# Project 2

Ames-housing Price Prediction

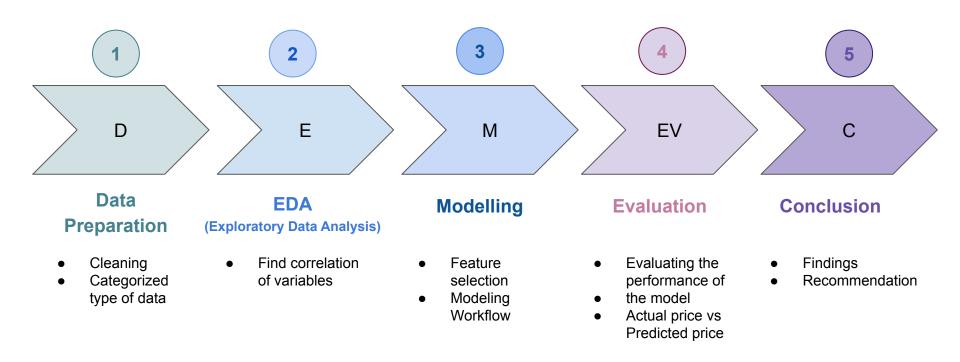
#### **Problem Statement**

If want to build a house for selling, and what main factors to think about? And what price should it be?

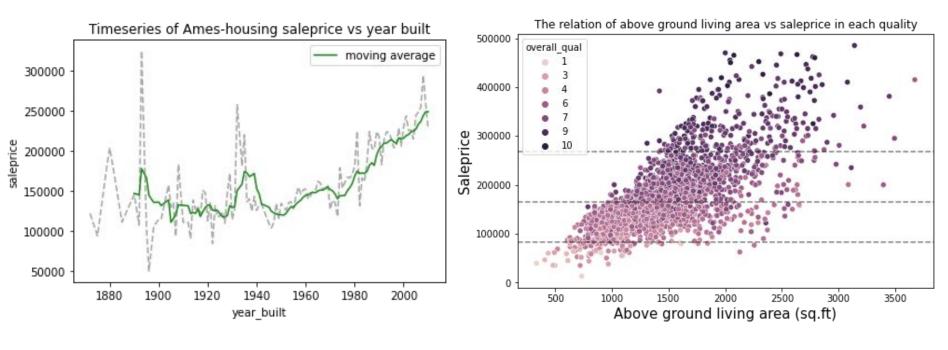
### Objective

To create the linear regression model, for predicting the price of housing in ames city.

#### Outline



### EDA - What affect to the Sale price?



 Sale prices of houses trend to increase as the newer of the house  The larger of living area, the higher of the sale price, and overall quality

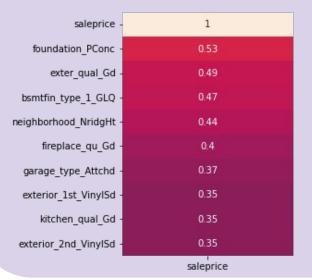
### Feature engineering (I) - Imputation & Filtering

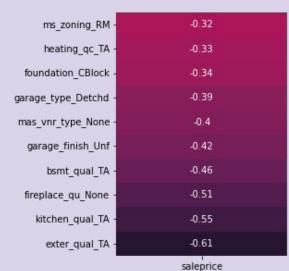


Numeric type: Impute with '0'; And use all of numeric features

Object type: Impute with 'NONE';

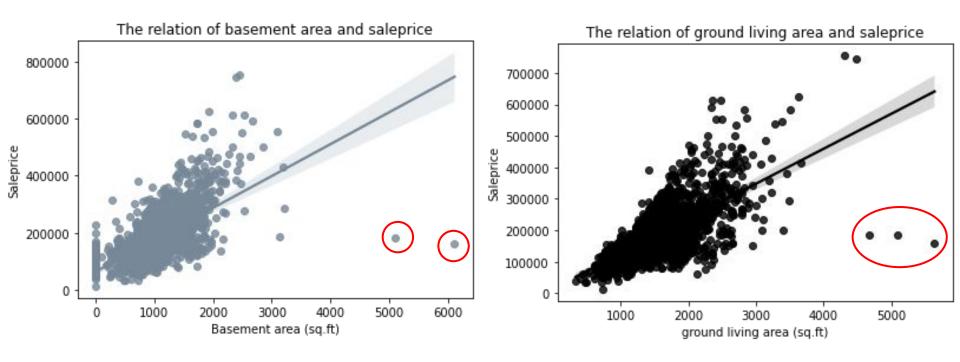
Dummy and use high correlated features to sale price





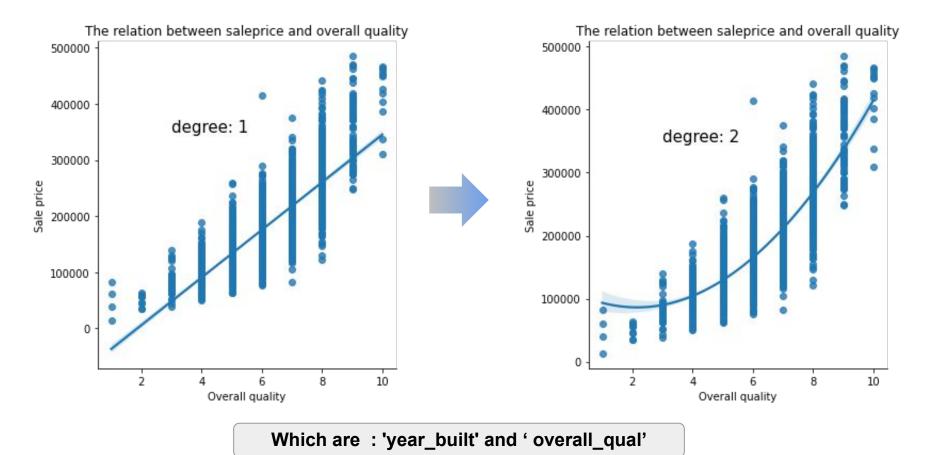


### Feature engineering (II) - Outliers



Which are: 'total\_bsmt\_sf', 'gr\_liv\_area', '1st\_flr\_sf', 'saleprice', 'mas\_vnr\_area', 'garage\_area'

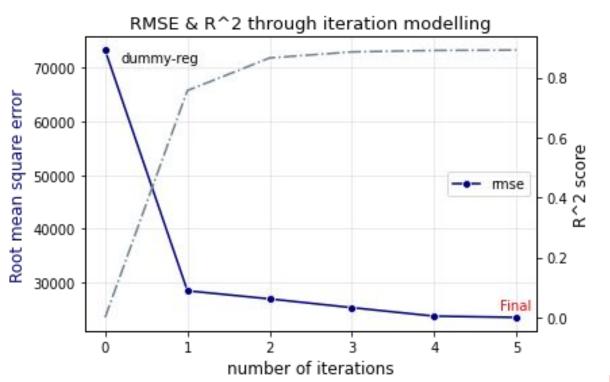
### Feature engineering (III) - Polynomial term



### **MODELING FLOW - 5 STEPS**

Using Train-Test split to split the data Evaluate the model by cross-validation **ONLY NUMERIC** + CLEAN OUTLIER + DUMMY + Poly term + LASSO

#### **EVALUATION**

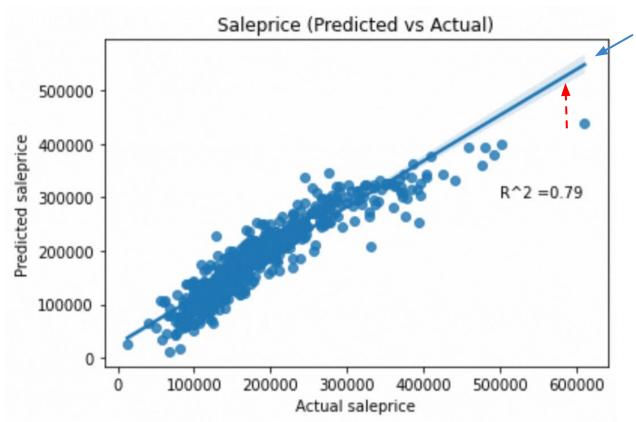


## The development of modeling performance

Iterations	RMSE	R^2
Dummyreg	73220	0.000
1	28503	0.757
2	26997	0.865
3	25373	0.886
4	23797	0.891
5	23555	0.892

Final model: RMSE decreased from first model around 5000

### **EVALUATION (II) -** Actual vs Predicted of Sale price



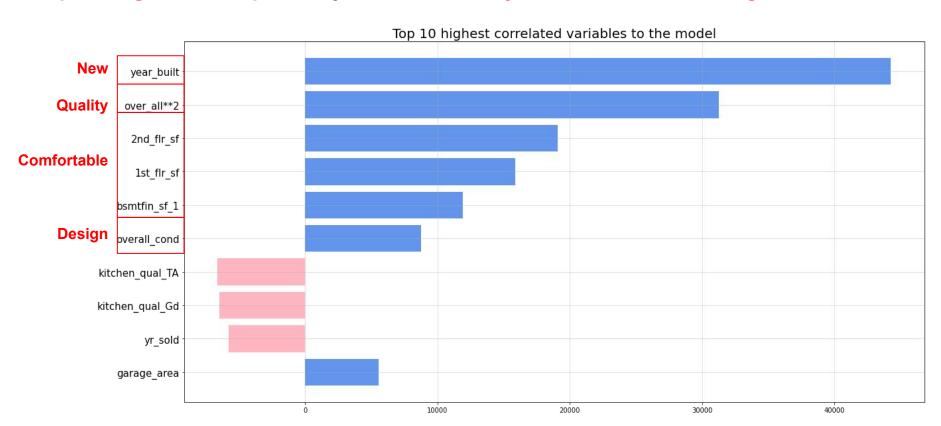
#### The reference line:

Actual price = Prediction price

 This developing route of the model will increase the accuracy of sale price prediction, as the increased of R^2

#### List of high correlated variables to the model

Top 4 things that sale price rely on : New, Quality, Comfortable, and Design.



#### Conclusion:

- The top 4 things that influence the sale price are: new (year built); quality (overall quality); comfortable (size of living area); and design (overall condition).
- The model has 60 features, including 41 numeric features and 19 dummied features.
- This model includes the use of train-test split to split the data for training and cross validation to test the performance of the model.
- This model includes cleaning outliers, adding polynomial terms & dummy variables, and regularization by LASSO.
- R^2: 89.2% of the variability in sale price of ames houses can be explained by this model.
- RMSE: The final model can reduce RMSE by around 5000. As a result, the price predicted by this model may differ from the actual value of around 23555 dollars.

#### Recommendation:

should add more the interaction term, and other variables

**THANK YOU** 

**FOR LISTENING** 

(KHOB KHUN KRUB)