

Introduction



 Find the best ML Model to predict US movie gross earnings to decrease financial risk

Methodology



Data Scraping

01

- Scraping IMDb Website
- 3000 records and 9 features



Data Cleansing

02

- Disintegrating some columns
- Detecting and handling missing data and outliers



Exploratory Data Analysis

03

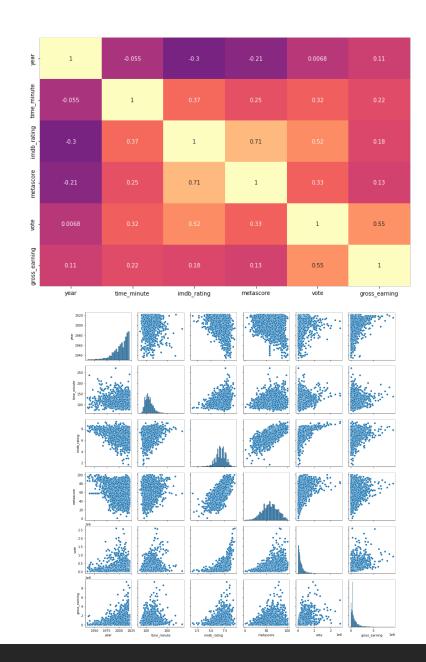
- Looking into feature correlations, pairplots, etc.
- Understanding the transformations needed to make data suitable for regression model



Building Models

04

- Creating different regression models.
- Using cross-validations, and using evaluation metrics to select the final model



- 0.75

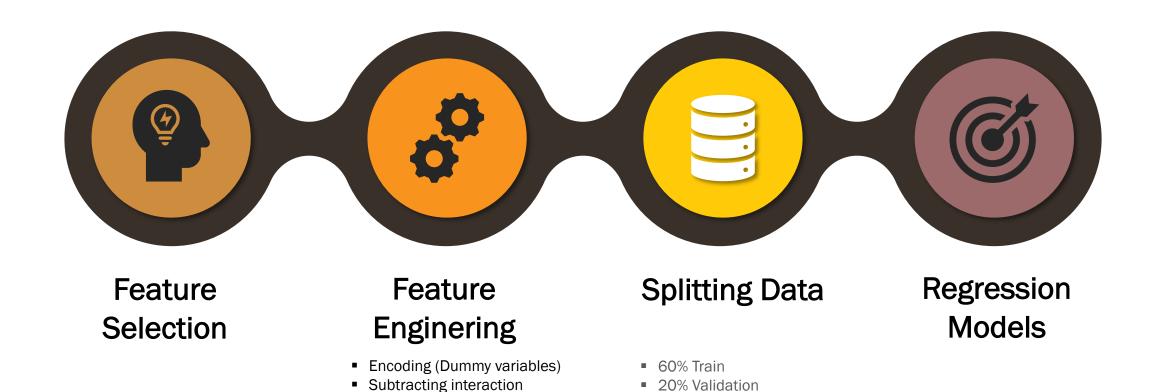
- 0.50

- 0.25

-0.25

Data Preparation

terms

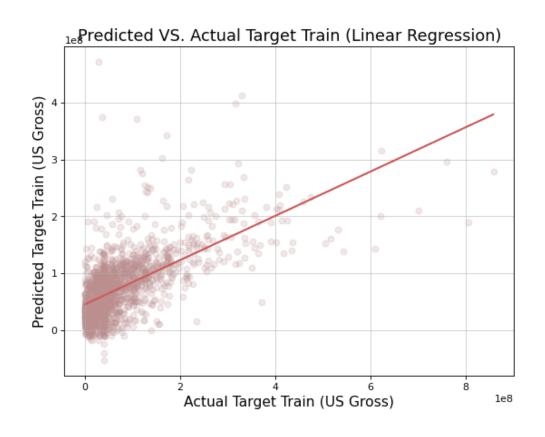


20% Test

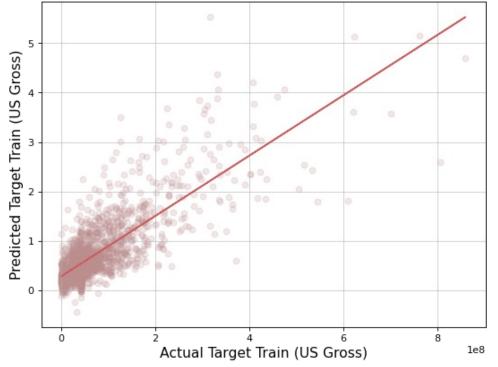
Analysis And Results

Regression Models	Training Score	Validation Score
Normal Linear Regression	0.389285939	0.461151975
K-fold Linear Regression	0.389285939	0.379048417
Polynomial Regression Degree 2	0.00471088	-0.07468319
Polynomial Regression Interaction	0.50838276	0.3800856
Ridge Regression	0.46113795	0.38928325
Ridge Regression Cross-Validation	0.389283247	0.379102316
Lasso Regression Cross-Validation	0.389285939	0.379048764
Ridge Regression on polynomial		
interaction only	0.61451453	0.54744346

Analysis And Results

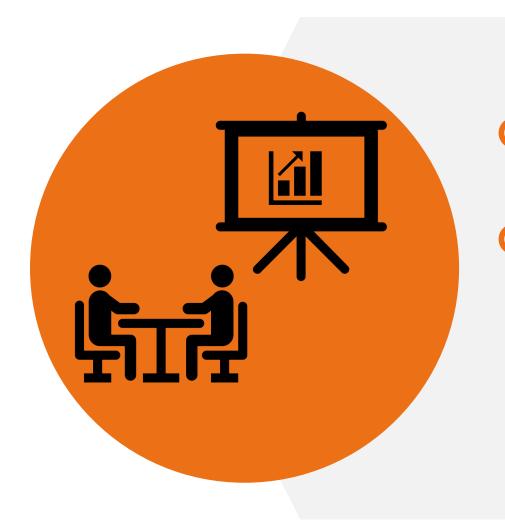


Predicted Y.S. Actual Target Train (Ridge Regression on poly interaction only)



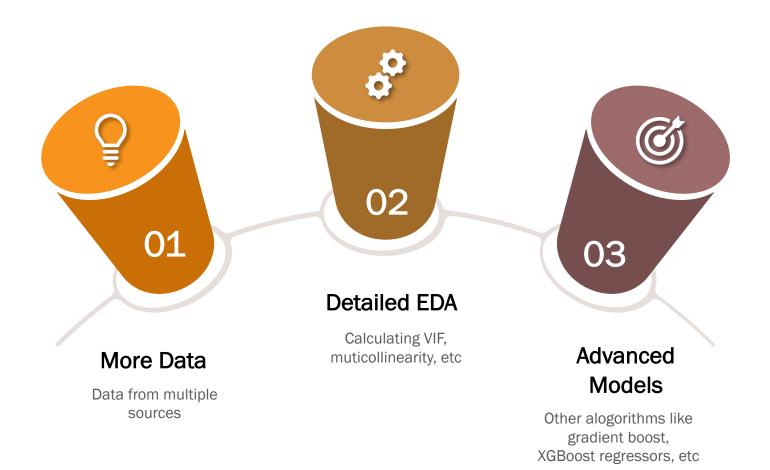
Testing Score: 0.59082610

Conclusion



- The objective of this project was to predict US movie gross earnings to decrease financial risk
- Thus, based on the evaluation metrics, Ridge Regression on polynomial interaction only features is the best model for prediction of gross earnings.

Future Work



THANK YOU

Appendix

Analysis And Results



