



# *Budget vs. Box Office: ROI in the Film Industry*

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# Introduction and Contents

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# Business Understanding

## Background

Company looking to produce movies in order to diversify their revenue streams

## Goals

Provide three recommendations on which movie features are the most *cost efficient*

## Success Criteria

Define '*cost efficient*' to be measured using return on investment (ROI):

$$\text{ROI} = \frac{\text{Net profit}}{\text{Cost of investment}}$$

# Data Understanding

IMDB's film database (8 tables total):

*movie\_basics:*

- **6 features** with **146,144 entries**
- Each entry represents a different movie

*persons:*

- **5 features** with **606,648 entries**.
- Each entry represents a person who has worked in/on a movie

*principals:*

- **6 features** and **1,028,186 entries**
- Each entry represents a movie with a corresponding individual who was part of the movie

# Data Understanding

The Numbers' *budgets* dataset:

- 6 features and 5,782 entries
- Each entry represents a different movie



# Data Preparation

- Cleaned four datasets: *budgets*, *movie\_basics*, *persons*, *principals*
- Converted columns to correct data types with help of custom *get\_info()* function
- Filtered out outliers and irrelevant columns, and addressed NaN values



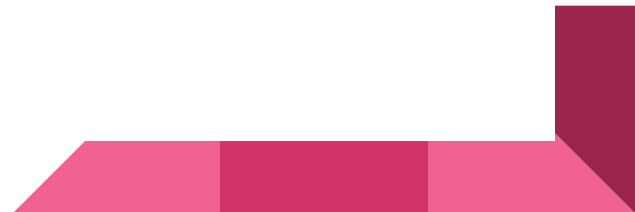
# Data Preparation

- Created three new dataframes (*top\_people\_budgets*, *top\_roi\_movie\_basics*, *budgets\_no\_outliers*)
- Focused on movies with the top 25% ROIs



# Data Analysis

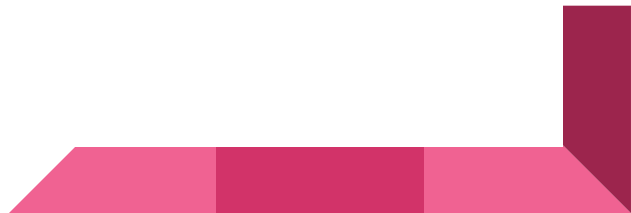
- 37% of all movies did **NOT** achieve **positive ROI**
- Typical movie:
  - **\$16 million production budget**
  - **\$26 million worldwide gross revenue**
  - **66% ROI**





# Data Analysis: Top 25% ROI Movies

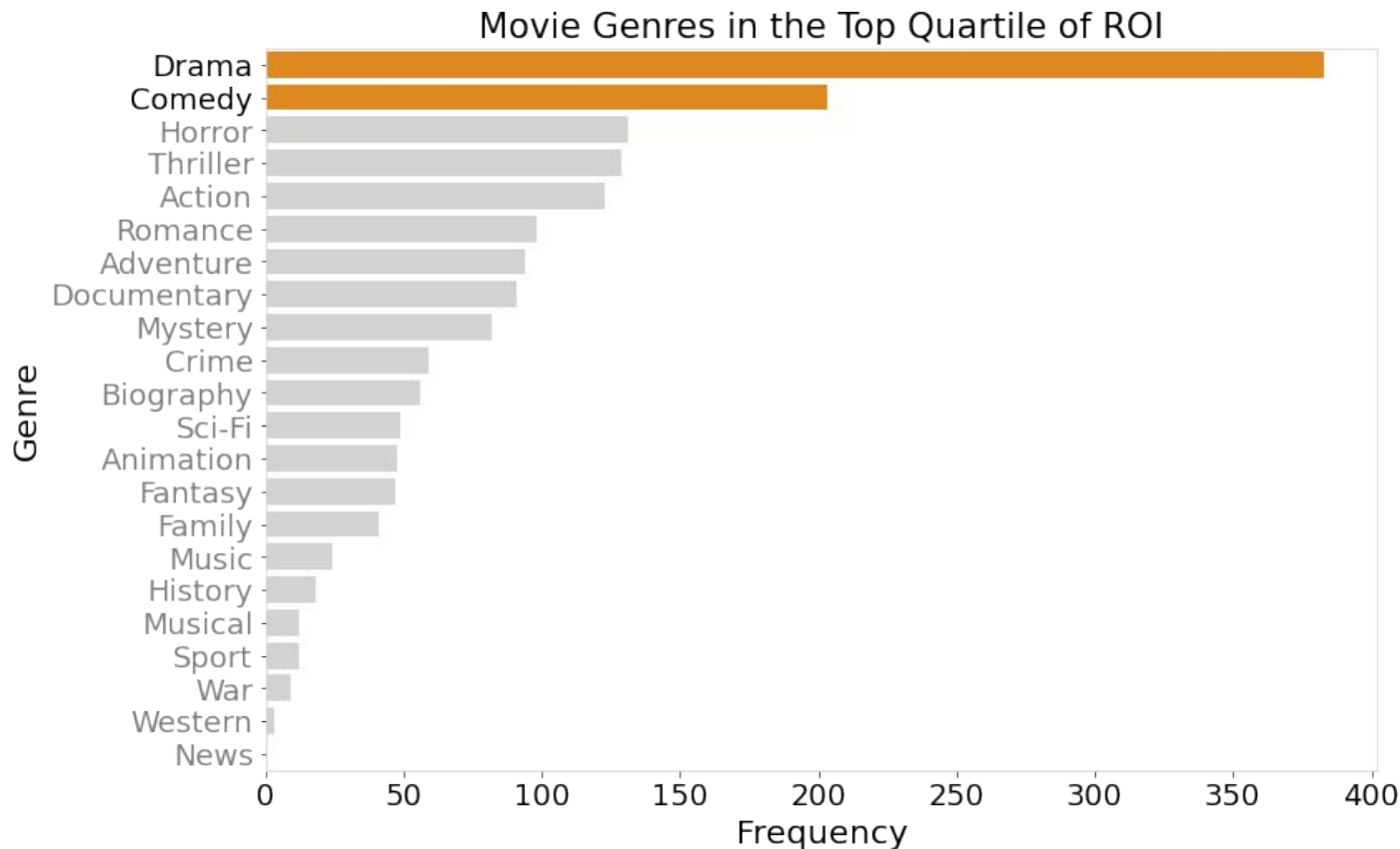
- **More than 33%** of movies classified as: **Dramas & Comedies**
- Middle 50% of movies had **runtimes** ranging **87** to **113 minutes**



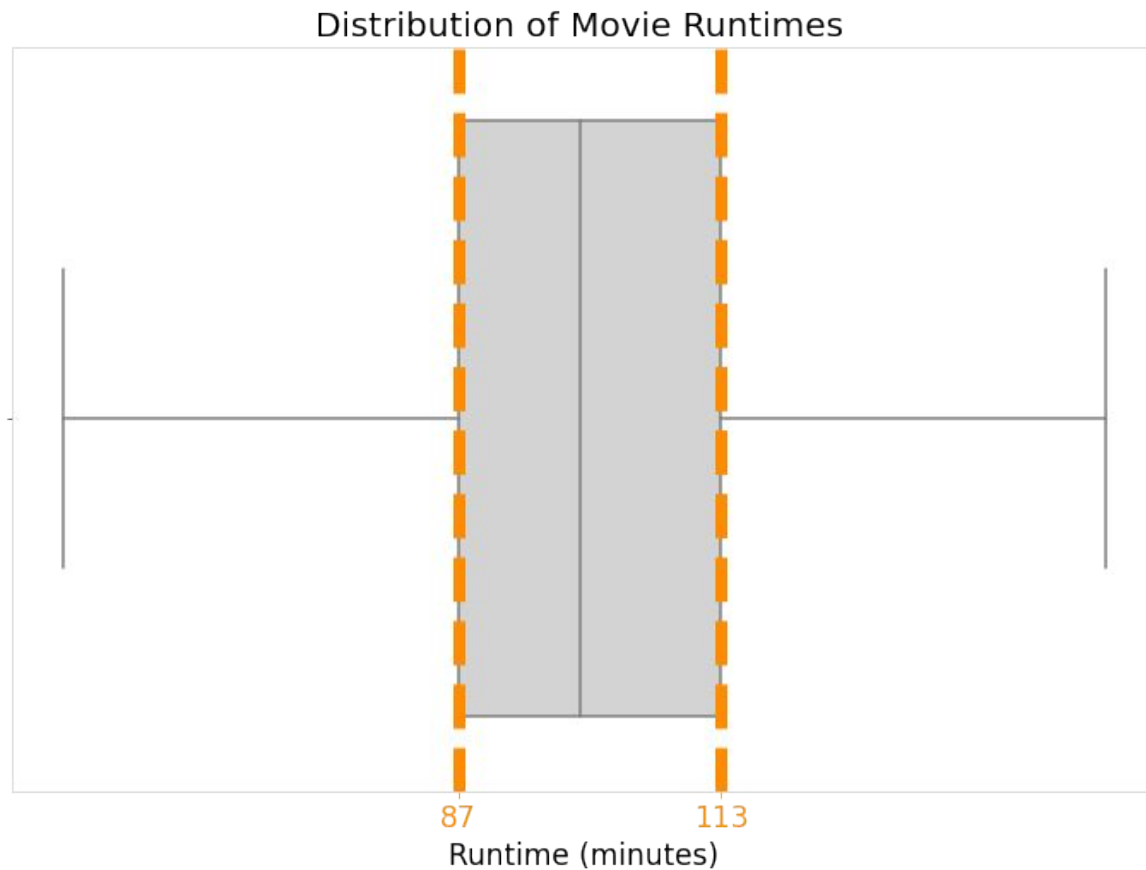
# Data Analysis: Top 25% ROI Movies

- Drama & comedy **professions** generating **highest ROI**:
    - **Composers**
      - Danny Elfman, Alexandre Desplat, Marco Beltrami, Thomas Newman, Theodore Shapiro
    - **Directors**
      - David O. Russell, Steven Spielberg, Damien Chazelle
    - **Producers**
      - Simon Kinberg, Michael De Luca, Dana Brunetti
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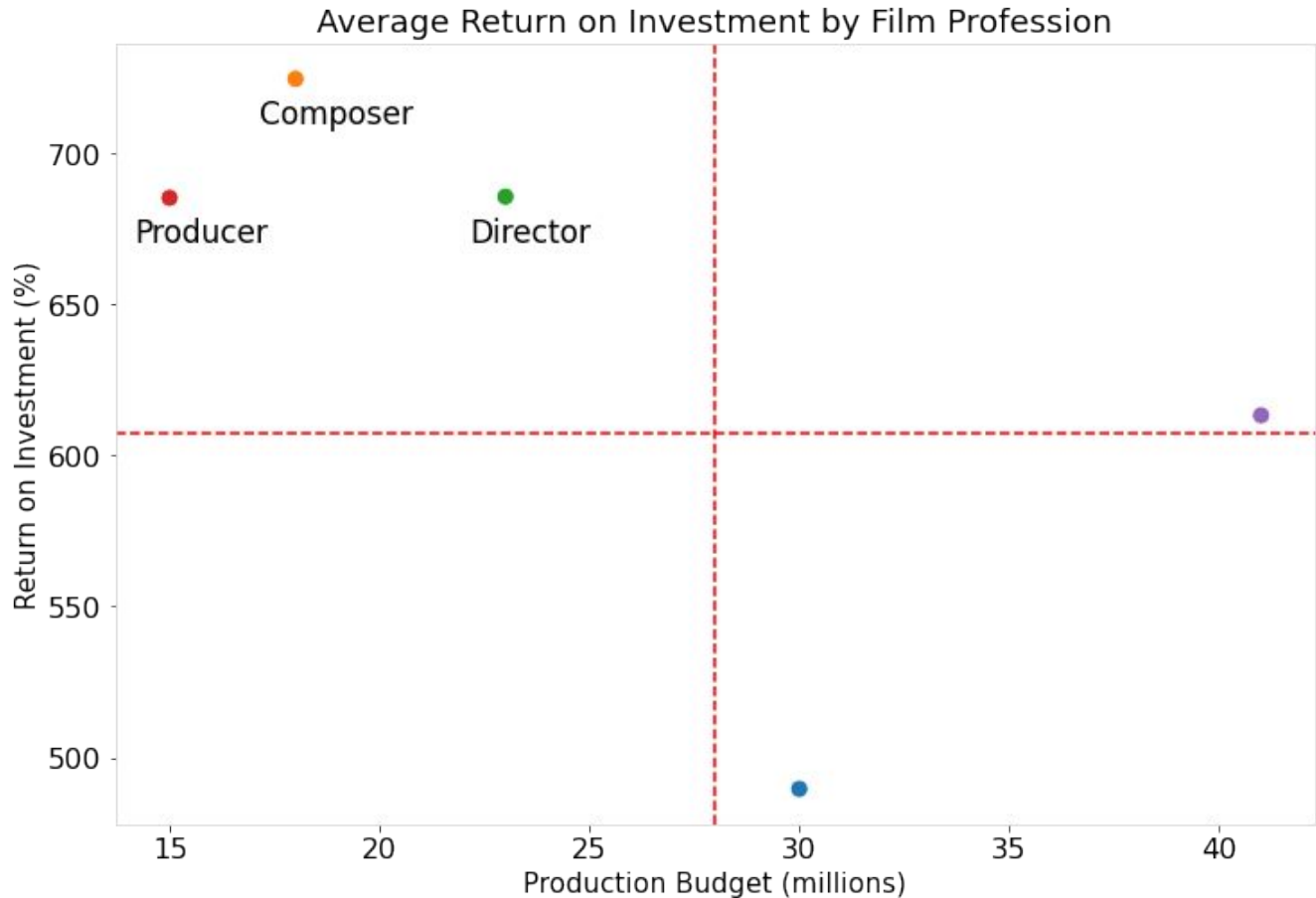
## More than 33% of movies classified as Dramas & comedies



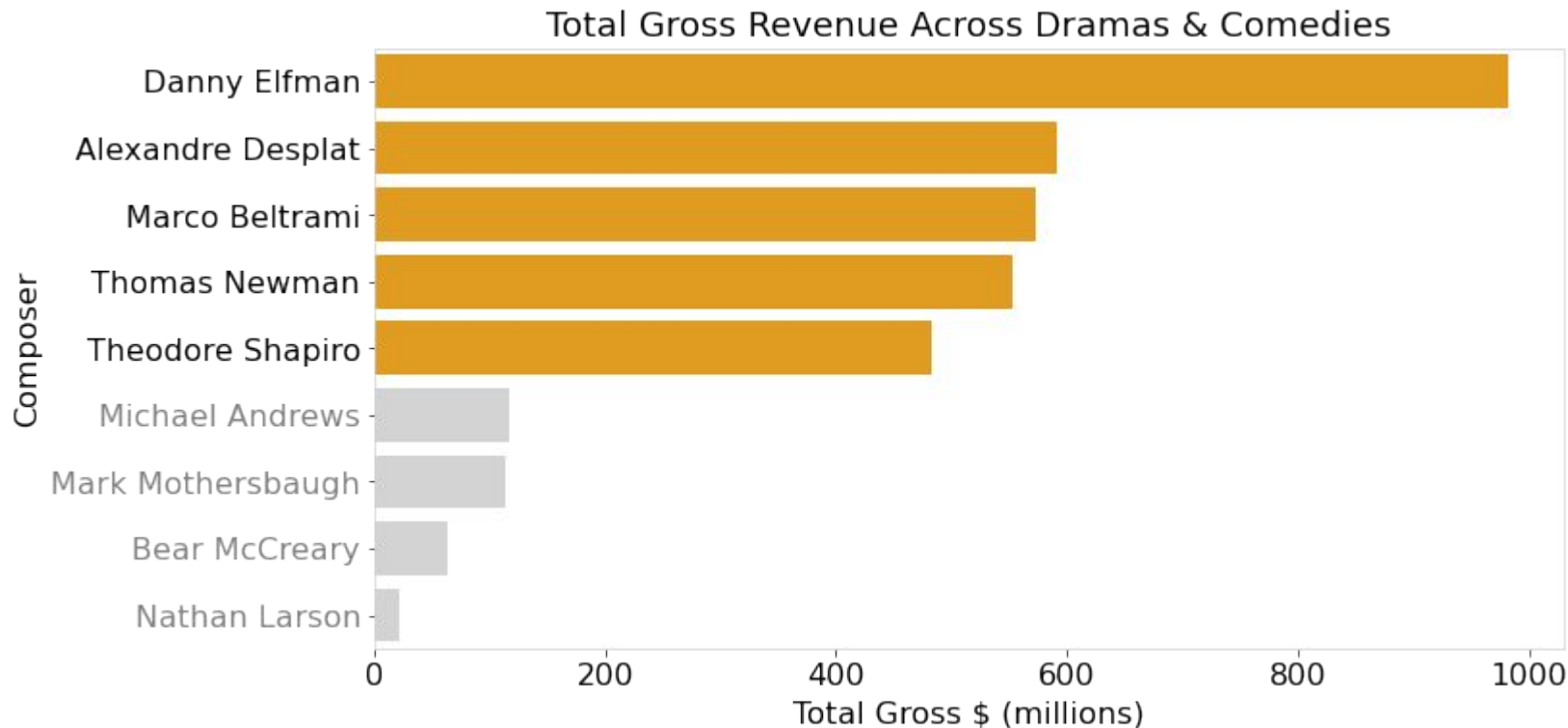
## Runtimes between 87-113 minutes for the middle 50% of movies



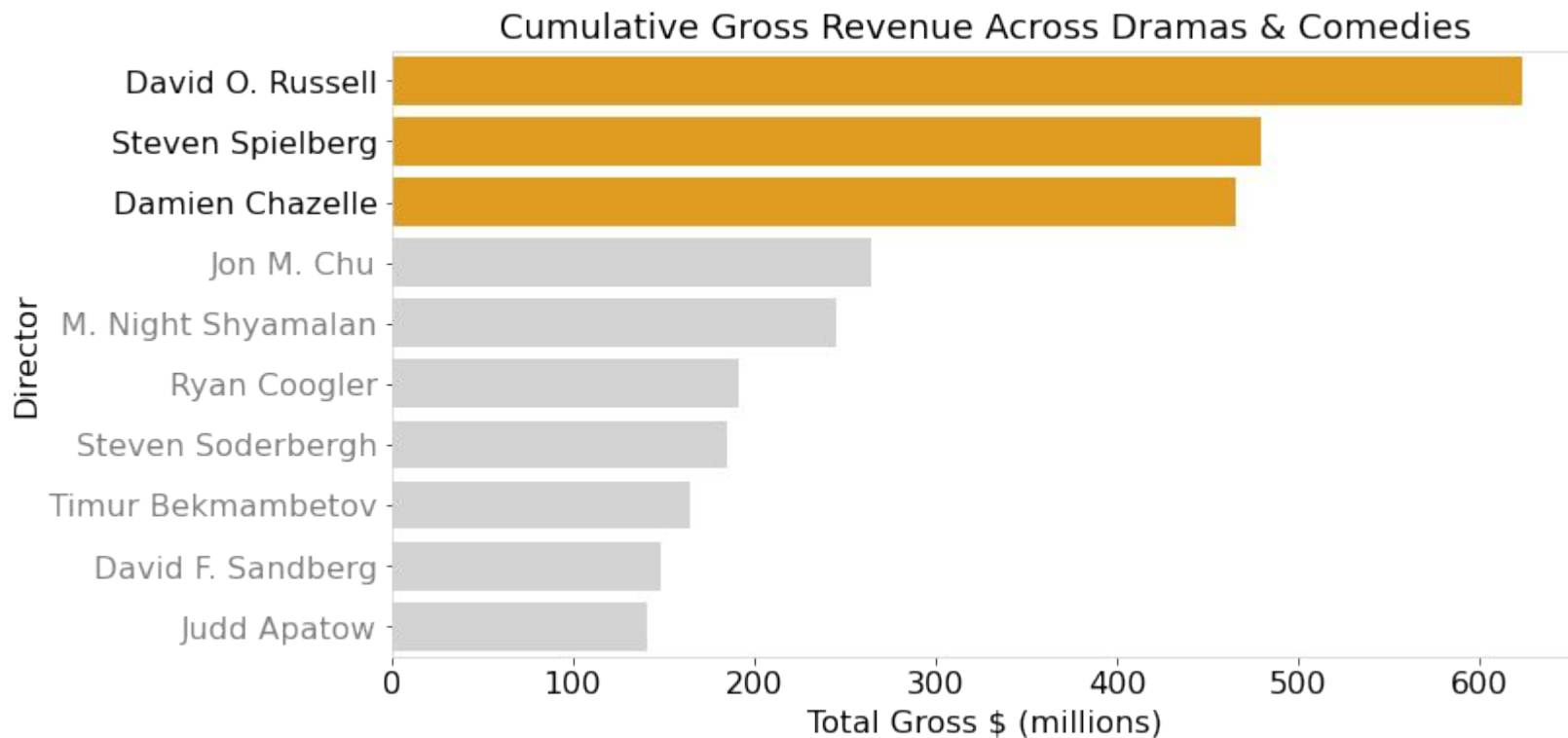
Producers, writers, & directors generate the **highest ROI**



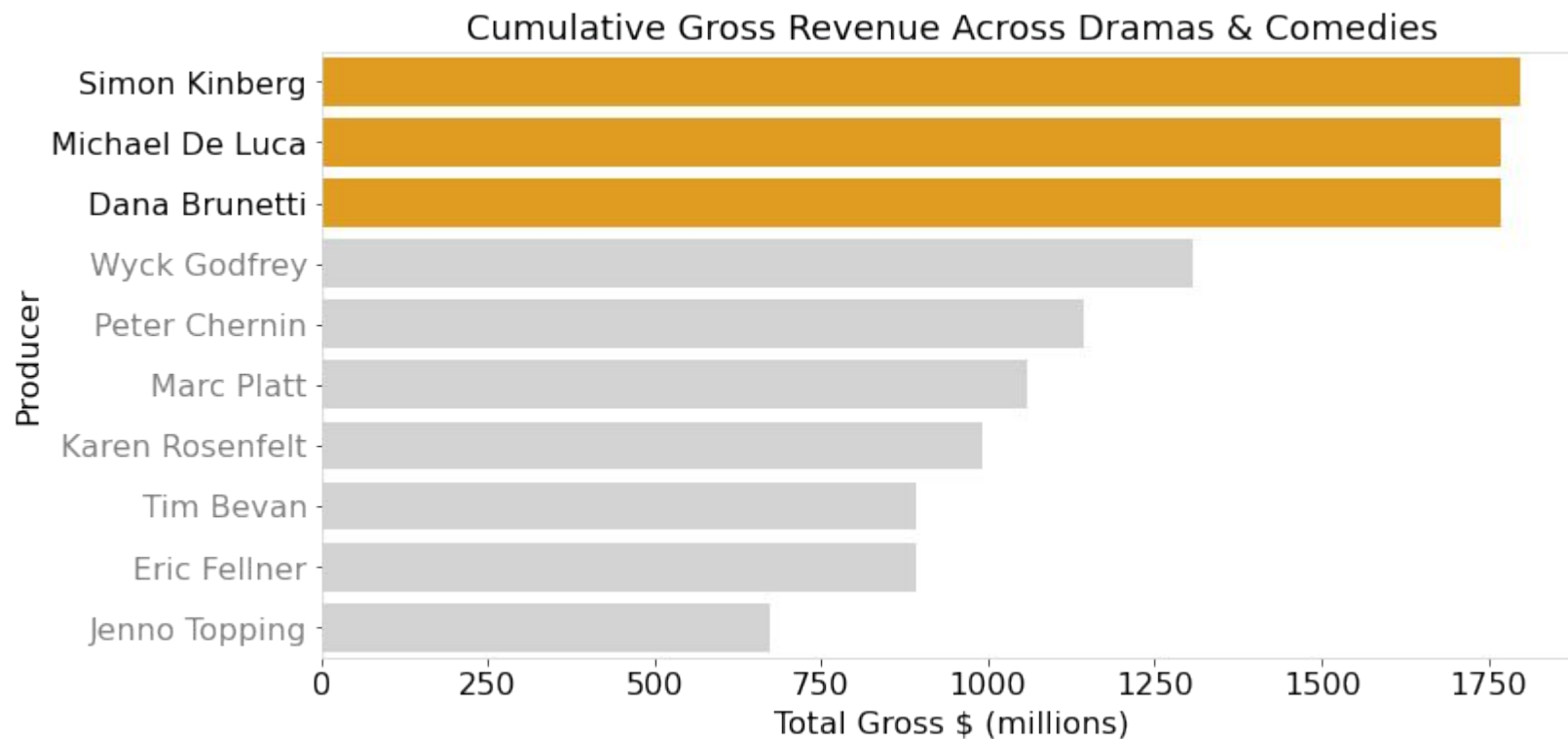
Drama & comedy **composers**: Danny Elfman, Alexandre Desplat, Marco Beltrami, Thomas Newman, Theodore Shapiro



## Drama & comedy **directors**: David O. Russell, Steven Spielberg, Damien Chazelle



## Drama & comedy **producers**: Simon Kinberg, Michael De Luca, Dana Brunetti





# Limitations

- Limited amount of budget data = Less data for ROI analysis



# Recommendations

## Recommendation 1: Genre

Focus on creating:

- **Dramas**
- **Comedies**

## Recommendation 2: Runtime

Ensure movies are  
between **87** and **113**  
**minutes long**

## Recommendation 3: Hiring

Hire high-quality  
**composers, directors,**  
& **producers**  
specializing in **drama**  
& **comedy**

# Next Steps

1. Moving forward I am interested in:
  - a. Gathering budget data on a wider number of movies
  - b. Performing regression analysis:
    - i. Which factors most strongly correlate with movie ROI?



# Thank you!



Github Repository:

[https://github.com/ckucewicz/movie\\_analysis\\_project](https://github.com/ckucewicz/movie_analysis_project)

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