

A silhouette of a cameraman wearing a cap, operating a professional video camera mounted on a dolly. The dolly is on a track, and a long boom arm extends from the right side. The background is a warm, orange-hued sunset or sunrise sky. A solid red square is located in the top-left corner of the image.

# Effect of Rating on Movie Production Costs Rate of Return

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# Business Case

Whether producing an indie film, or next summer's blockbuster, it helps to know what factors predict a favorable rate of return on investment.

This project is designed to help a production company shape their movie project portfolio for the best return on investment.

# Inspiration



De Vany, A., & Walls, W. (2002). **Does Hollywood Make Too Many R-Rated Movies? Risk, Stochastic Dominance, and the Illusion of Expectation.** *The Journal of Business*, 75(3), 425-451. doi:1. Retrieved from <http://www.jstor.org/stable/10.1086/339890> doi:1

- Using movie data from 1982 – 1996, De Vany & Walls showed by stochastic dominance, that movie studios could maximize their rate of return by making more PG and PG-13 movies, thereby reducing the proportion of R rated films.
- This study in this paper, uses multiple linear regression to identify which ratings generate the best rate of return for the time period from 2000 to 2015, while taking into consideration several other factors of interest.

# Methodology: Data Used

## Target:

Rate of return =  $\frac{\text{Worldwide Gross}}{\text{Total Production Budget}}$

## Example:

\$30 million return on \$10 million budget = rate of return = 3

## Data Sources:

the-record.com – financial data

IMDB database

awardsdatabase.oscars.org – Academy Awards database

## Predictors:

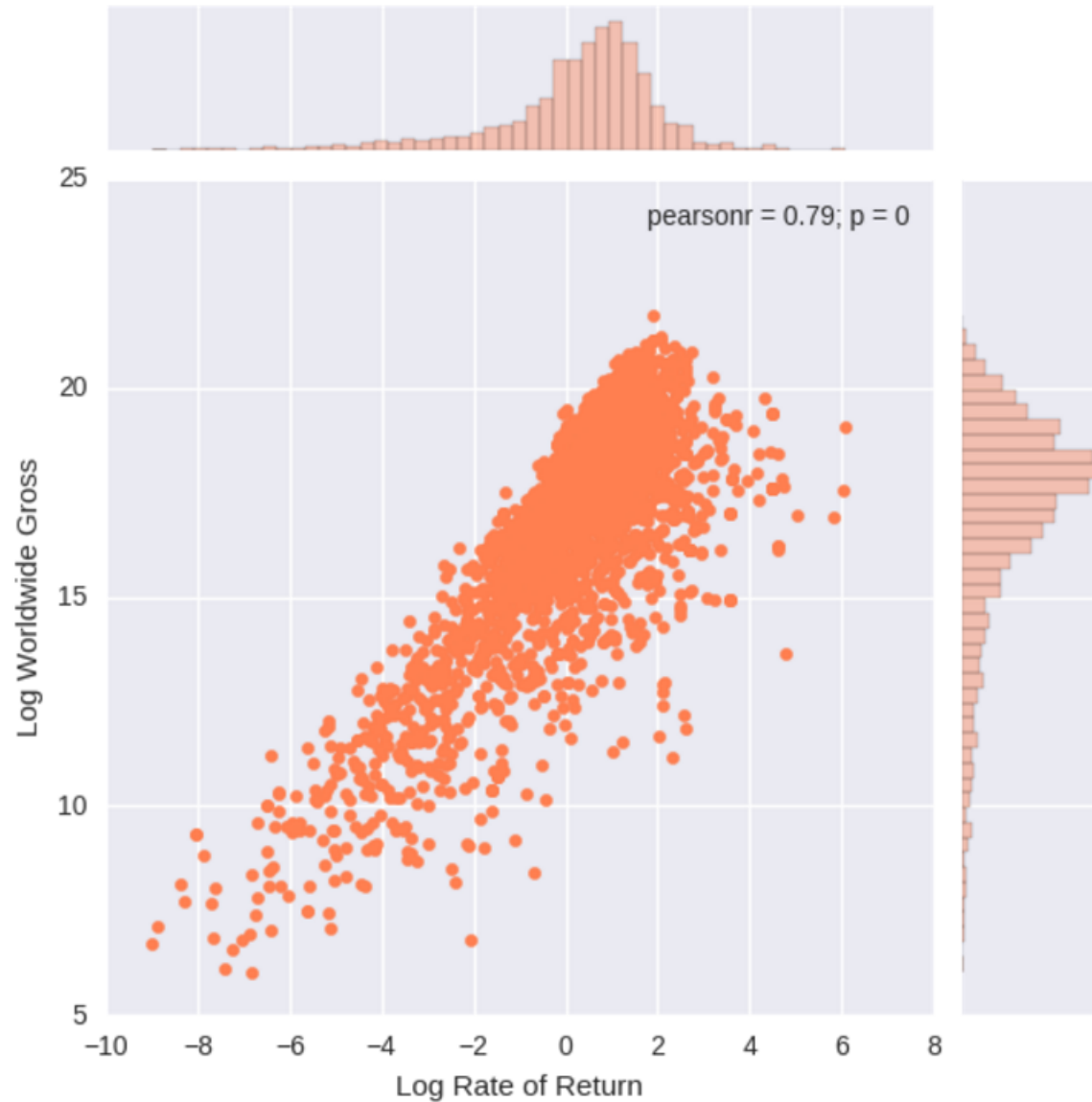
- Film duration – in minutes
- Content Rating – G, PG, PG-13, R
- Star power – Number of prior Oscar nominations for any of the 3 principle actors
- Color / black & white
- Year of release



# Methodology: Data Transformations

- Target:
  - $\ln(\text{worldwide gross} / \text{total production budget})$
- Predictors:
  - Imputed 0.2% of movie durations with mean
  - Dummy variables for movie ratings and color vs. black & white

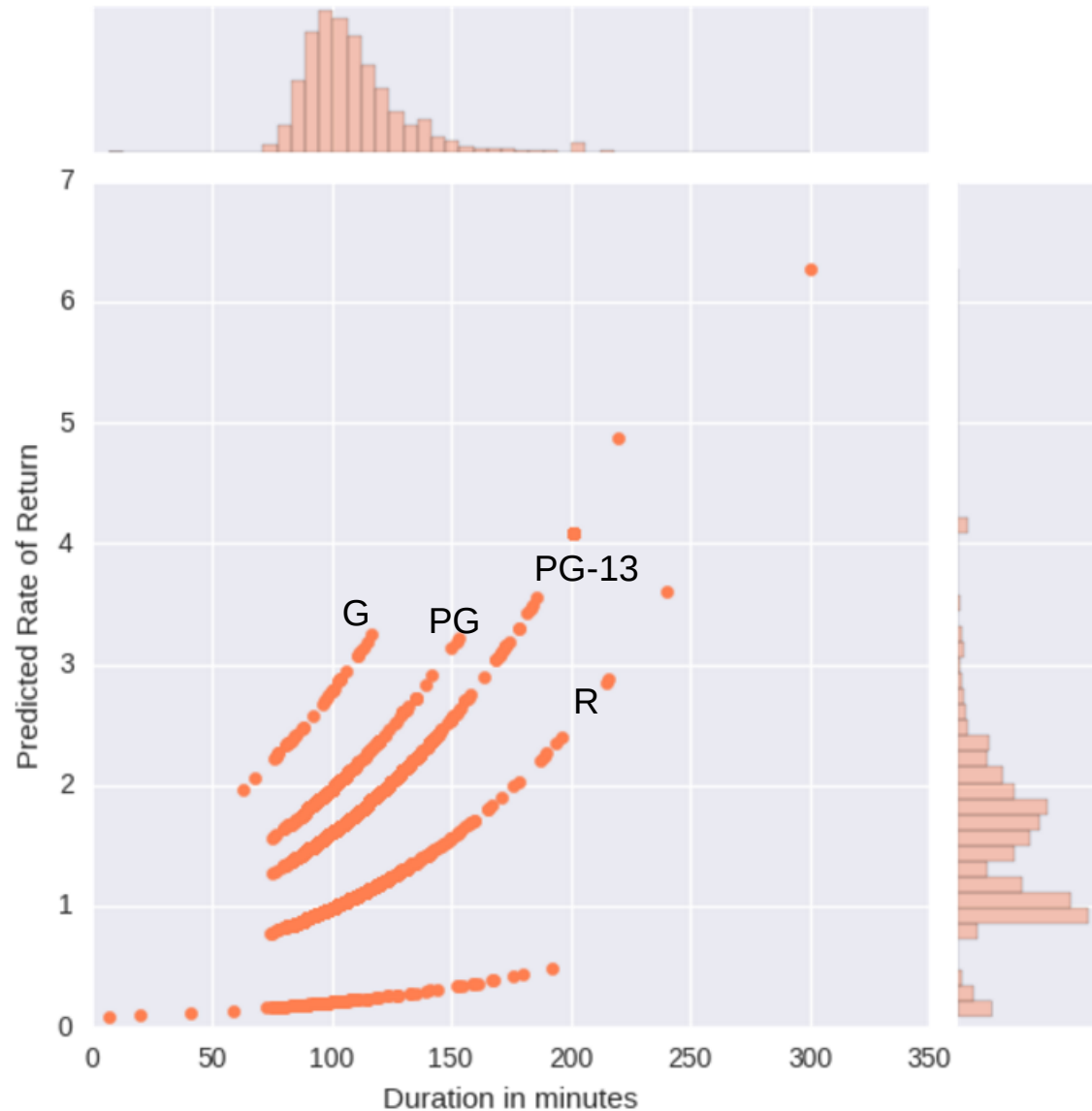
# Target Variable: Log Rate of Return



# Results: Regression Models

Response variable	Initial Model		Final Model	
	ln(rate of return)		ln(rate of return)	
R-squared	0.076		0.103	
Adjusted R squared	0.072		0.099	
N (sample size)	2227		2227	
Model degrees of freedom	10		5	
F Statistic / Prob F-stat	18.26	1.73E-32	25.42	4.10e-46
	coef	P> t	coef	P> t
duration	0.0093	0.000	0.0093	0.000
title_year	0.0088	0.261		
num_stars	0.0014	0.946		
dum_ Black and White	-0.1466	0.524		
dum_NC-17	1.9993	0.046		
dum_Not Rated	0.2088	0.508		
dum_PG	2.3762	0.000	2.2538	0.000
dum_PG-13	2.1717	0.000	2.0483	0.000
dum_R	1.6878	0.000	1.5591	0.000
dum_G	2.7414	0.000	2.5971	0.000
Intercept	-12.9504	0.197	-2.5009	0.000

# Results: Model Prediction







# Conclusion and Recommendations

- Although the low  $r$ -squared value indicates other significant sources of variability not captured here, we were able to identify significant effects of both the film's rating and duration on the return on investment.
  - It seems that the lower the content rating, and therefore the wider the potential audience, the higher return on production costs. So the making of R-rated films should perhaps be downplayed in favor of G, PG and PG-13 films.
  - The finding that films with longer duration showed an improved rate of return should not be taken to the extreme as very little data was represented in very long or very short films and it is unlikely that the underlying function is truly linear.