

# Movie Recommendation System

This project demonstrates a personalized movie recommendation system using the <u>MovieLens 100K dataset</u>. It leverages collaborative filtering and hybrid modeling techniques to recommend movies based on user preferences.

# **Project Overview**

- Goal: Build a movie recommender system that provides accurate and relevant movie suggestions to users.
- Dataset: MovieLens 100K, including ratings.csv, movies.csv, and tags.csv.

# **Business Understanding**

Modern streaming platforms must recommend content users will enjoy to drive engagement. This system is built to simulate such a platform by learning from historical ratings and genres.

## **Key Business Questions**

- What user behaviors or rating patterns can be leveraged to predict future movie preferences?
- Which recommendation technique offers better predictive performance: memory-based collaborative filtering or model-based filtering
- Can the model generate high-quality, personalized Top 5 recommendations for individual users?

#### **Data Understanding & Cleaning**

The MovieLens 100K dataset includes:

- ratings.csv, which contains four columns: userld, movield, rating, and timestamp, representing a user's interaction with a movie.
- movies.csv, which includes metadata such as the movie title and associated genres.
- tags.csv, which logs user-submitted tags on movies, potentially useful for content-based or hybrid filtering.
- links.csv, which provides external identifiers such as IMDb and TMDb IDs for integration with third-party data.

### **Data Preparation**

- Cleaned datasets which include ratings.csv and movies.csv to produce a structured dataset.
- Filtered to users with ≥20 ratings and movies with ≥50 ratings, yielding 41,360 ratin
- Transformed data using the Surprise Reader object.
- Used an 80/20 train-test split to evaluate performance.

### **Exploratory Analysis**

- Found users favor highly-rated movies in Drama, Comedy, and Thriller genres.
- Active users and popular movies provide strong signals for recommendations.

### Modeling

Implemented two collaborative filtering models using the Surprise library:

Model	RMSE	MAE
SVD	0.8368	0.6395
KNNBasic	0.8844	0.6753

SVD outperformed KNNBasic on both metrics.

### Top Recommendations (User 555)

- 1. Alien (1979) 🖈 4.55
- 2. Raiders of the Lost Ark (1981)
- 3. Clockwork Orange, A (1971)
- 4. Star Wars: Episode V The Empire Strikes Back (1980)
- 5. Terminator 2: Judgment Day (1991)

### **Hybrid Filtering**

- Combined SVD with genre-based cosine similarity.
- Addressed cold-start issues for new users.
- Blended scores with alpha=0.7 (70% CF, 30% content-based).

### **Insights**

- Drama, Comedy, and Thriller are the most frequently rated genres.
- Hybrid models combining collaborative and content-based filtering improve cold-start performance.
- Regular model retraining is important to capture evolving user behavior.

#### Conclusion

The system combines collaborative and hybrid filtering to deliver personalized movie recommendations, addressing user engagement challenges effectively.

#### Recommendations

- Deploy SVD-based recommender as the core engine.
- Introduce hybrid filtering for improved cold-start handling.

• Periodically retrain the model on new data.

#### Releases

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#### **Packages**

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#### Contributors 3



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#### Languages

Jupyter Notebook 100.0%