

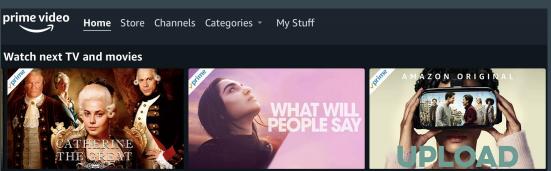
What can we learn from Movie Ratings?

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Introduction

- What are recommendation systems?
- How are they built and trained?
- What is the design trend?
 - Collaborative filtering and content-based filtering
 - Deep learning models
 - Pipelined systems: frontend & backend
- What are the design challenges now?
- How can we help?



Our Project :

- Use data to extract knowledge
- Use visualization as a means for both discovery and presentation
- Use cases:
 - o ML Engineer
 - Movie Producer
 - Theatre Manager

Data Collection & Sources:

- Kaggle
 - Netflix
 - Crawled
 - The Movie Database
- Ratings
 - User privacy
- Movies
 - Meaningful properties
- Merging

Data abstraction and data aggregation

• Datasets of 1,100 items (movies) and 26 million items (ratings)

Categorical	Ordered
Original language: 9	Budget: 0-400,000,000
Genre: 19	Revenue: 0-1,900,00,000
View rating: 9	Year: 1930-2020
Release month: 12	Rating: 1-5
Languages spoken: 26	Runtime: 70-210
Original languages: 10	Number of votes: 1-5,639



- Splitting a list of genres into multiple strings
- Aggregating budget/revenue into 10-15 groups
- Aggregating languages spoken into 6 groups
- Filtering outliers and items with missing attributes







Live demo:

https://public.tableau.com/views/MovieRatings_16395453139120/Whatisthebestgenre?:language=en-GB&publish=yes&:display_count=n&:origin=viz_share_link

What is the best genre?

Genres correlation:

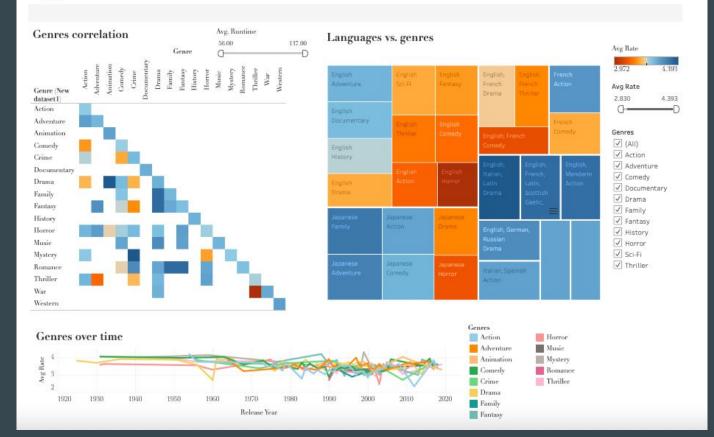
 explore combinations of different genres according to their average runtime (e.g. animated movie is the most popular when it is a dram at the same time, but not very good when it is an animated horror)

Genres over time:

- see the rating trend for each genre by selecting one or more genre on the right

Languages vs. Genres:

 explore the best combination of genres and languages spoken in the movie using the filters on the right



Rating vs. time

Ratings over time

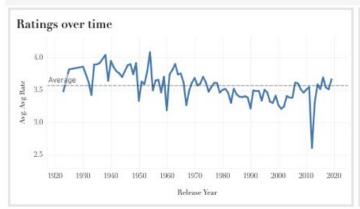
- see the development of average rating over time (e.g. movies were generally better before 1975, then it became less consistent)

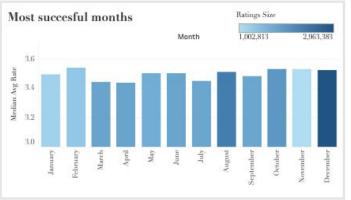
Most successful months:

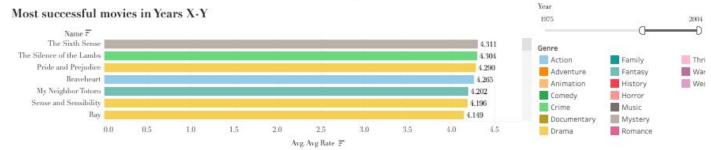
- see how which months the highest rated movies came out and what's the proportion to other months (e.g. in Deceber the most famous movies come out, however, the highest rated movies usually came out in October November)

Most successful movies in years X-Y:

- explore the hoghest rated movies throughput history and their main genres using the filter on the right







Movies Across the World

Average rating vs Country availability:

explore and compare different countries throughout the years by filters on see the rating based on how many languages are spoken in the movie the right (e.g. movies available in Brazil tend to get low rating across all

Spoken Languages:

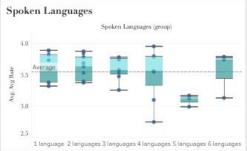
[e.g. movies with 5 languages spoken are statistically less popular]

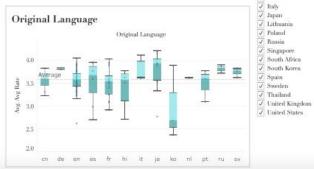
Original Language:

see the rating based on original languages of the movies (e.g. original korean movies are on average the least popular)

Average rating vs Country availability









Do money matter?

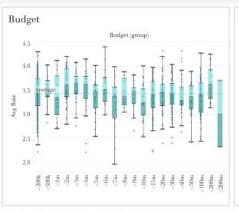
explore the user rating based on the budget for the movie and however movies for \$1-2m typically have higher than average rating) no matter the revenue)

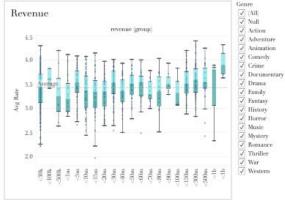
Spoken Languages:

- explore the user rating based on the revenue for the movie and different genres (e.g. low budget movies span across the whole scale, different genres (e.g. documentaries tend to have above average rating

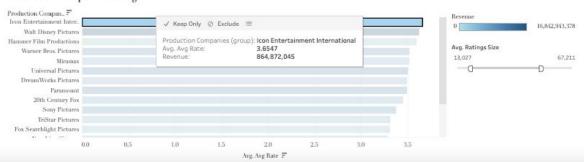
Production companies rating:

see the rating based on the production companies and explore the revenue based on the number of ratings on the movies from the respective companies [e.g. Icon Entertainment Internaitonal has a higher average rating than Paramount, who has significantly higher number of ratings



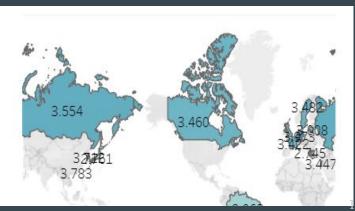


Production companies rating

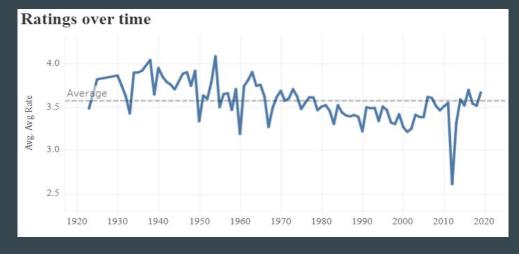


Main Findings/Results:

- Movies with the Genre Mystery have maintained a consistent rating of 3 and above.
- Movies released in February, October received the highest ratings.
- December has the highest number of ratings.
- The genres drama and comedy are more popular.
- Movies with lower budget experienced fluctuations in the average rating.
- Documentaries maintained a higher rating overall.
- Movies more popular in Russia and australia.



Main Findings/Results:



- Production companies like Icon Entertainment International had the highest average rating.
- Movies with less number of languages spoken are more popular.
- Original Korean movies are least popular.
- Crime-Comedy movies are not as popular as the crime-mystery movies.
- On the overall, movies received higher ratings before 1975.

Limitations/Future Work

- Predict trends for the future.
- Using larger dataset for better analyze the trends.
- Include more Interactivity in the visualizations by providing options of different viz techniques for the user to choose from.
- Could explore other complex visualizations which could help to analyze data in a more detailed manner.

Thank you

Q & A