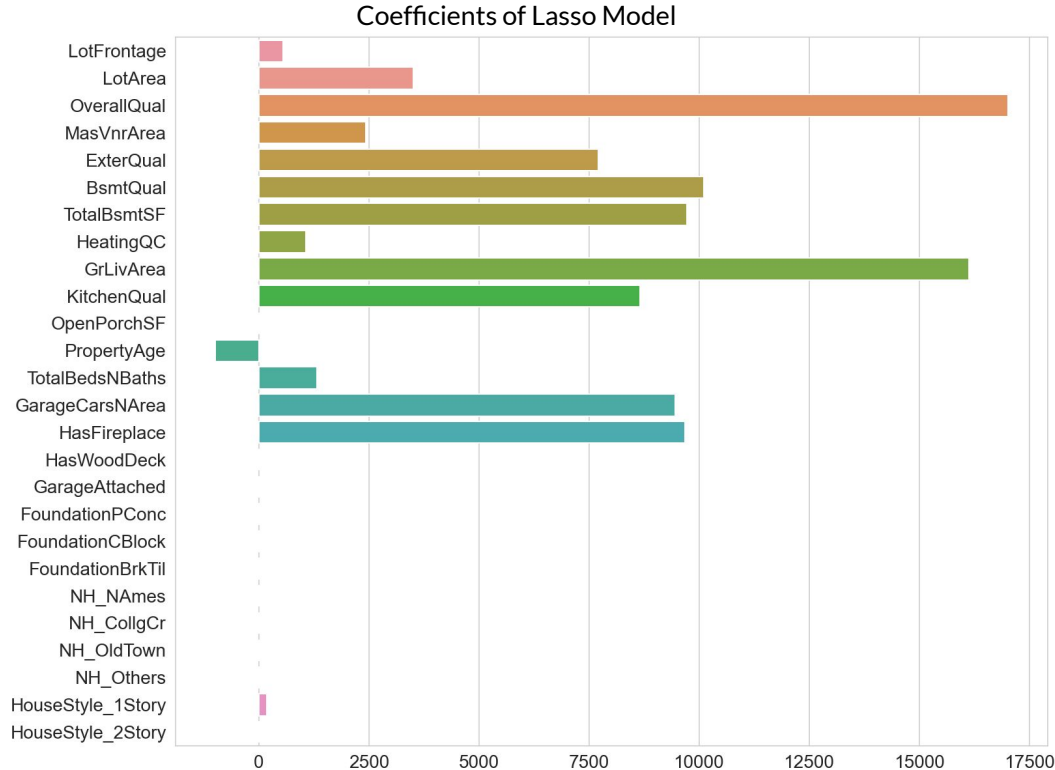Lot Area box plot showing outliers > 40000

Modeling & Feature Selection
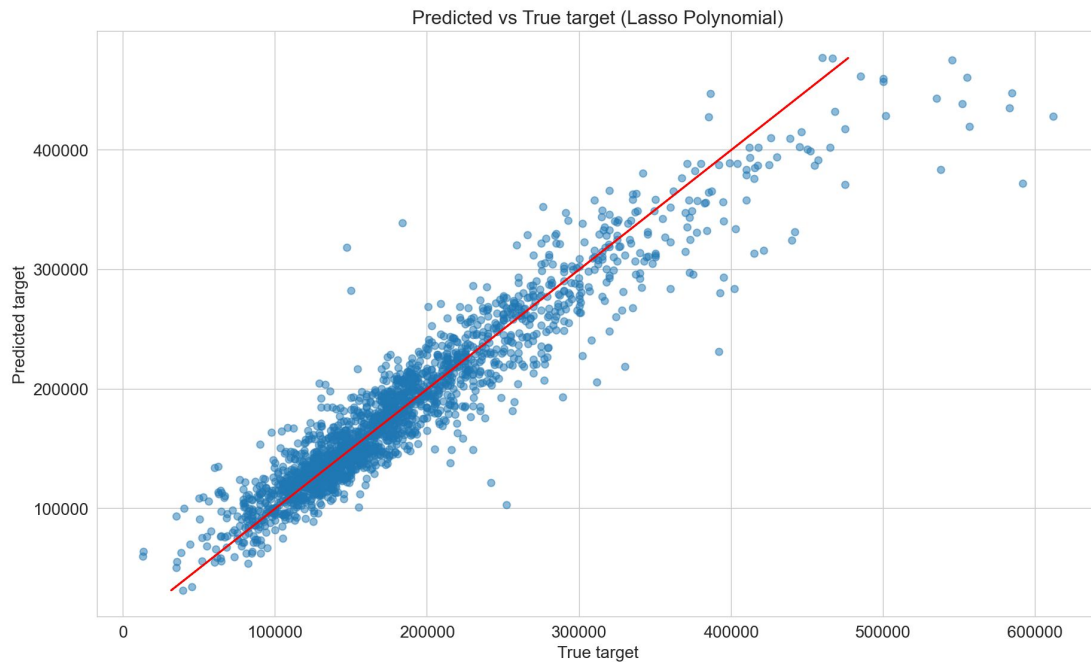

SalePrice increases with Ground Living Area

SalePrice decreases with PropertyAge

Modeling & Feature Selection

Coefficients of Lasso Model

Results

Predicted vs True target (Lasso Polynomial)

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

- Most positive impact features:

  - Basement Quality

  - Exterior Quality

  - Kitchen Quality

  - Overall Quality

  - Has Fireplace

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GENERAL ASSEMBLY

Thank You!