Ames Housing Sale Price Prediction

Team 5:

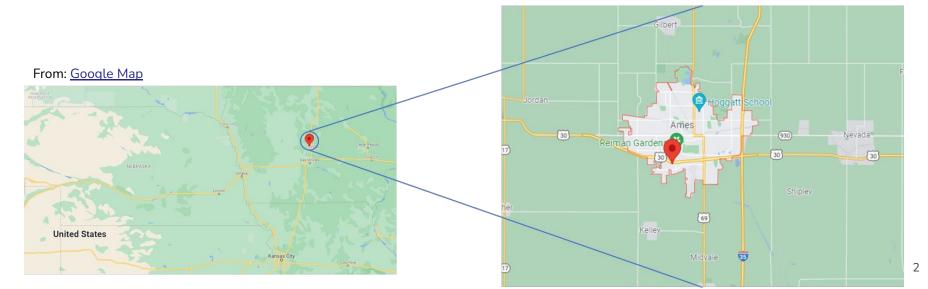
Bryan Soh Mitchelle Chua Vincent Chua Zavier Soon



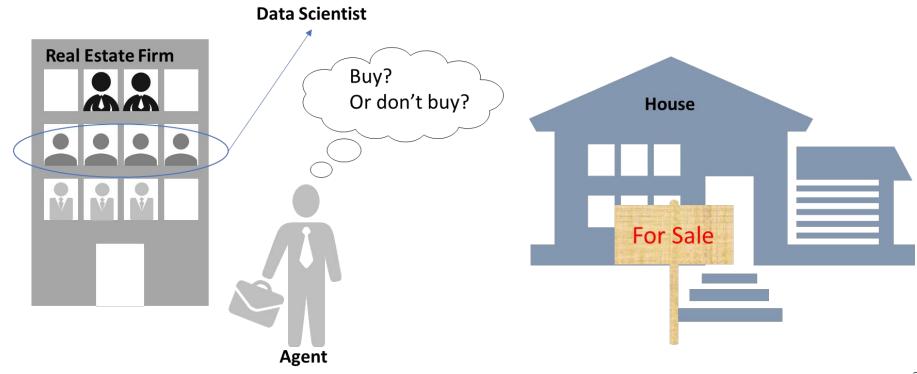


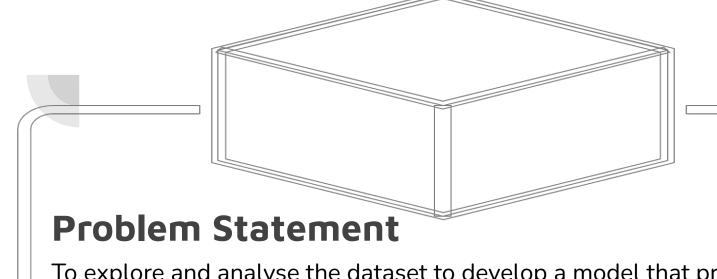
Official Website: https://www.cityofames.org

Ames is a city in Story County, Iowa, United States, located approximately 30 miles north of Des Moines in Central Iowa. It is best known as the home of Iowa State University, with leading agriculture, design, engineering and veterinary medicine colleges.



Background





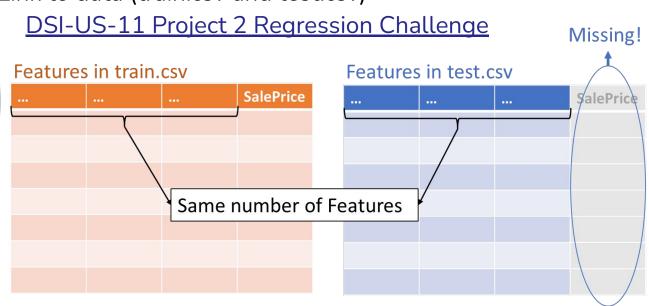
To explore and analyse the dataset to develop a model that predicts the housing sale price in Ames, Iowa.

Using the model built, identify

- 3 features that will increase the sale price 3 features that will lead to a decrease in the sale price 3 features recommended for renovation



Link to data (train.csv and test.csv)



Data Dictionary:

https://web.archive.org/web/20201203235151/http://jse.amstat.org/v19n3/Decock/DataDocumentation.txt

Feature and Data imputation

Features dropped:

1. Lot Frontage (more than 5% of the data are missing)

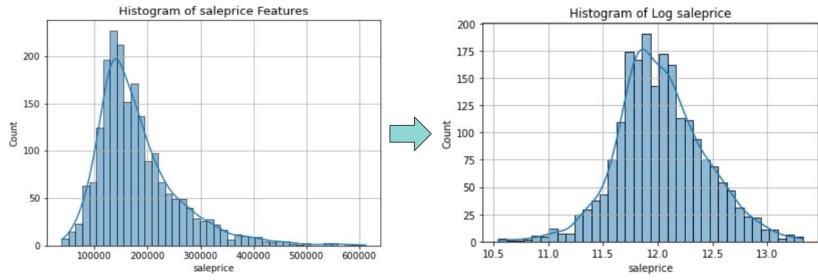
Data dropped in train.csv

- 1. Data with lots of missing values
- 2. Data with extreme values (outliers)



Feature Engineering

1. Saleprice Log Transformed



- 2. Three New Features:
 - Total square feet of a House
 - Quality Score
 - House Age

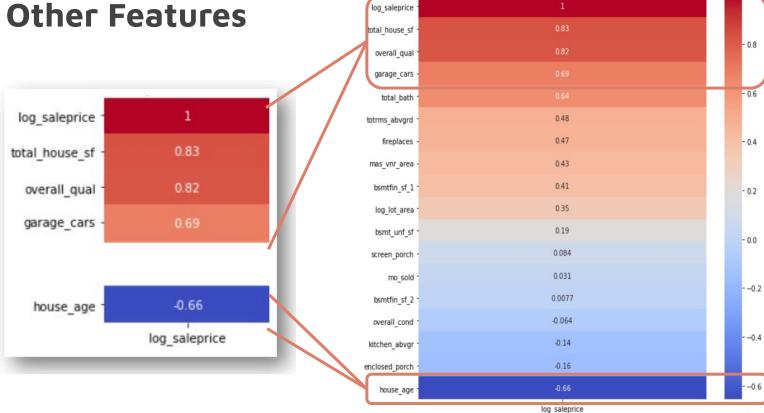
Median Property Value in Ames, IA (2006-2020)



References:

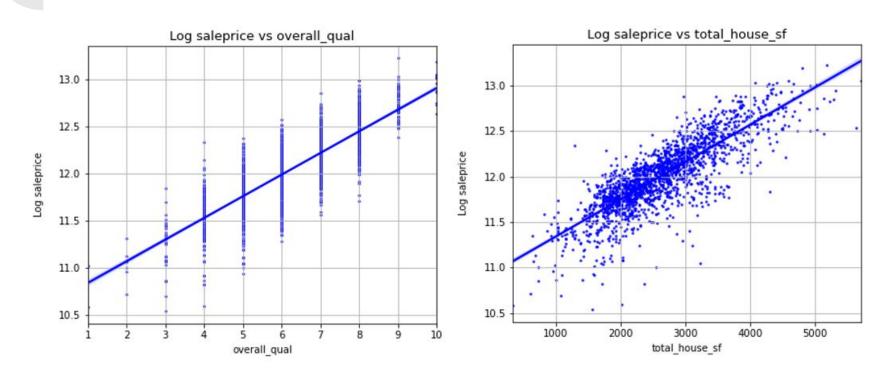
- 1. https://walletinvestor.com/real-estate-forecast/ia/story/ames-housing-market
- 2. https://data.census.gov/cedsci/table?g=DP04&tid=ACSDP5Y2019.DP04

Correlations of Sale Price vs Other Features

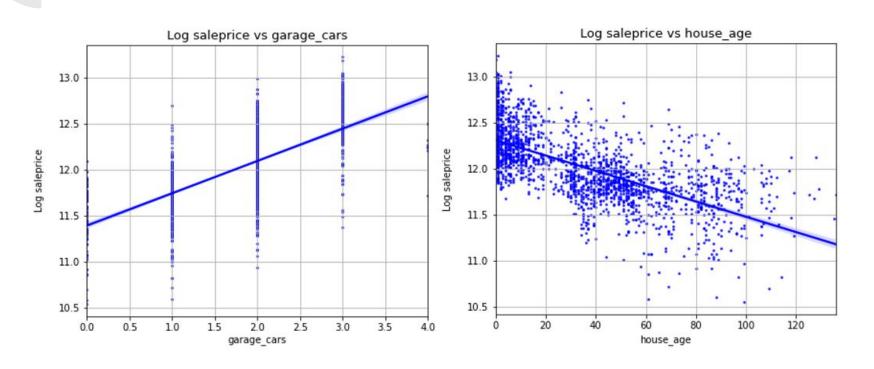


Correlations of log saleprice vs other numerical features

Top 4 Features Highly Correlated Sale Price



Top 4 Features Highly Correlated Sale Price



Our Modelling Process Final Model & Prediction Use our selected model for prediction **Model Evaluation & Selection** Use evaluation metrics R² Score and Root Mean Square Error (RMSE) **Regularization Model** Ridge Regression & Lasso Regression **Baseline Model** Linear Regression **Model Preparation** One Hot Encoding, Scaling, Train-Test-Split

Model Preparation

- One-Hot Encoding our categorical variables
- Moving from 77 features to 235 features
- Create more noises and increase risk of overfitting
- Scaling our features
- Train-Test Split, 80% of train datasets



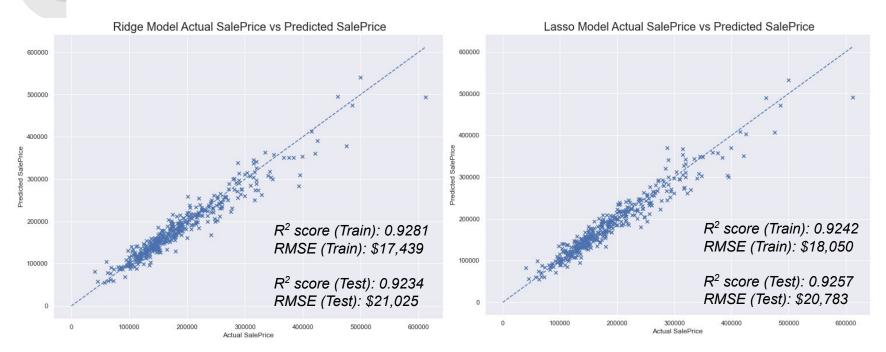


Baseline Model: Linear Regression



★ Coefficient of Determination (R²) - This is to explain the accuracy of our model fits the observed data. The higher the r-squared indicated a better fit for the model

Regularization Model: Ridge vs Lasso Regression



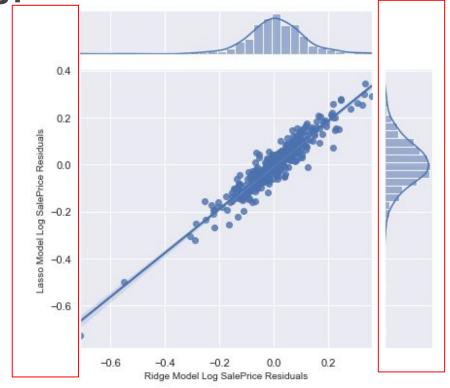
- ★ Coefficient of Determination (R²) This is to explain the accuracy of our model fits the observed data. The higher the r-squared indicated a better fit for the model
- ★ Root Mean Squared Error (RMSE) the square root of the variance of the residuals. Lower values of RMSE indicate better fit.



Evaluation of Metrics:

Ridge Model Residuals against Lasso Model Residuals

	Ridge Regression	Lasso Regression	
Cross Validation Score	90.25%	90.15%	
R ² Score (Train)	92.81%	92.42%	
R ² Score (Test)	92.35%	92.57% 0.	
RMSE (Train)	\$ 17,439	\$ 18,050	
RMSE (Test)	\$ 21,025	\$ 20,783	



Residuals - Difference between actual sale price and predicted sale price

Making sense of the Lasso Model Results



Variability or fluctuation in the test data sale price can be explained by the predictor features.



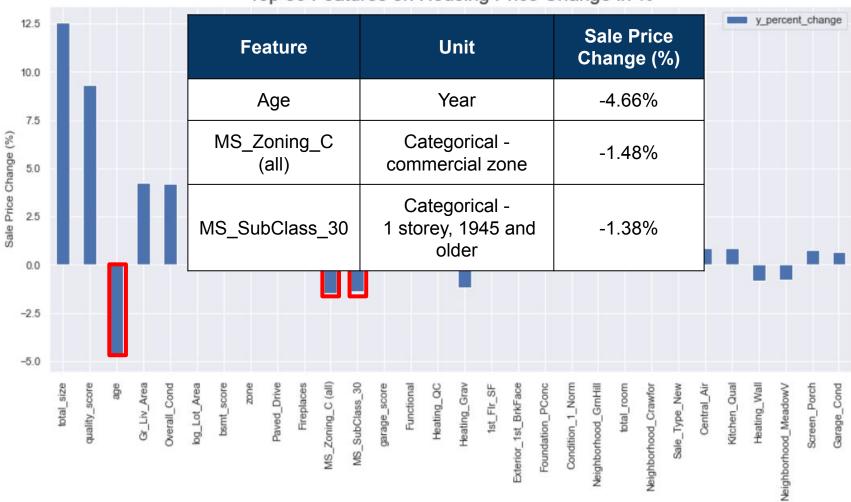
USD 20,000

Predict the sale price within +/- USD 20,000





Top 30 Features on Housing Price Change in %



Top 3 Impactful Features for Renovation

Features	Breakdown	Levels	Impact per increase in level	Action
Overall Cond	Very Poor to Very Excellent	10	+4.21%	Repair and painting works
Paved Driveway	Dirt/Gravel to Fully Paved	3	+1.83%	Pave the full driveway
Home Functionality	Salvage Only to Typical Functionality	8	+1.22%	Repair damages



Conclusion & Recommendations

- High impact features are consistent with common sense, which lends credibility to the model
 - High quality, in good condition, newly built, large in size

- In terms of whether to renovate the house, the cost should be lower than the predicted increase in Sale Price

 The underlying guiding principle will always be to buy a house with a selling price lower than its true value

