## PREDICTING IMDB STAR RATINGS

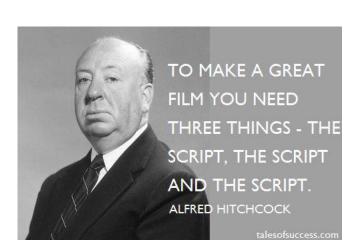
Michael Green

## TOPICS

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- Features In Data Source Used In Model
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- Linear Equation with Coefficients
- Final Test Score
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## DESCRIPTION OF THE PROBLEM

There are many ideas of how to make a good movie:



Cinema is a matter of what's in the frame and what's out.

— Martin Scorsese



## DESCRIPTION OF THE PROBLEM

- The problem is that these are not easily quantifiable.
- Many content creation companies (like Netflix) want be able to make movie making/buy decisions faster and with greater efficacy and over a larger volume of movie proposals.
- This project develops a model that uses as inputs features of movies that can be well known on or before its release date to predict if the movie will be well liked by viewers.

## DESCRIPTION OF THE PROBLEM

- This model could then assist executives by doing a quick first pass scoring of large volumes of movie proposals.
- This would enable movie executives to then focus valuable deep analysis time on already quantitatively promising movies.

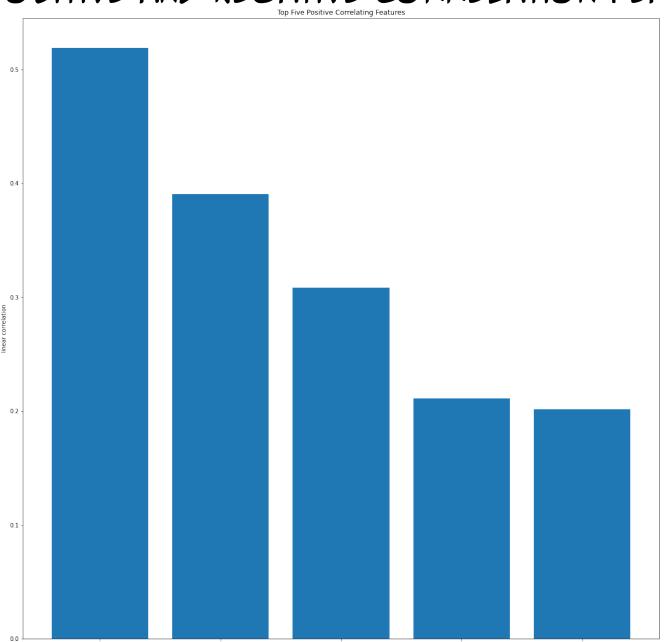
## DATA SOURCE

- The starting data set are 4000 movies released in the US with an MPAA rating between January 1, 2010 and December 31, 2019.
- This original data set was then cleaned to 2127 movies that had all required data fields defined.
- The data was scraped from imdb.com using Python requests and BeautifulSoup.

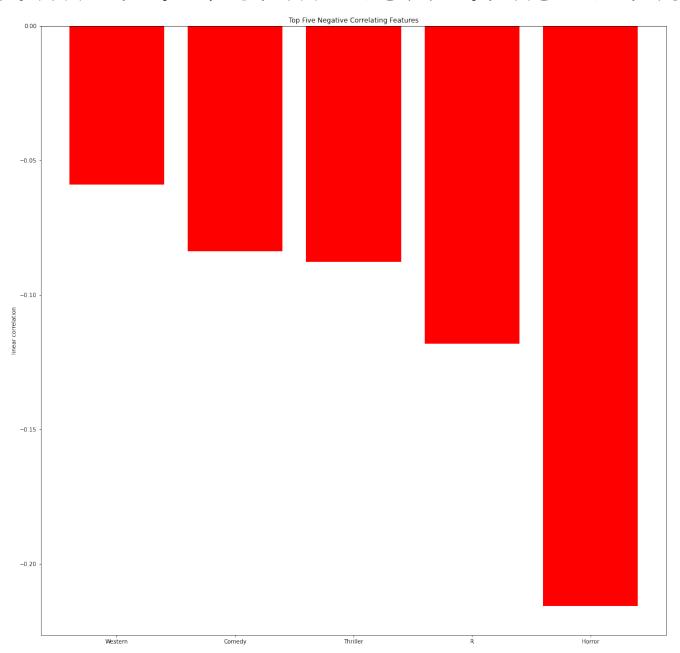
# FEATURES IN DATA SOURCE USED IN MODEL

- The following features were used in the model:
  - Runtime: length of the movie in minutes
  - Budget: Amount of money it took to make the movie (in US \$)
  - Release month of the movie: One-hot encoded (12 categories)
  - Genre of the movie: One-hot encoded (17 categories)
  - MPAA Rating of the Movie: (4 categories)
  - One-hot encoded value of the month the movie was released (12 categories encoded)
  - One-hot encoded value of the genre of the movie (17 categories encoded)
  - **Director Star Power**: 1 point for each Best Director Award for earned or nominated by the director of said movie
  - Cast 1 Star Power: 1 point for each Best Actor or Best Actress Award earned or nominated by the 1<sup>st</sup> cast member listed for the movie
  - Cast 2 Star Power: 1 point for each Best Actor or Best Actress Award earned or nominated by the 2<sup>nd</sup> cast member listed for the movie
  - Cast 3 Star Power: 1 point for each Best Actor or Best Actress Award earned or nominated by the 3<sup>rd</sup> cast member listed for the movie
  - Log(Budget): Logarithm of the Budget

#### TOP 5 POSITIVE AND NEGATIVE CORRELATION FEATURES



#### TOP 5 POSITIVE AND NEGATIVE CORRELATION FEATURES



#### LINEAR EQUATION WITH COEFFICIENTS

```
y = 0.0258runtime
  + 1.878budget
  + 0.0449August
 - 0.1492December
  0.1206February
 - 0.0363January
    -0.0672July
   - 0.0296June
  - 0.0638March
    - 0.1692May
 + 0.06582November
 + 0.02940ctober
+ 0.005259September
+ 0.3251Adventure
+ 0.8829Animation
 + 0.7957Biography
  + 0.2936Comedy
  + 0.3785Crime
  + 0.5407Drama
  -0.3144Family
 - 0.42562Fantasy
 - 0.08323Horror
  + 1.6462Music
  + 2.198Musical
  + 1.003Myster
 + 0.45469Romance
  + 0.9918Sci-Fi
  + 1.93200Sport
 - 0.53621Thriller
  - 1.778Wester
    - 1.146PG
   -0.977PG-13
```

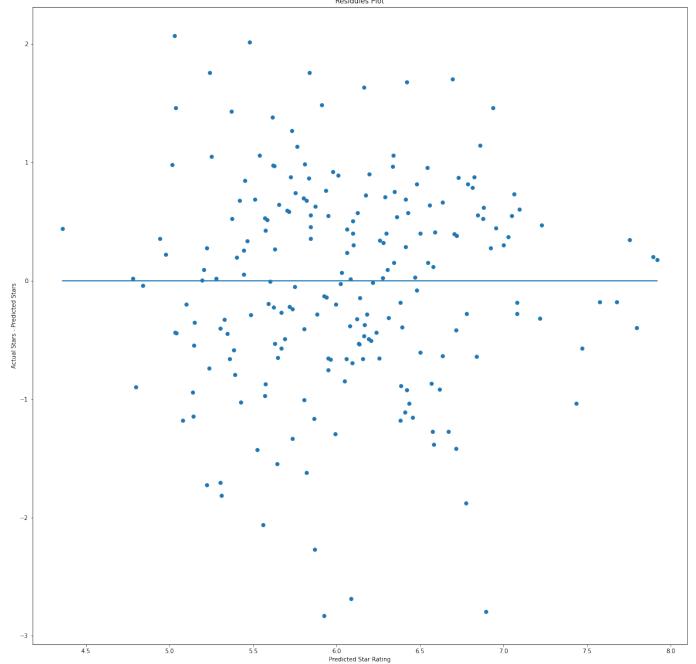
- 0.9303R + 0.18929star\_power + 0.10356\*cast1\_starpower +0\*(cast2\_starpower + cast3\_starpower) + 0.14988budget + 1.5468

## FINAL TEST SCORE

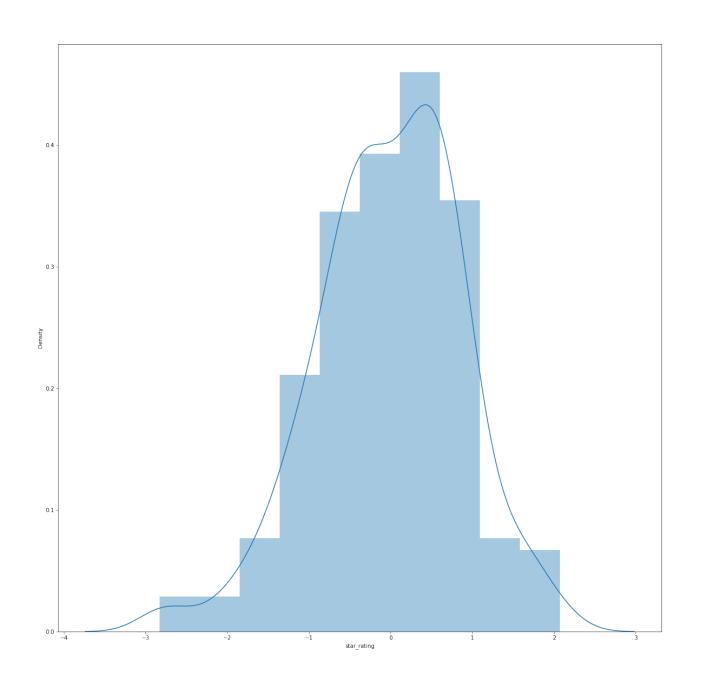
$$R^2 = 0.363$$

### RESIDUAL PLOT

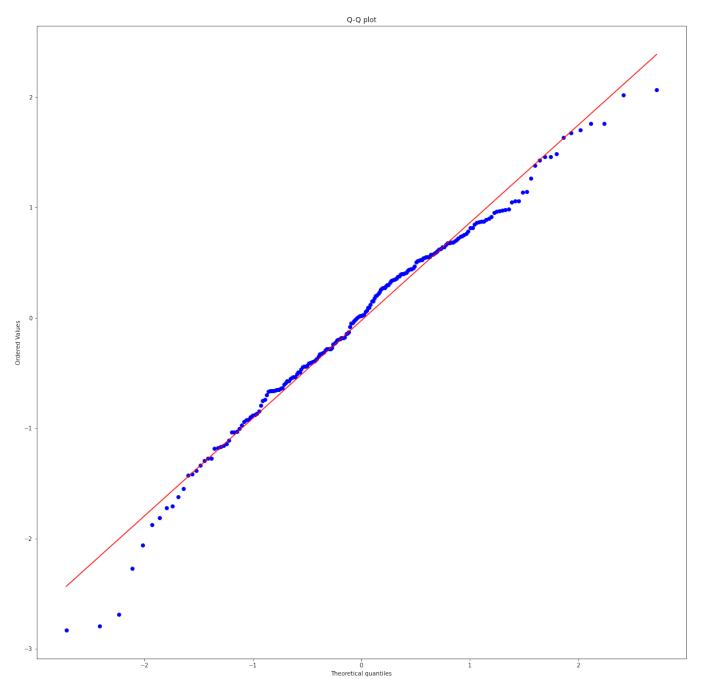




#### DISTRIBUTION OF ERROR



#### QQ PLOT



## FUTURE WORK

- Scrape more data. There are over 50,000 movies that have been released over the last 10 years.
- Collect more data on the cast and crew: How many producers on the movie have been awarded Best Picture, how many writers have been awarded for Best Adapted Screen Play.
- Dynamic Duos, Trios, Quartets, Etc.: Anecdotally there are a lot of movies where you see the same (director, cast) combinations.
   Quantify the level of influence the existence of these tuples have on the star rating of the movie.

## APPENDIX