



# The movie industry and the influential factors of IMDb movie ratings

**Mafalda Fragoso**

Data Analytics Bootcamp  
27/03/2020

# Table of contents

---

**1. Goals of the project**

**2. Methodology**

**3. Main findings**

The IMDb logo is displayed in a yellow rounded rectangle. It consists of the letters "IMDb" in a bold, black, sans-serif font. The background of the slide is a grid of various movie posters and actor portraits, including titles like "Leon", "Fringe", "The Education of a Young Woman", and "21 Jump Street".

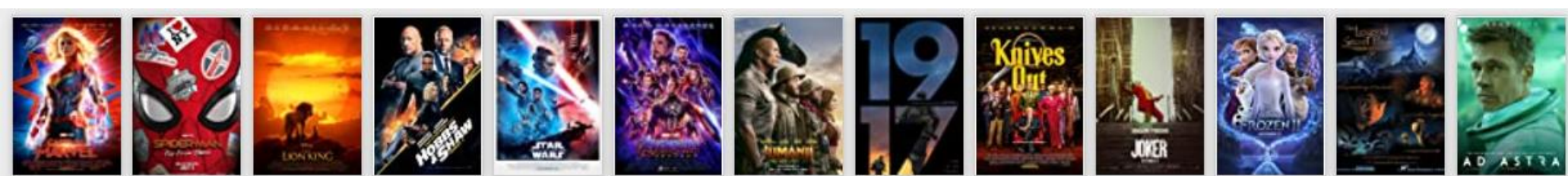
**IMDb**

# Goals of the project

---

## Research purpose

1. Exploratory analysis on the movie industry (2019)
2. Understand the influential factors of IMDb movie ratings:
  - Is the movie rating variable correlated with the duration of the movie, genre of the movie, number of votes and revenue?




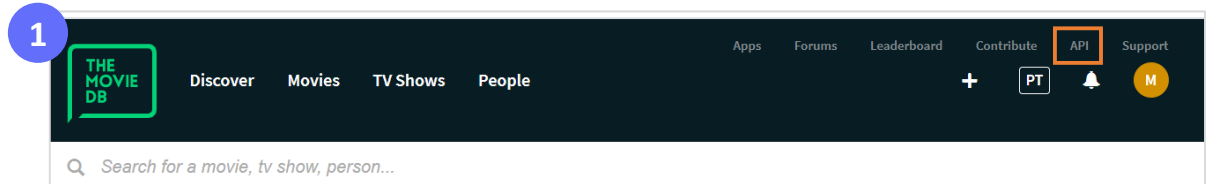
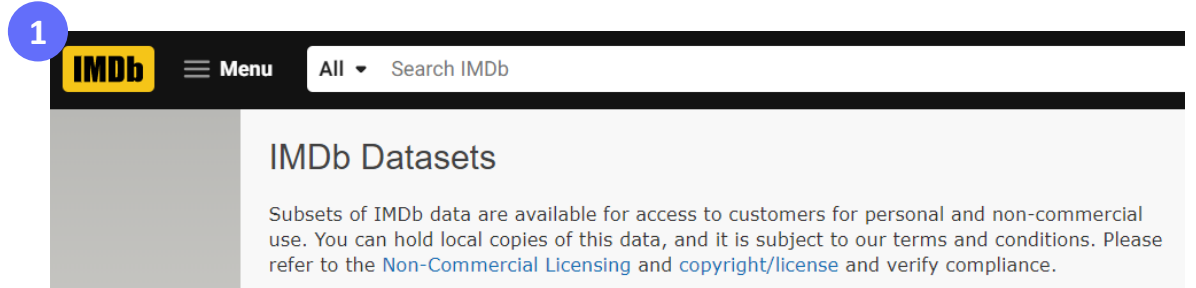
# Methodology

## 1. Data gathering

- IMDb » open datasets
- The Movie Database » API

## 2. Data cleaning

- Joined several tables from the IMDb open datasets with the information needed for the study
- Joined the final IMDb dataset with the data gathered from the API of The Movie Database (revenue and budget variables)
- Data cleaning included checking for missing values, number of duplicates, low variance and data types
- Sample of 1640 movies after cleaning the data

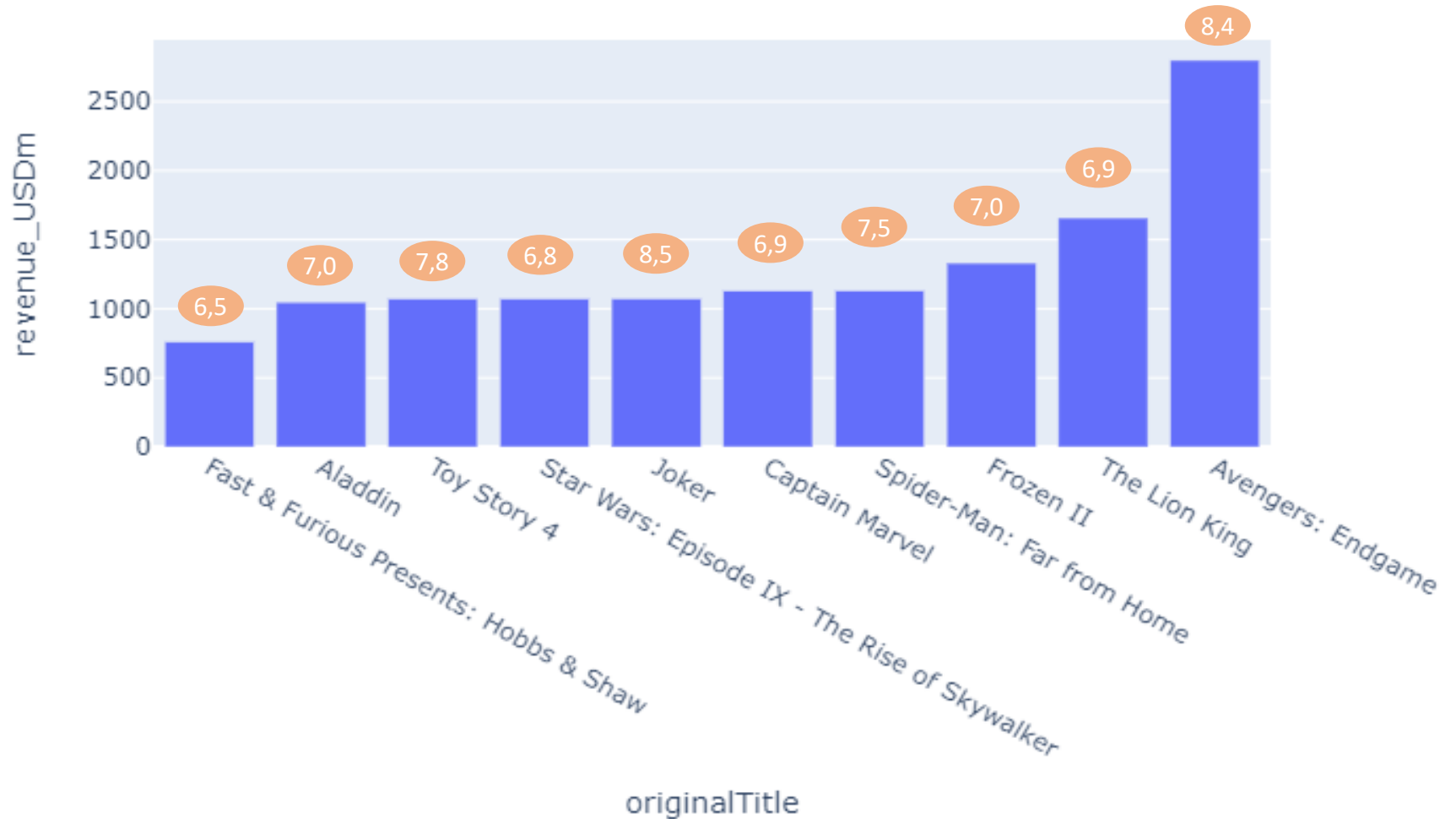
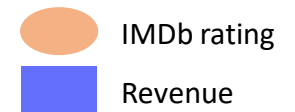


	id	tconst	revenue_USDm	budget_USDm	titleType	originalTitle	startYear	runtimeMinutes	averageRating	numVotes	Audience	Genre
0	419704.0	tt2935510	127.175922	87.5	movie	Ad Astra	2019	123.0	6.6	152221	not adult only	Adventure
1	181812.0	tt2527338	1073.604458	250.0	movie	Star Wars: Episode IX - The Rise of Skywalker	2019	142.0	6.8	277640	not adult only	Action
2	512200.0	tt7975244	310.830000	125.0	movie	Jumanji: The Next Level	2019	123.0	6.7	111270	not adult only	Action
3	330457.0	tt4520988	1330.764959	33.0	movie	Frozen II	2019	103.0	7.0	88950	not adult only	Adventure
4	475557.0	tt7286456	1074.151311	55.0	movie	Joker	2019	122.0	8.5	721212	not adult only	Crime

# Main findings

## The top movies released in 2019 based on revenue

(revenue in UDS millions - Cumulative Worldwide Gross)



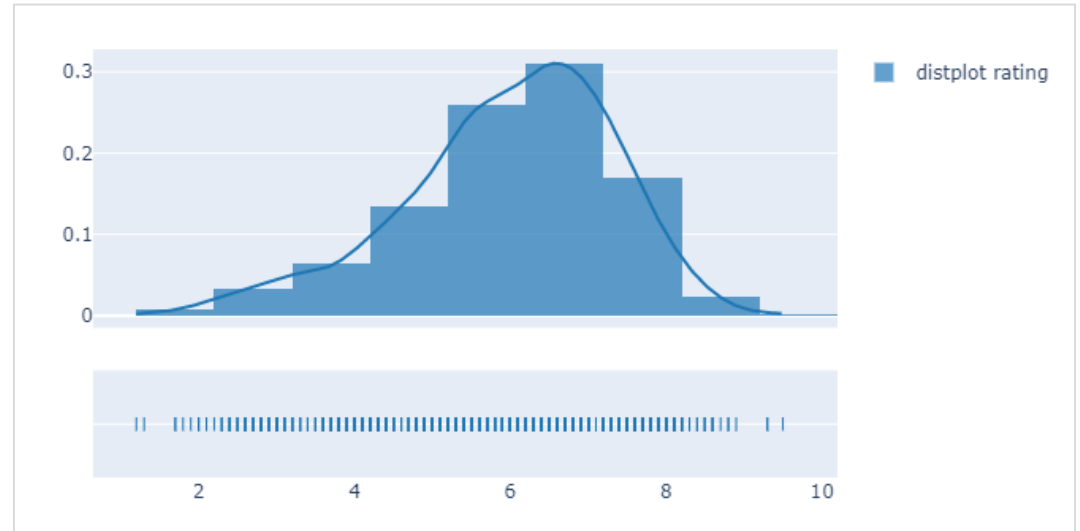
# Main findings

---

## Rating distribution

The average rating variable follows a normal distribution with a mean of 5,97

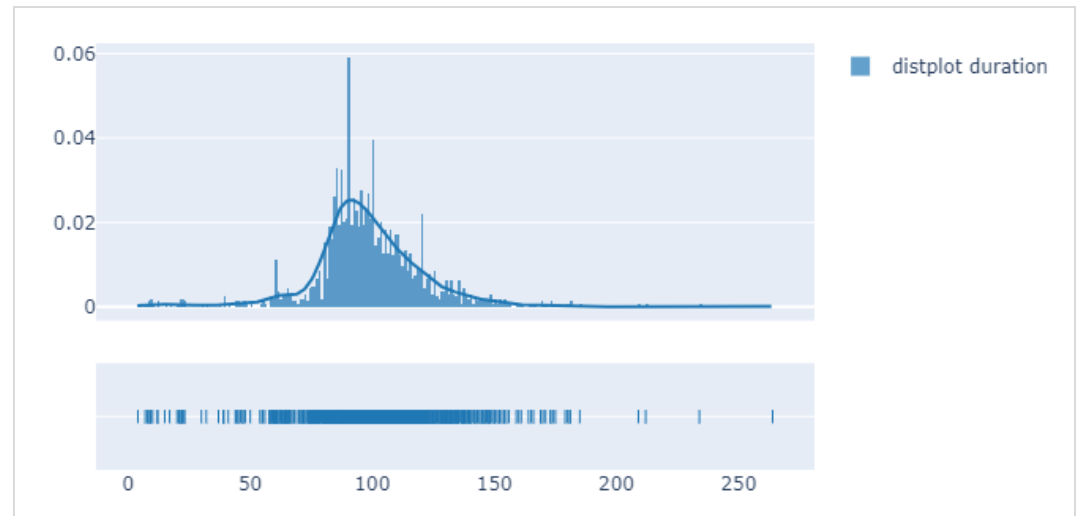
(rating scale from 1 to 10)



---

## Duration of the movie distribution

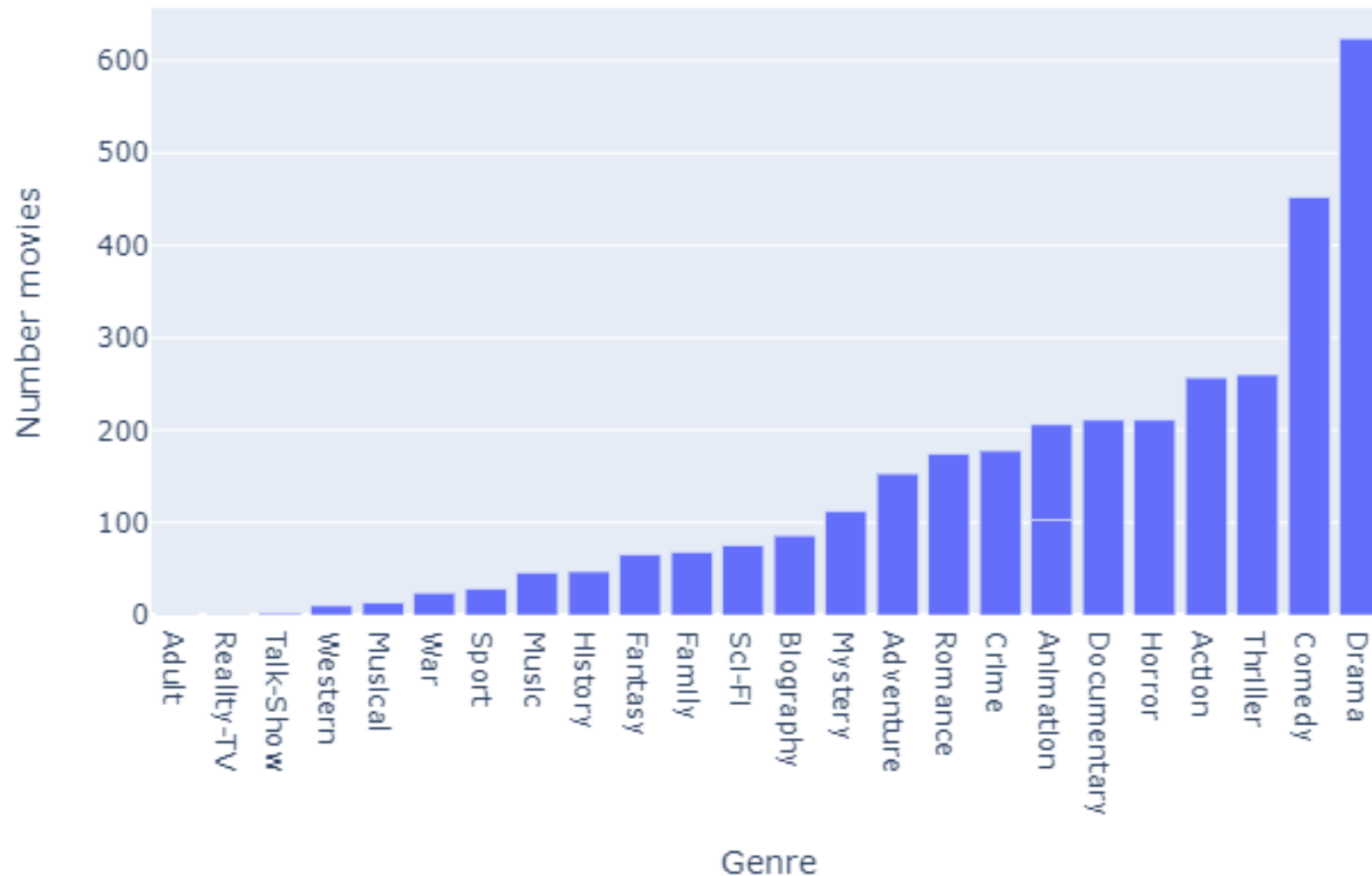
The duration of the movie variable follows a normal distribution with a mean of 98 minutes



# Main findings

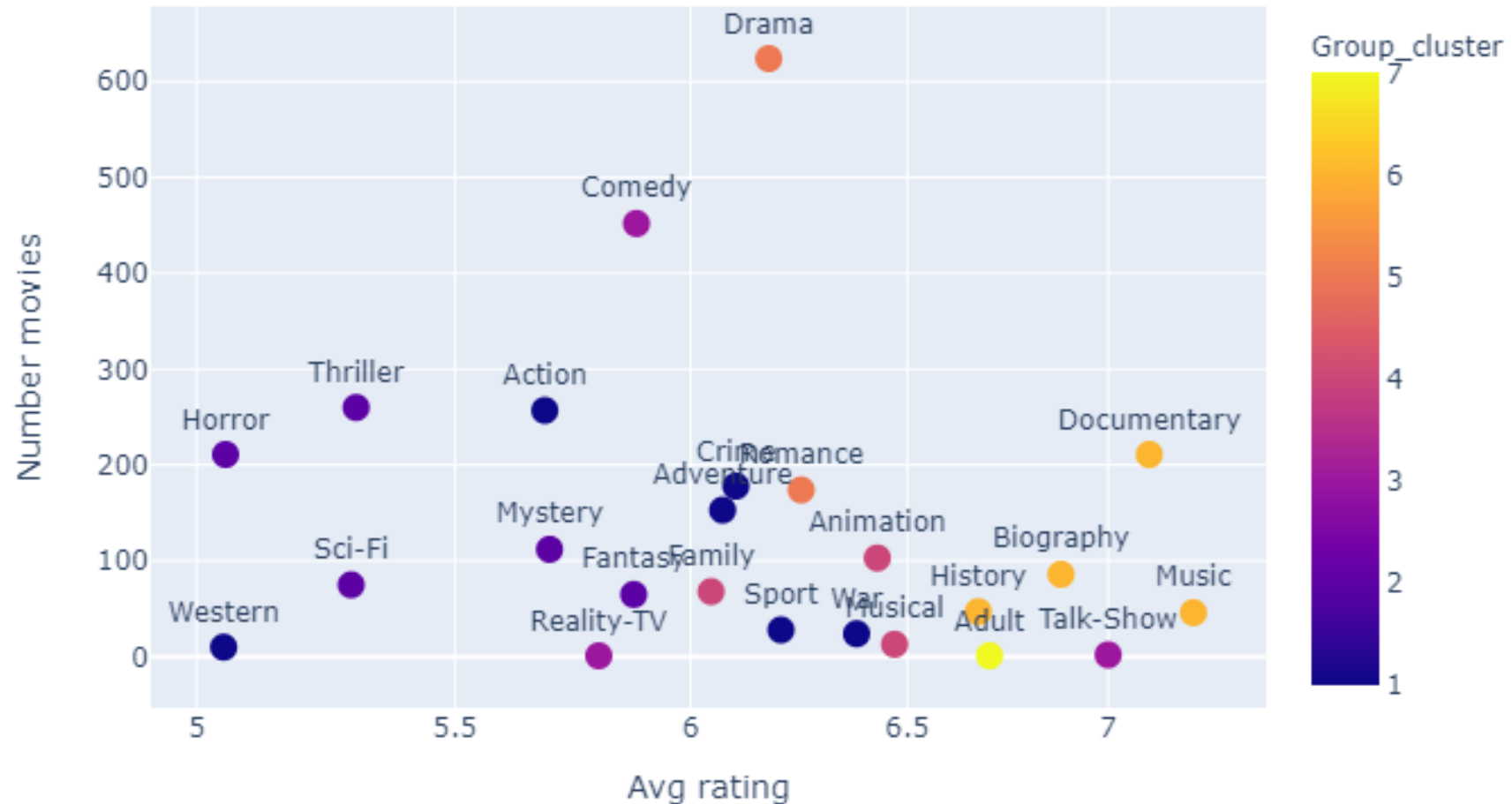
---

**The most representative genres in my sample are Drama, Comedy, Thriller and Action**  
(note: each movie can have until 3 genres classification)



# Main findings

## Distribution of average rating vs number of movies per gender



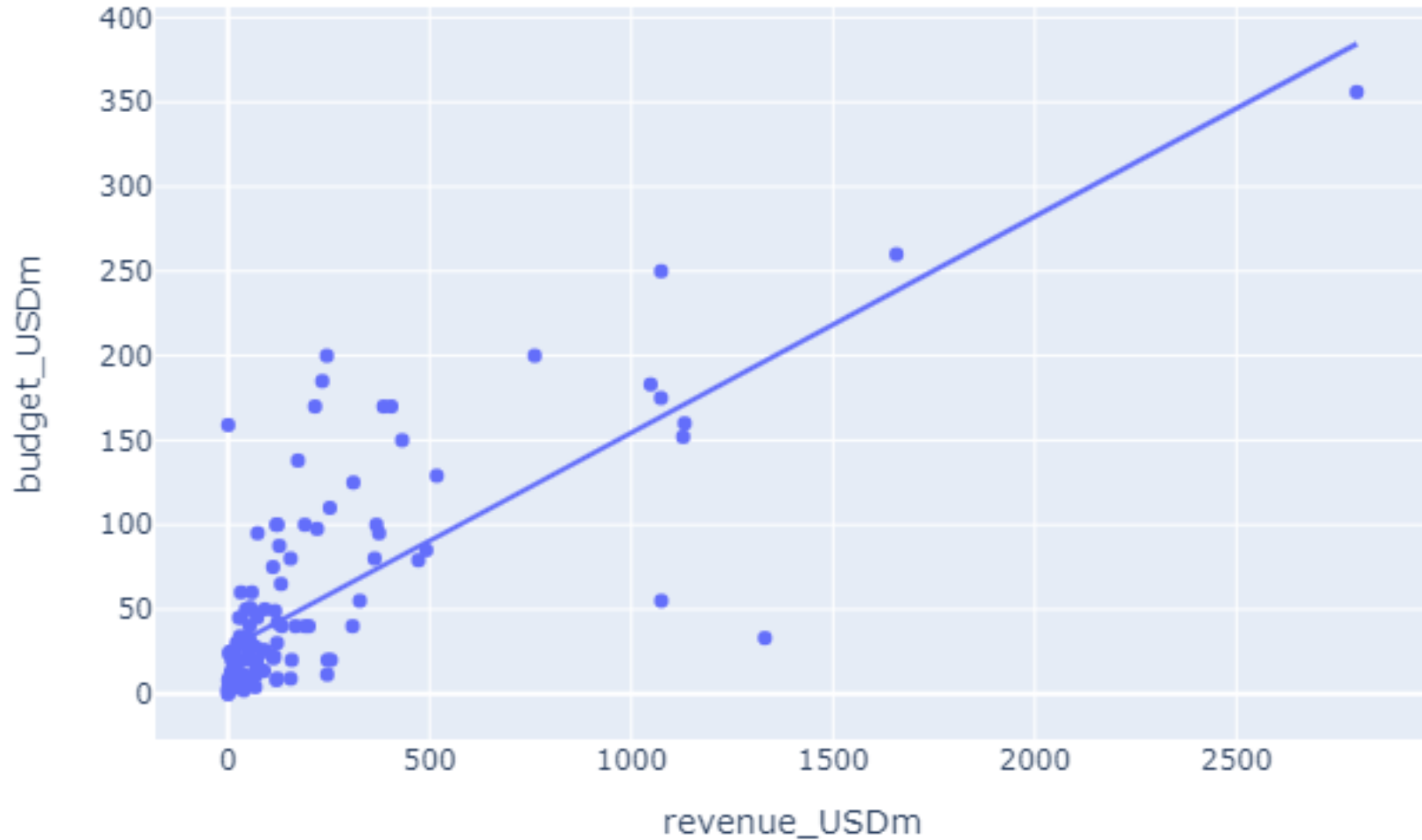
- There are less movies classified as technical/informative (documentary, Biography, History) but they are more likely to have higher ratings
- Movies in the cluster of action, horror, sci.fi are more likely to have lower ratings



# Main findings

---

**Strong positive linear relationship between revenue and budget**  
(correlation coefficient **0.76**)

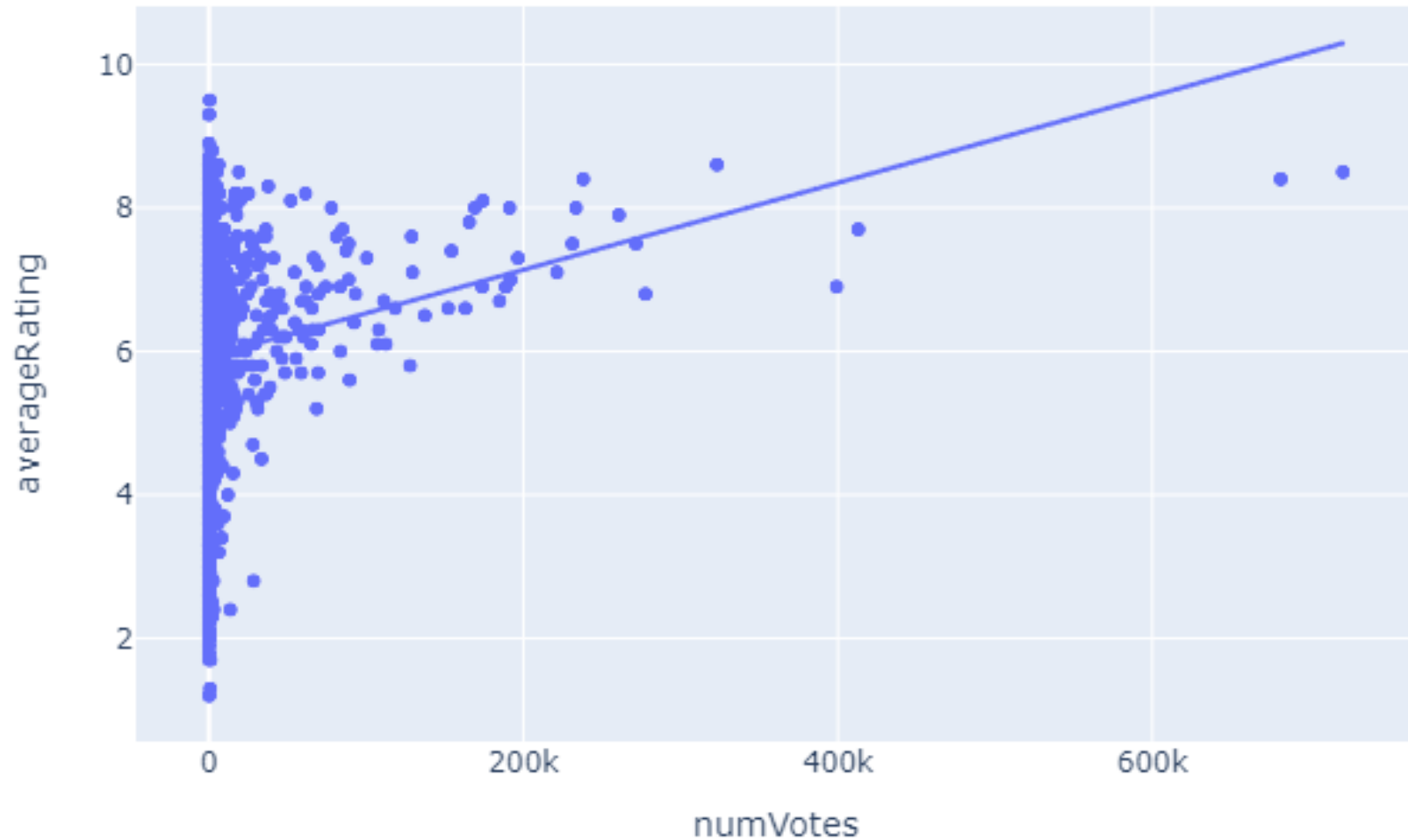


More investment results in more payback

# Main findings

---

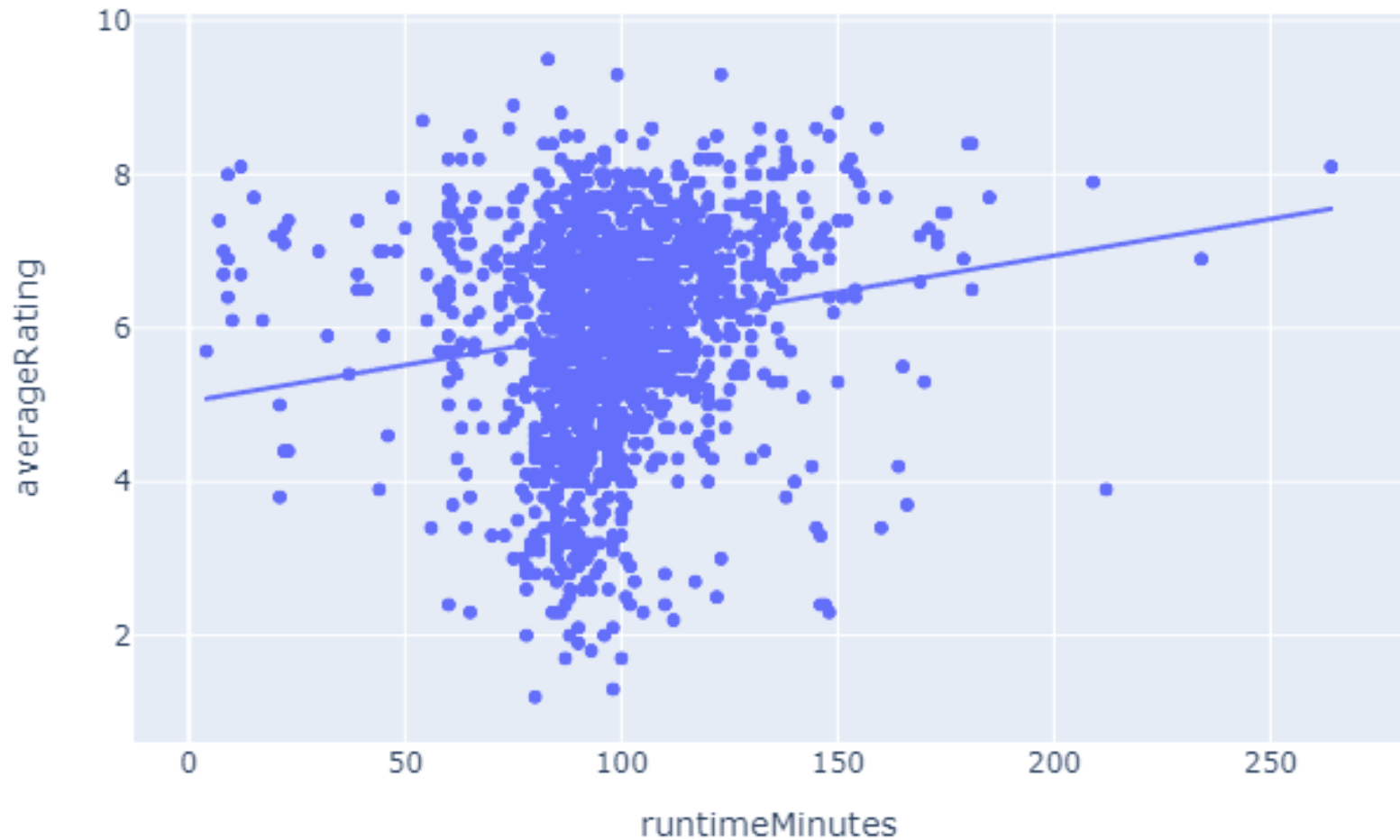
**No relationship found between average rating and number of votes**  
(correlation coefficient **0.17**)



# Main findings

---

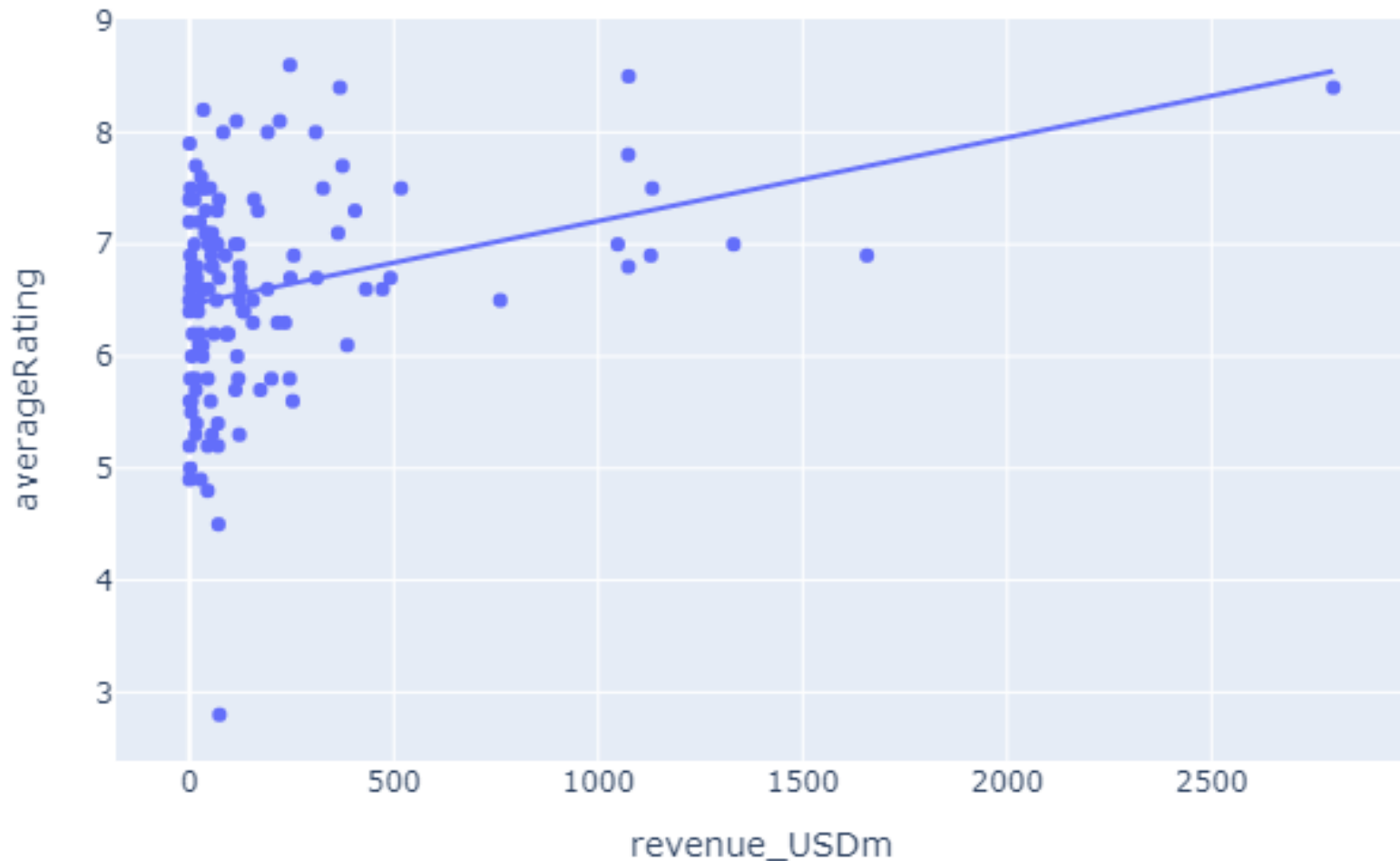
**No relationship found between average rating and duration of the movie**  
(correlation coefficient **0.16**)



# Main findings

---

**No relationship found between average rating and revenue of the movie**  
(correlation coefficient **0.3**)

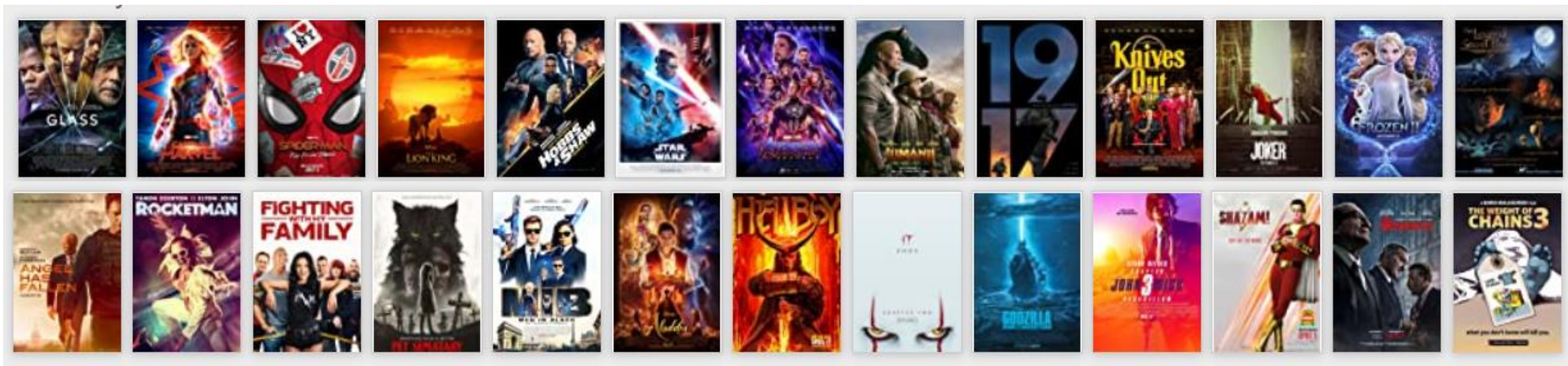


Higher rating doesn't necessarily correlate with higher profits for a movie

# Conclusions

---

- The movie rating variable is not correlated with the duration of the movie, number of votes and revenue
- Technical/informative type of movies (documentary, Biography, History) are more likely to have higher ratings while movies in the cluster of action, horror, sci.fi are more likely to have lower ratings
- Strong positive linear relationship between revenue and budget meaning that more investment results in more payback





# Thank you.

**Mafalda Fragoso**

Data Analytics Bootcamp  
27/03/2020

# Appendix

# OLS Regression Results

---

## OLS Regression Results

```
=====
Dep. Variable:          averageRating    R-squared:                0.045
Model:                  OLS              Adj. R-squared:          0.044
Method:                 Least Squares    F-statistic:             38.70
Date:                   Fri, 27 Mar 2020  Prob (F-statistic):      3.80e-17
Time:                   05:29:44         Log-Likelihood:          -2787.8
No. Observations:      1640             AIC:                    5582.
Df Residuals:          1637             BIC:                    5598.
Df Model:               2
Covariance Type:       nonrobust
=====

               coef      std err          t      P>|t|      [0.025      0.975]
-----
const          5.2056      0.147      35.436      0.000       4.917       5.494
runtimeMinutes  0.0074      0.001       5.025      0.000       0.005       0.010
numVotes       4.986e-06  8.66e-07      5.757      0.000      3.29e-06      6.68e-06
=====

Omnibus:            89.585    Durbin-Watson:           1.922
Prob(Omnibus):      0.000    Jarque-Bera (JB):        103.296
Skew:               -0.594    Prob(JB):                3.71e-23
Kurtosis:           3.315    Cond. No.                1.80e+05
=====
```



# OLS Regression Results

---

```
=====
                        OLS Regression Results
=====
Dep. Variable:          averageRating    R-squared:                0.118
Model:                  OLS              Adj. R-squared:           0.103
Method:                 Least Squares    F-statistic:              8.215
Date:                   Fri, 27 Mar 2020  Prob (F-statistic):      0.000448
Time:                   05:31:42         Log-Likelihood:           -160.77
No. Observations:       126             AIC:                     327.5
Df Residuals:           123             BIC:                     336.0
Df Model:                2
Covariance Type:        nonrobust
=====

               coef      std err          t      P>|t|      [0.025      0.975]
-----
const          6.5573      0.102     64.552     0.000      6.356      6.758
revenue_USDm    0.0012      0.000      3.768     0.000      0.001      0.002
budget_USDm    -0.0035      0.002     -1.876     0.063     -0.007      0.000
=====

Omnibus:                 10.413    Durbin-Watson:           1.911
Prob(Omnibus):            0.005    Jarque-Bera (JB):        12.643
Skew:                    -0.502    Prob(JB):                0.00180
Kurtosis:                 4.184    Cond. No.                 562.
=====
```

# Revenue and budget distribution

---

