

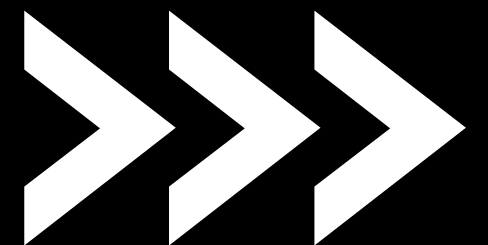


# Movie Recommendation System

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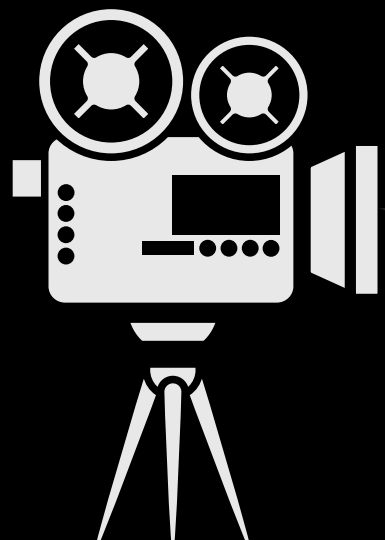
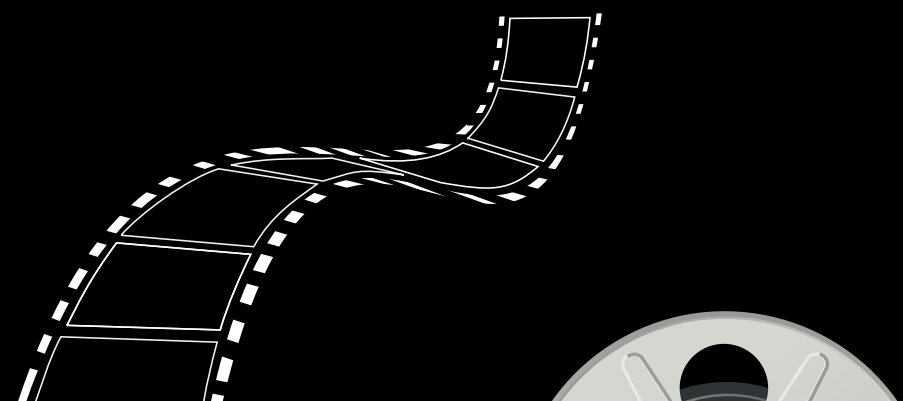
**AU1940165 Neel Popat**



# Introduction

## Movie Recommendation system :

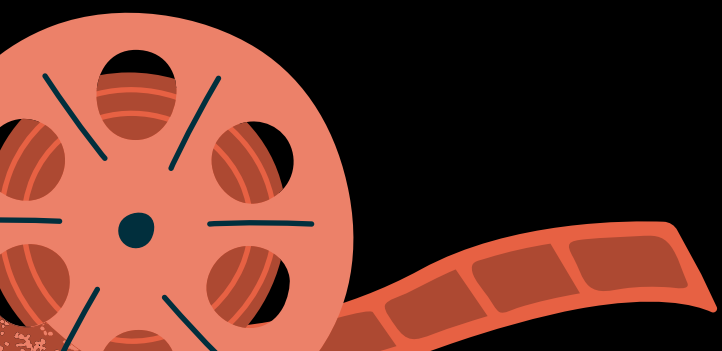
- Various OTT (over the top) platforms uses a recommendation system for its users making the interface user-oriented.
- Users spends a lot of time in finding what they might like because they have 1000s of options.
- Hence it is important to have such an algorithm which helps the users to find their interests.





## Problem Statement

The project aims to assist the Users in finding content which he/she likes using concepts based on weighted mean, cosine similarity and singular value decomposition and hence reducing the time spent by the users to find appropriate content.



# Existing body of work

- Currently, all the OTT platforms, google, youtube and all kinds of search engines use recommendation strategy to retain their users for maximum amount of time.
- Recommendations are given based on popularity, views, similar products, super affinity, etc.
- Netflix, "Netflix prize data," Kaggle, 13-Nov-2019. [Online]. Available: <https://www.kaggle.com/netflix-inc/netflix-prize-data>. [Accessed: 30-Mar-2022].







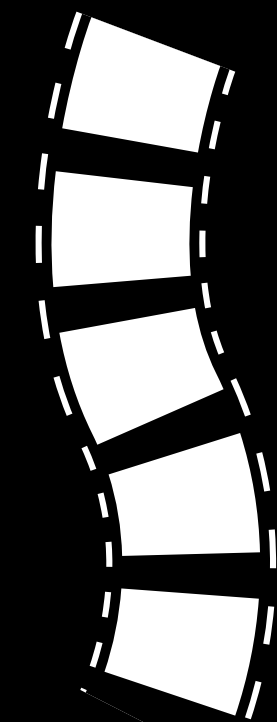
# Weighted mean

$$\text{Weighted Rating (WR)} = \left( \frac{v}{v+m} \cdot R \right) + \left( \frac{m}{v+m} \cdot C \right)$$

where,

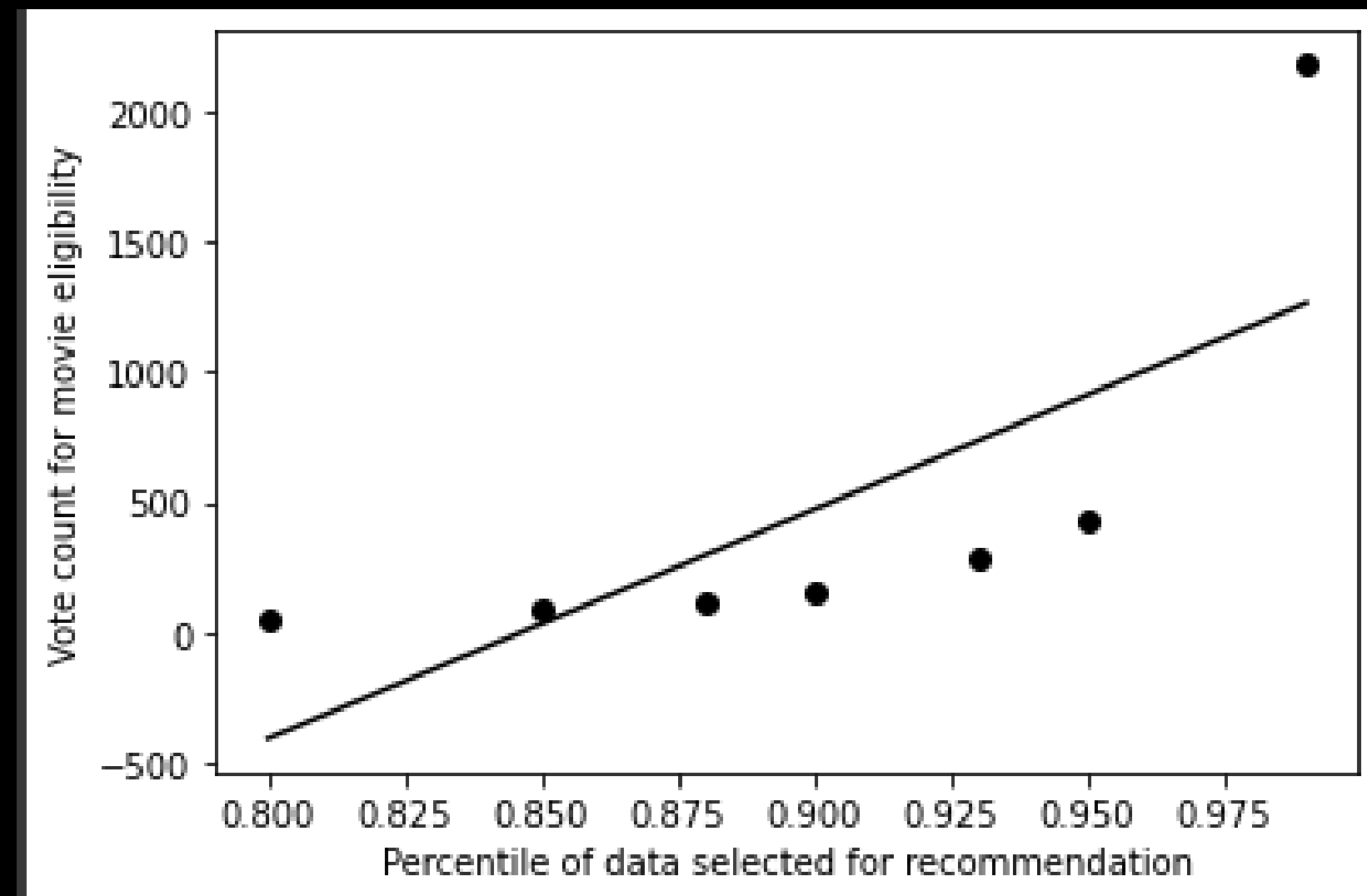
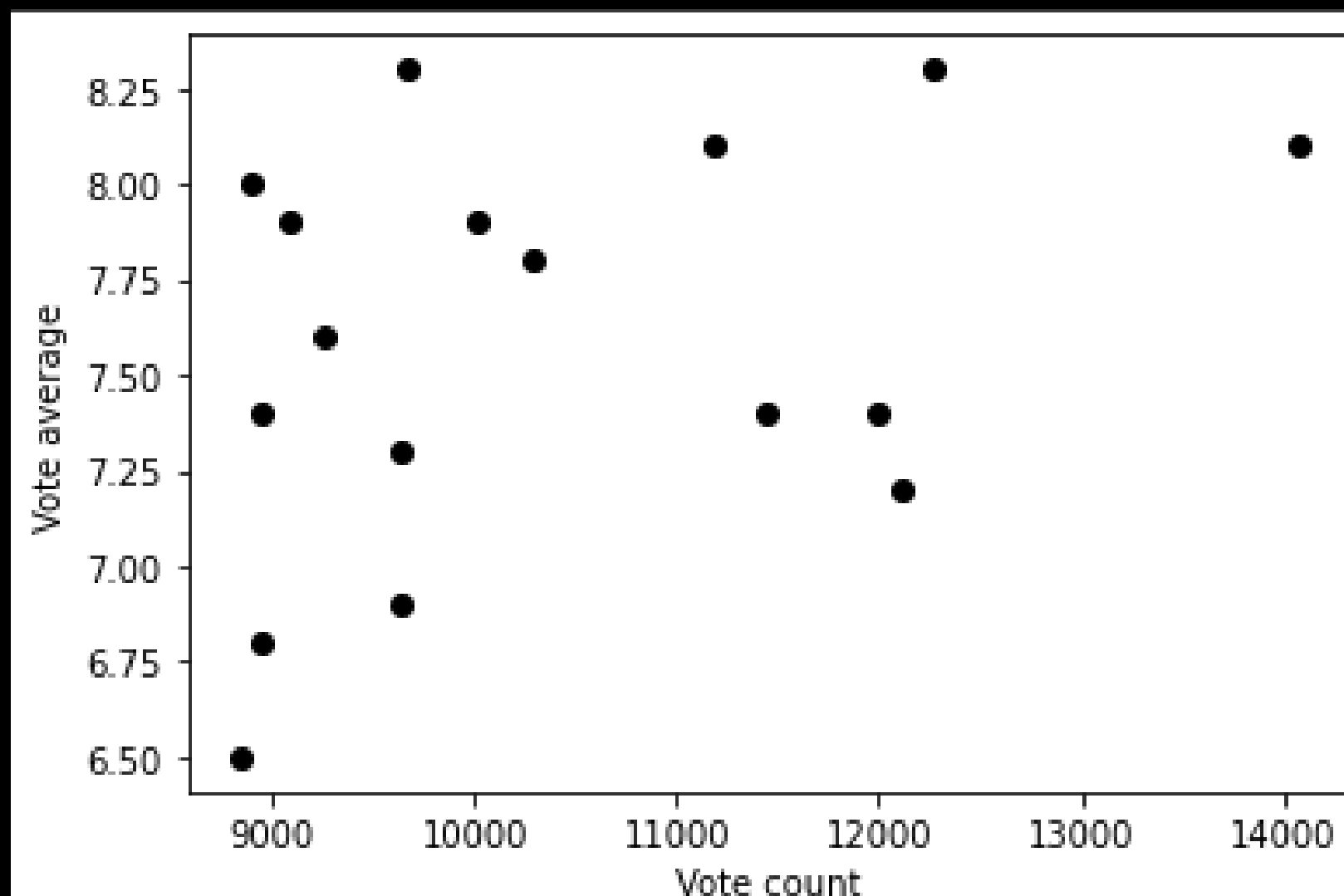
- $v$  is the number of votes for the movie
- $m$  is the minimum votes required to be listed in the chart
- $R$  is the average rating of the movie
- $C$  is the mean vote across the whole report

- Create one parameter "weighted rating" which considers all 4 parameters above for a movie.
- Does not give personalized recommendations.

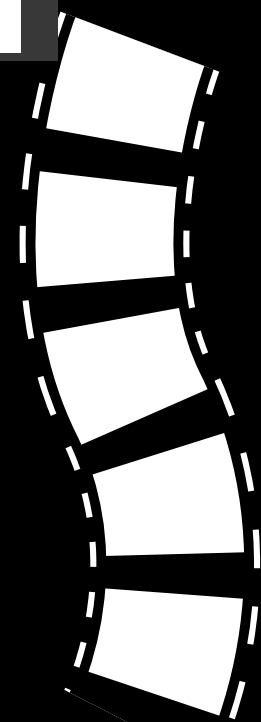




# Weighted mean



- In the first plotted graph, data is filtering at 0.95 th percentile of the movies present in the data set containing 17 movies.
- In the second graph, we are fitting a linear regression to the percentile and vote count graph.



# Initial Results

- Genre based movie recommendation
- Output for genre: ROMANCE

```
build_chart('Romance').head(5)
```




	title	year	vote_count	vote_average	popularity	wr
10309	Dilwale Dulhania Le Jayenge	1995	661	9	34.457024	8.565285
351	Forrest Gump	1994	8147	8	48.307194	7.971357
876	Vertigo	1958	1162	8	18.20822	7.811667
40251	Your Name.	2016	1030	8	34.461252	7.789489
883	Some Like It Hot	1959	835	8	11.845107	7.745154





# Initial Results

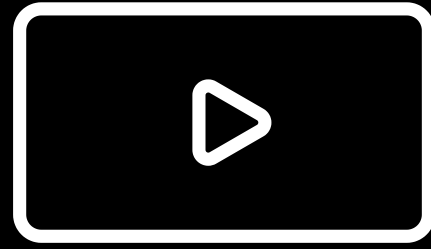
## Output for genre: ACTION



The image shows a Jupyter Notebook interface. At the top, there is a code cell with the command `build_chart('Action').head(5)`. Below the code cell, there is an output cell displaying a table of movie data. The table has columns for rank, title, year, vote\_count, vote\_average, popularity, and weighted rating (wr). The top 5 movies are listed, with Inception at the top, followed by The Dark Knight, and three movies from The Lord of the Rings.

	title	year	vote_count	vote_average	popularity	wr
15480	Inception	2010	14075	8	29.108149	7.955099
12481	The Dark Knight	2008	12269	8	123.167259	7.948610
4863	The Lord of the Rings: The Fellowship of the Ring	2001	8892	8	32.070725	7.929579
7000	The Lord of the Rings: The Return of the King	2003	8226	8	29.324358	7.924031
5814	The Lord of the Rings: The Two Towers	2002	7641	8	29.423537	7.918382

INFERENCE : The movies are recommended based on one parameter - weighted rating. Higher weighted rating suggests more priority to a movie.

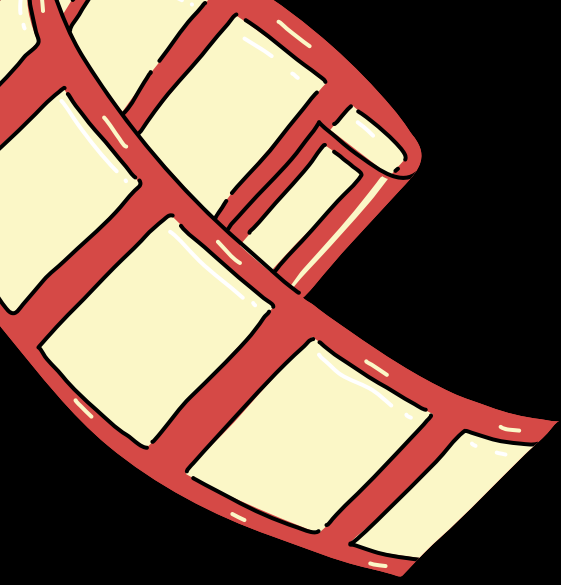


## Contribution of each member

**Yashvi Navadia** : Implemented weighted mean and explored various functions in python used in the coding part, coding in Simple recommender system.

**Kashish Jivani** : Reading various articles on the topics which have ML models to solve the problem, coding in Simple recommender system.

**Neel Popat** : Handled and gathered the datasets, figuring out the cosine similarity, coding in Content based filtering.



# FUTURE WORKS

Application and Implementation :

- Content based filtering
- Collaborative based filtering
- Hybrid Recommendation System



# References

- C. Z. Omega and Hendry, "Movie Recommendation System using Weighted Average Approach," 2021 2nd International Conference on Innovative and Creative Information Technology (ICITech), 2021, pp. 105-109, doi: 10.1109/ICITech50181.2021.9590147.
- S. Agrawal and P. Jain, "An improved approach for movie recommendation system," 2017 International Conference on I-SMAC (IoT in Social, Mobile, Analytics and Cloud) (I-SMAC), 2017, pp. 336-342, doi: 10.1109/I-SMAC.2017.8058367.
- C. M. Wu, D. Garg and U. Bhandary, "Movie Recommendation System Using Collaborative Filtering," 2018 IEEE 9th International Conference on Software Engineering and Service Science (ICSESS), 2018, pp. 11-15, doi: 10.1109/ICSESS.2018.8663822.

Thank you

