

# Regression Analysis:

## Predicting the International vs. Domestic Share of Box Office Revenue

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# INTRODUCTION

## OVERVIEW

- Global markets are significant sources of revenue for US made movies
- Predicting what kind of movie will produce a high international response would be helpful for a movie distributor to have in its business decision making toolkit

## PROJECT GOAL:

Using data from Box Office Mojo, build a predictive model for international revenue percentage (%) of revenue for domestic made movies

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# METHODOLOGY

## DATA

**Sample Frame:** 1000 domestic movie web pages from Box Office Mojo Top Lifetime Grosses page

### Variables:

- *Target Variable (1):* International Revenue %
- *Feature Variables (95):*
  - Numeric Vars: [Year, Run Time, Budget (adjusted for inflation)]
  - Categorical Vars: [Distributor, Rating, Genres, Directors, Actors, Release Month]

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# METHODOLOGY

## TOOLS

- **Web Scraping:** BeautifulSoup, Pickle
- **Data Visualization:** Matplotlib and Seaborn
- **Regression Analysis:** SciKitLearn, statsmodels

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# METHODOLOGY

## METRICS

- **Model variable review (feature engineering)**
  - Fit linear regression to sample data
  - Pairplots, heatmaps, and VIF analysis to ID/address collinearity
  - Residuals scatter plot to ID/address heteroskedasticity
- **Trained and tested 4 regression models** scored for explanation power ( $R^2$  value) and magnitude of error value (MAE):

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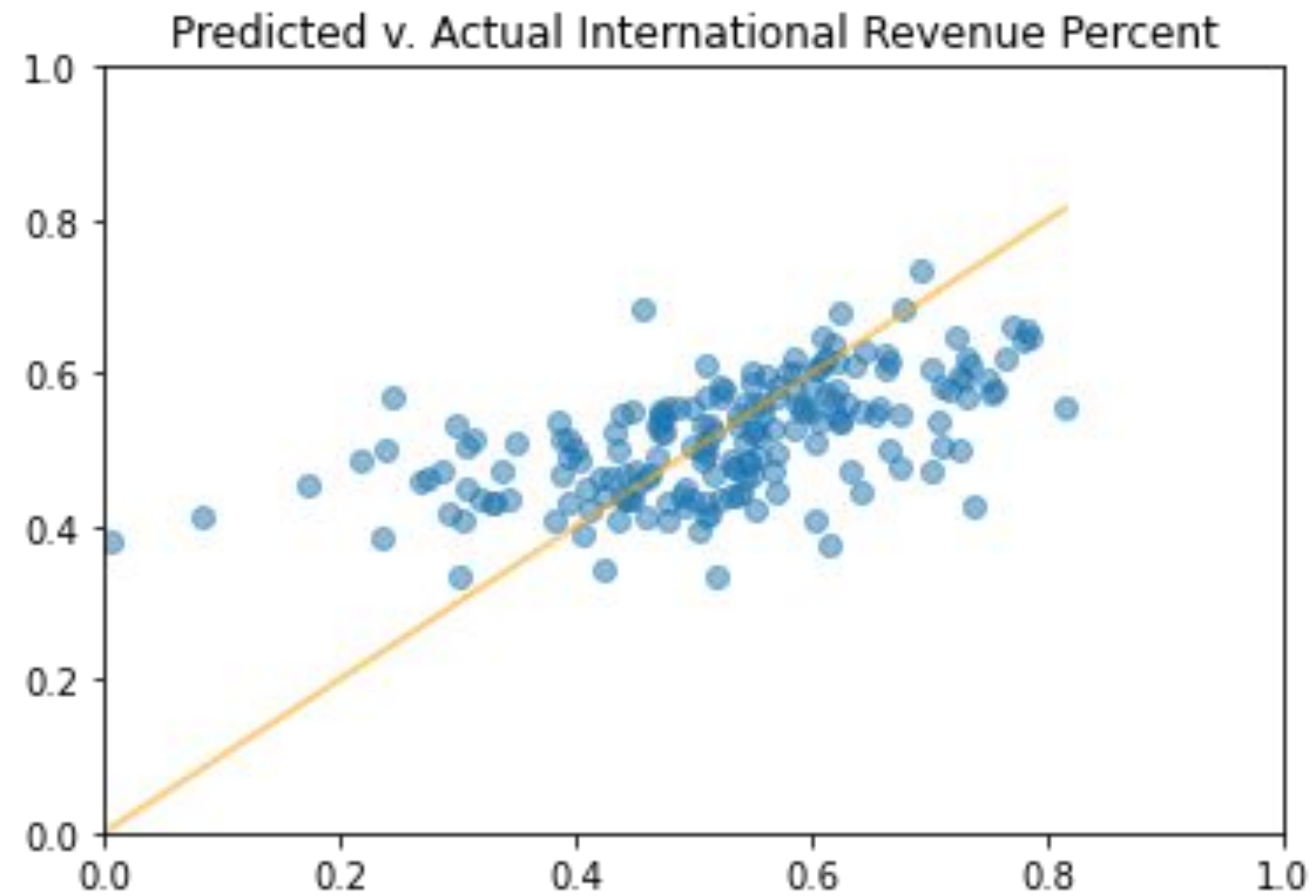
# METHODOLOGY

## METRICS

- **Standard OLS, cross validated (KFold)\*:**
  - $R^2 = 0.33 \pm 0.11$
  - $MAE = 0.1 \pm 0.01$
- **Polynomial OLS, cross validated (KFold)\*\*:**
  - $R^2 = -0.21 \pm 0.36$
  - $MAE = 0.13 \pm 0.02$
- **Ridge Regression, cross validated (RidgeCV)\*\*:**
  - $R^2 = .29$
  - $MAE = 0.09$
- **Lasso Regression, cross validated (LassoCV)\*\*:**
  - $R^2 = .33$
  - $MAE = 0.09$

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# RESULTS



Final Model:

Lasso Regression, cross validated (LassoCV)

$R^2 = .33$

MAE = 0.09

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# RESULTS

## Variables with highest coefficient (standardized)

- 'budget\_adj' 0.034
- 'year' 0.032
- 'Adventure' 0.022
- 'Animation' 0.012
- 'Thriller' 0.011
- 'Comedy' 0.015
- 'Will Ferrell' 0.016

## Variables that were dropped to zero during Lasso Regression:

- Run Time
- All Directors
- All Distributors
- All Months Released
- Most Actors
- Many Genres



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# CONCLUSIONS

Unfortunately, the model in its current form is not very predictive or useful for a business case.

## Ways to improve the model

- **Better sample Frame:**
  - Use a much larger data set
  - Use a data set within a more recent time frame (last 10 years for example)
- **Better target:**
  - International % may be too broad and capturing too many variables
  - Model can be refined to predict more specific targets, e.g. Chinese box office revenue
- **Better features:**
  - Search for another movie data website that has data that promises to be more explanatory for predicting international market targets