Building a Movie Recommender

Overview

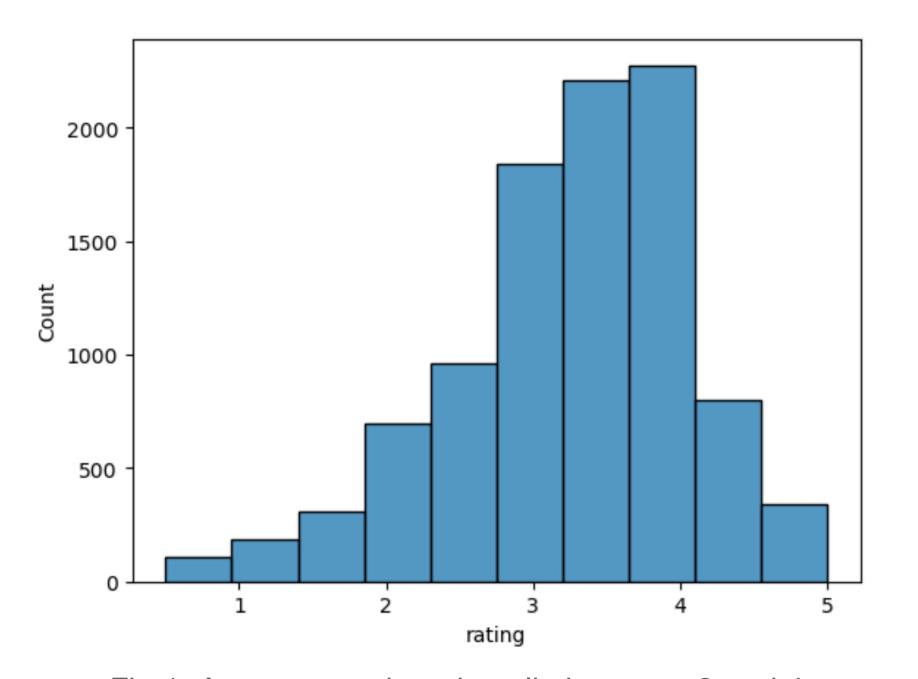
- Business problem
- Data and model
- Recommendations and areas for further analysis
- Conclusion

Business Problem

- Recommendation systems for media content are utilized by streamers to help users make decisions
- Streamers effectively collect their own data given user decisions every day

Data

- 100k IMDB dataset movie title and genres for ~10k movies
- ~100k user ratings from a total of 610 users



6000 - 4000 - 2000 - 50 100 150 200 250 300 Number of Ratings

8000

Fig 1. Average movie ratings lie between 3 and 4

Fig 2. Most movies have a relatively small number of ratings

Recommender Approaches

- Collaborative filtering: recommendations solely based on user ratings
- Content filtering: recommendations based on movie content
- Hybrid approach uses both
- For this project:
 - Use collaborative filtering to produce the ratings
 - Use movie genres to further filter specific recommendations

Final Model

- Collaborative filtering using SVD model tuned on data
- No improvement from incorporating genre preferences
- Cold start problem (no existing user): give most highly rated films based on genre preferences
- Performance metrics:
 - RMSE 0.55
 - MAE 0.42

Final Model

Fig 3. Recommender output for a new user inputting genres Fantasy, Comedy, Action

```
1. 5634 Monday (2000)
Name: title, dtype: object
2. 8597 Dragonheart 2: A New Beginning (2000)
Name: title, dtype: object
3. 9394 Maximum Ride (2016)
Name: title, dtype: object
4. 899 Princess Bride, The (1987)
Name: title, dtype: object
5. 960 Evil Dead II (Dead by Dawn) (1987)
Name: title, dtype: object
```

Fig 4. Recommender output for an existing user inputting genres Fantasy, Comedy, Action

```
1. 899 Princess Bride, The (1987)
Name: title, dtype: object
2. 916 Army of Darkness (1993)
Name: title, dtype: object
3. 2798 Big Trouble in Little China (1986)
Name: title, dtype: object
4. 8881 Kung Fury (2015)
Name: title, dtype: object
5. 8357 The Lego Movie (2014)
Name: title, dtype: object
```

Future Improvements

- Incorporate hybrid content more broadly lots of data for movies
- Try using similarities between movies rather than users data size issue

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