

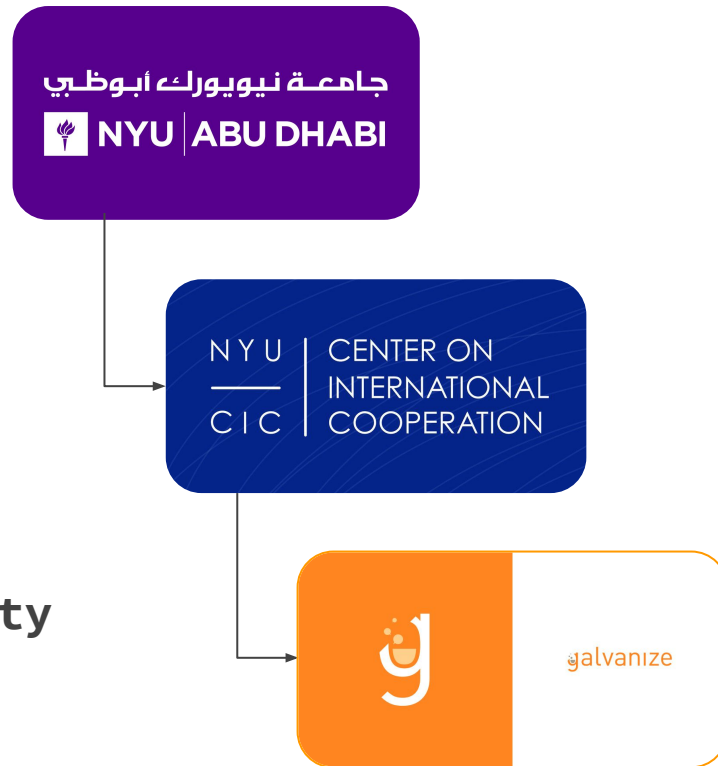
# Predicting Film Success

**Ryan Rappa**

# Me / my project

---

- Poli sci + econ background
- Interest in film & entertainment
- Project: predict film profitability



# Project workflow

Collect  
data



Clean  
data

pandas  
 $y_{it} = \beta' x_{it} + \mu_i + \epsilon_{it}$

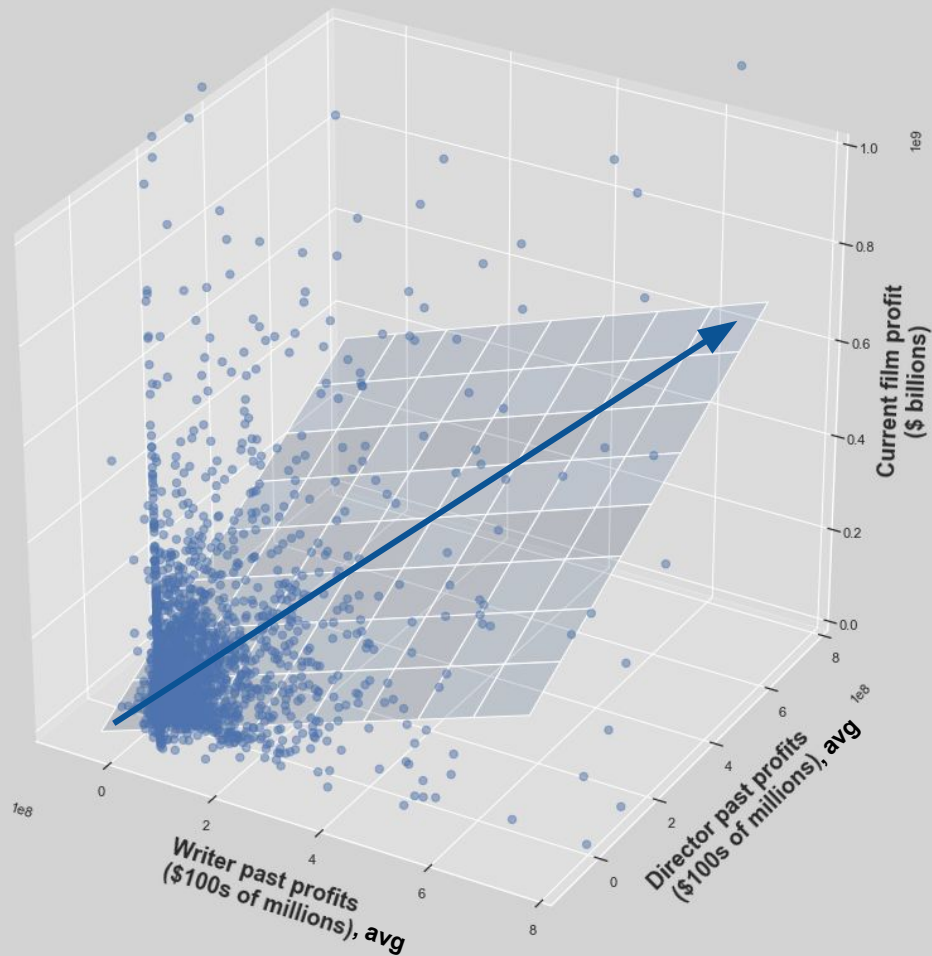
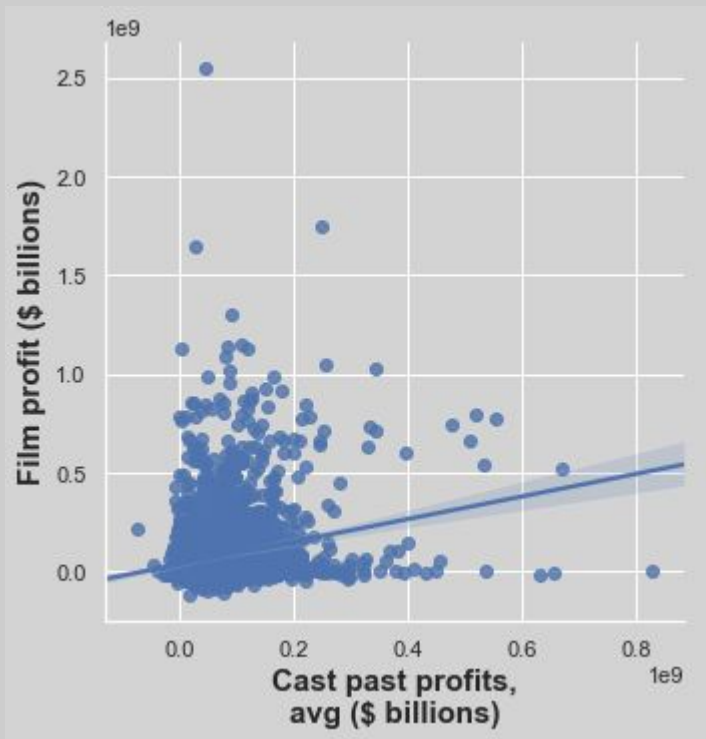
EDA

matplotlib  
seaborn

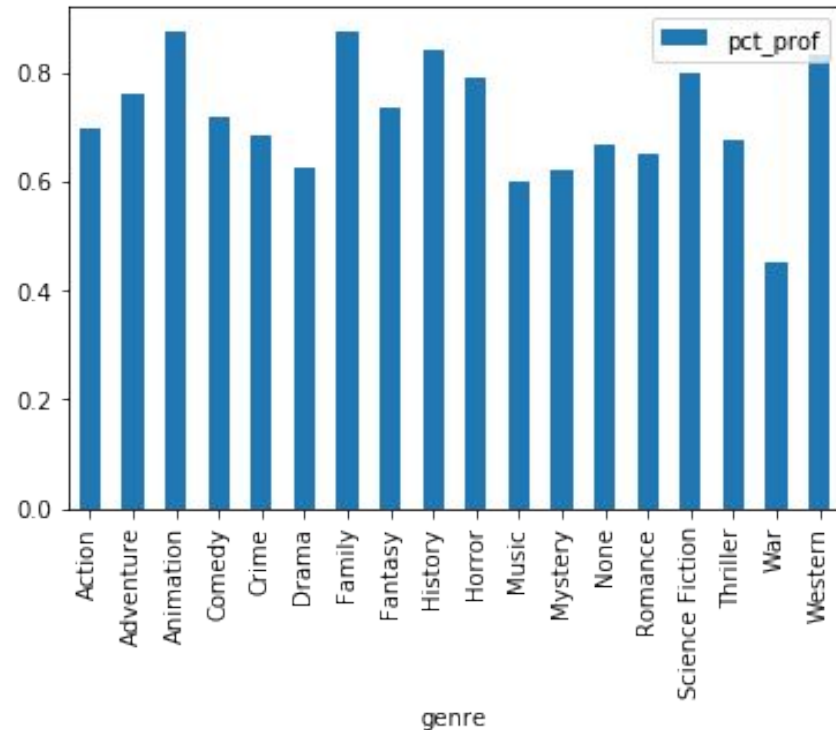
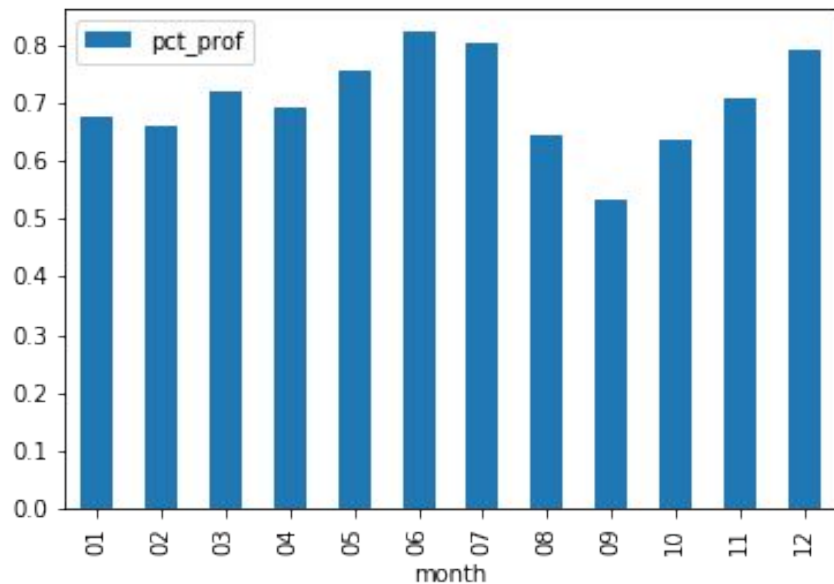
Model



# Profits vs. cast, writer, director



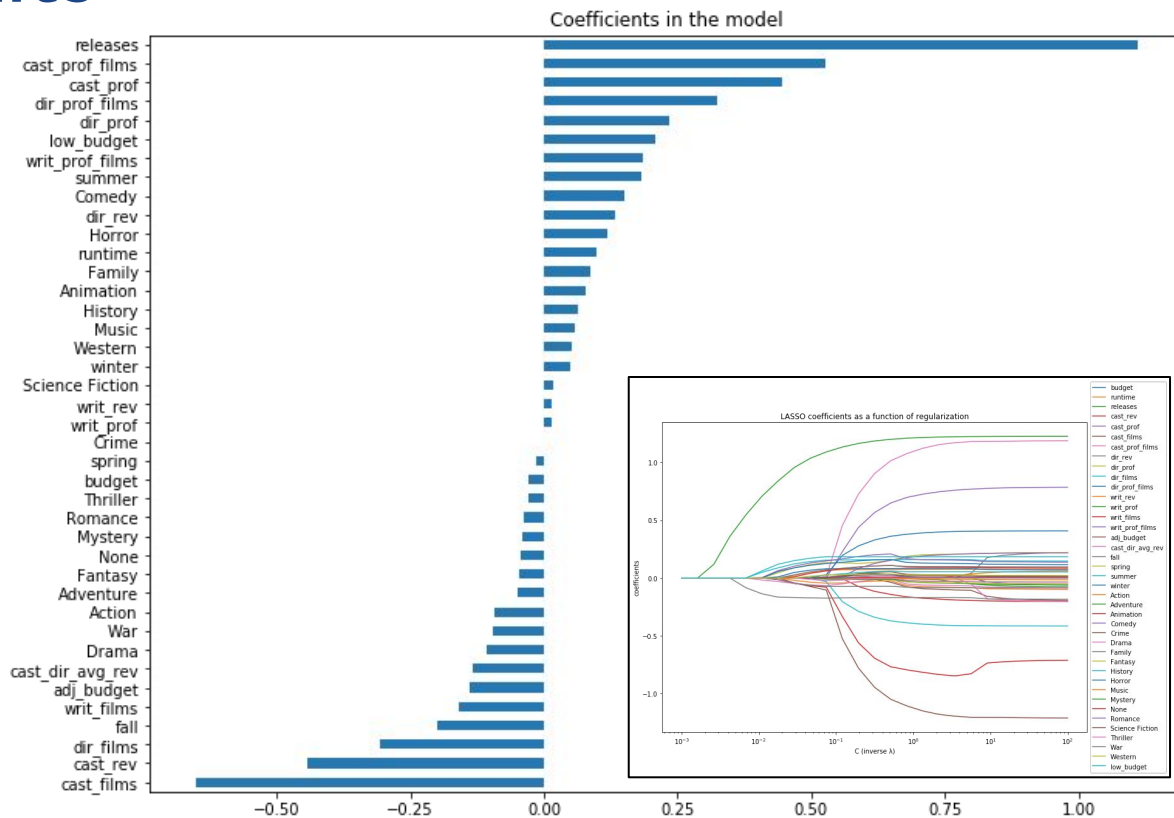
# Profitability by month, genre



# Robust coefficients

## Predictors of success:

- Int'l releases
- Cast past success
- Director past success
- “Low Budget” ( $\leq \$3\text{m}$ )
- Season
  - May-Jul best
  - Aug-Oct worst
- Genre
  - Animation/Family
  - Horror



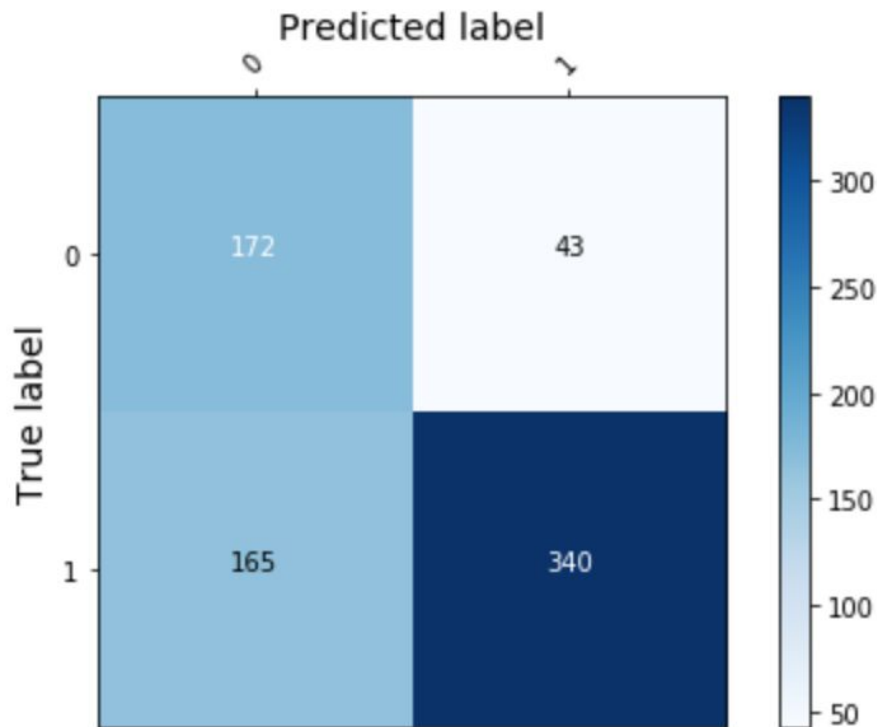
# Robust predictions

XGBoost w/ 75% decision boundary:

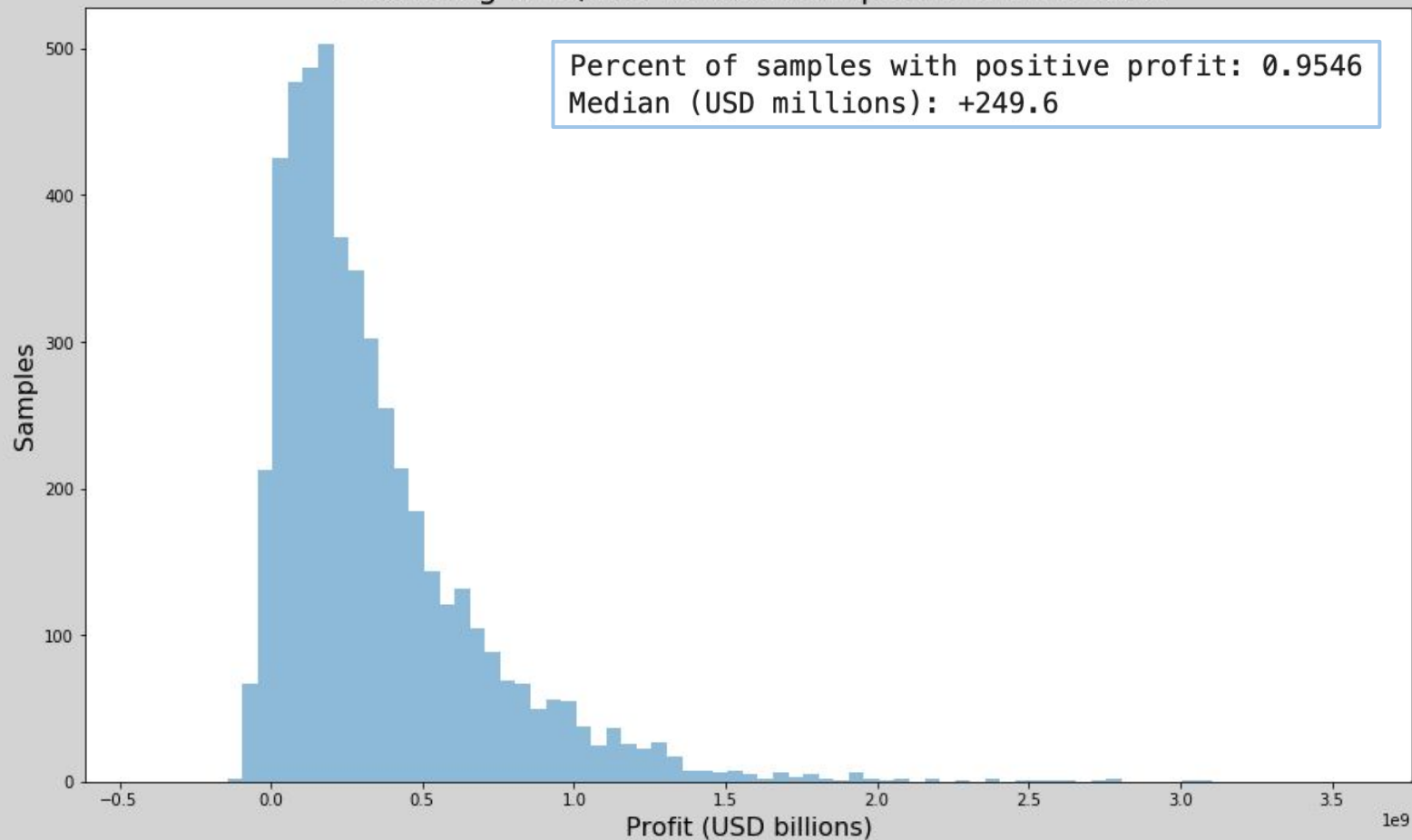
- **Accuracy: 71%**
- **Precision: 89%**
- **Recall (TPR): 67%**
- **Fallout (FPR): 20%**

```
---- Classification Report ----
              precision    recall  f1-score   support

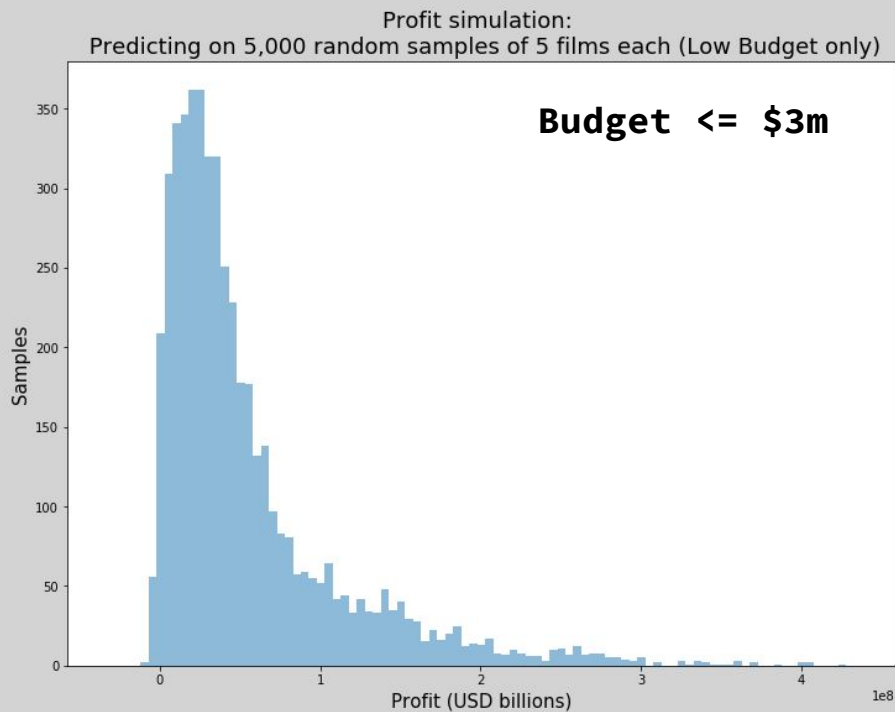
     0           0.51       0.80      0.62        215
     1           0.89       0.67      0.77        505
```



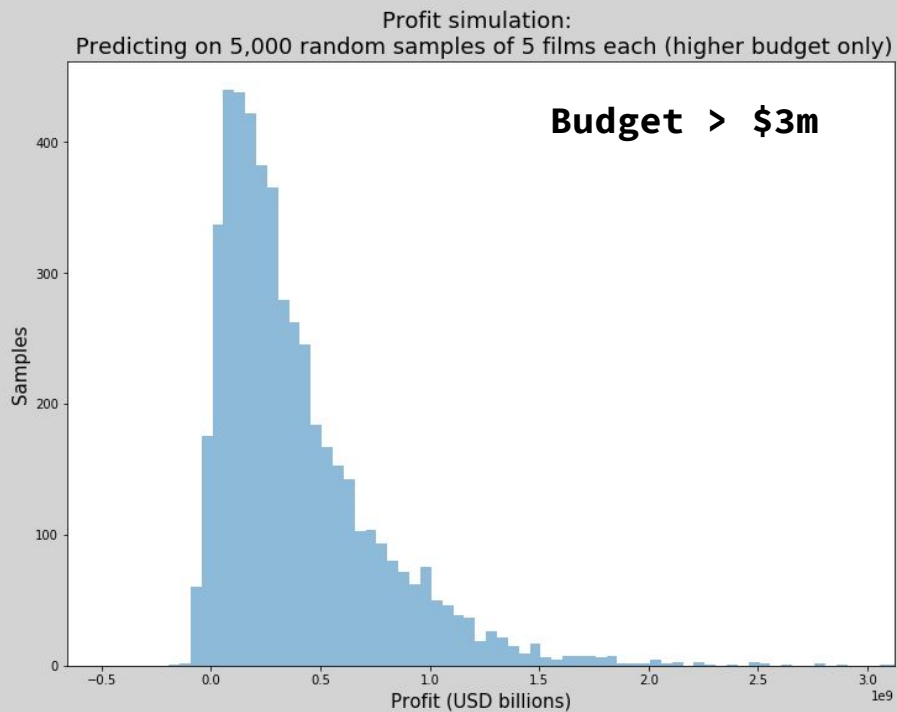
Profit simulation:  
Predicting on 5,000 random samples of 5 films each







Percent of samples with positive profit: 0.9744  
Median (USD millions): +36.1



Percent of samples with positive profit: 0.958  
Median (USD millions): +290.8

# Next steps / variables to consider

— — —

- **Collect more data**
- Story/script
- Posters
- Trailers
- Marketing strategy
- Domestic vs. foreign box office
- Ancillary revenues
- Piracy



Thank you

— — —

HOLLYWOOD

