



Predicting Profits for Movie Releases

Charlie Reiney and
Noah Johnson



Data Collection

Sources

- The Movie Database
- IMDb

Query Criteria

- Profit
- Time-adjusted Popularity
 - 1991-2010: Pop. score of at least 5
 - 2011-2020: Pop. score of at least 2

Pre-Processing



Binarize movie collection data



One-Hot Encode genres



Parse release date (m/d/y)



Remove text attributes (Sentiment Analysis ☹️)



Discretize revenue (Classification)



Normalization/Scaling (SVR, SVM)



Feature Exploration



Models Overview

Regression

- Linear
- KNN
- Support Vector
- Reg. Tree
- XGBoost

Classification

- KNN
- Naïve Bayes
- Decision Tree
- Random Forest
- XGBoost
- Neural Network

Regression Results

XGBoost

**$R^2 =$
0.77850**

**MAE =
32283606.03**

**Regression
Tree**

**$R^2 =$
0.73081**

**MAE =
31457472.86**

**Linear
Regression**

$R^2 = 0.74132$

**MAE =
41071663.34**

Classification Results

XGBoost

67.157%
accuracy

Random
Forest

66.818%
accuracy

Decision
Tree

62.967%
accuracy

Real Predictions

- Movies: Venom, Dune, The Eternals, Clifford, Belfast, King Richard, Ghostbusters Afterlife, C'mon C'mon, Encanto, House of Gucci, and Resident Evil
- KNN Regression (k=2)
 - $R^2 = .93495$
 - MAE = 34,724,816.05
- Naïve Bayes
 - Accuracy = 81.8181%



Discussion

1

Improving Models

- Edit Neural Network
- Perform sentiment analysis
- Find a new ensemble regression model

2

Collecting More Data

- YouTube views on official trailer
- Top-billed cast members
- Social media interaction