

# Movie Recommender *App*



**Built upon Deep Neural Network & Flask**

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# Why Movie Recommender?

- ★ Increasing entertainment market
- ★ Personalization on product marketing
- ★ Better engage customers on movie/tv products
- ★ Drive sales



# Data Source & Properties

- ★ Amazon Movies/TV reviews from [UCSD](#)
- ★ Json.gz file with 19 years data (8,765,568 reviews)
- ★ Subsample to 2018 ratings/reviews
  - reduce computational cost
- ★ Scrape review webpage link for each movie using BeautifulSoup
- ★ Drop unrelated variables

# Data Exploration



N = 209,060



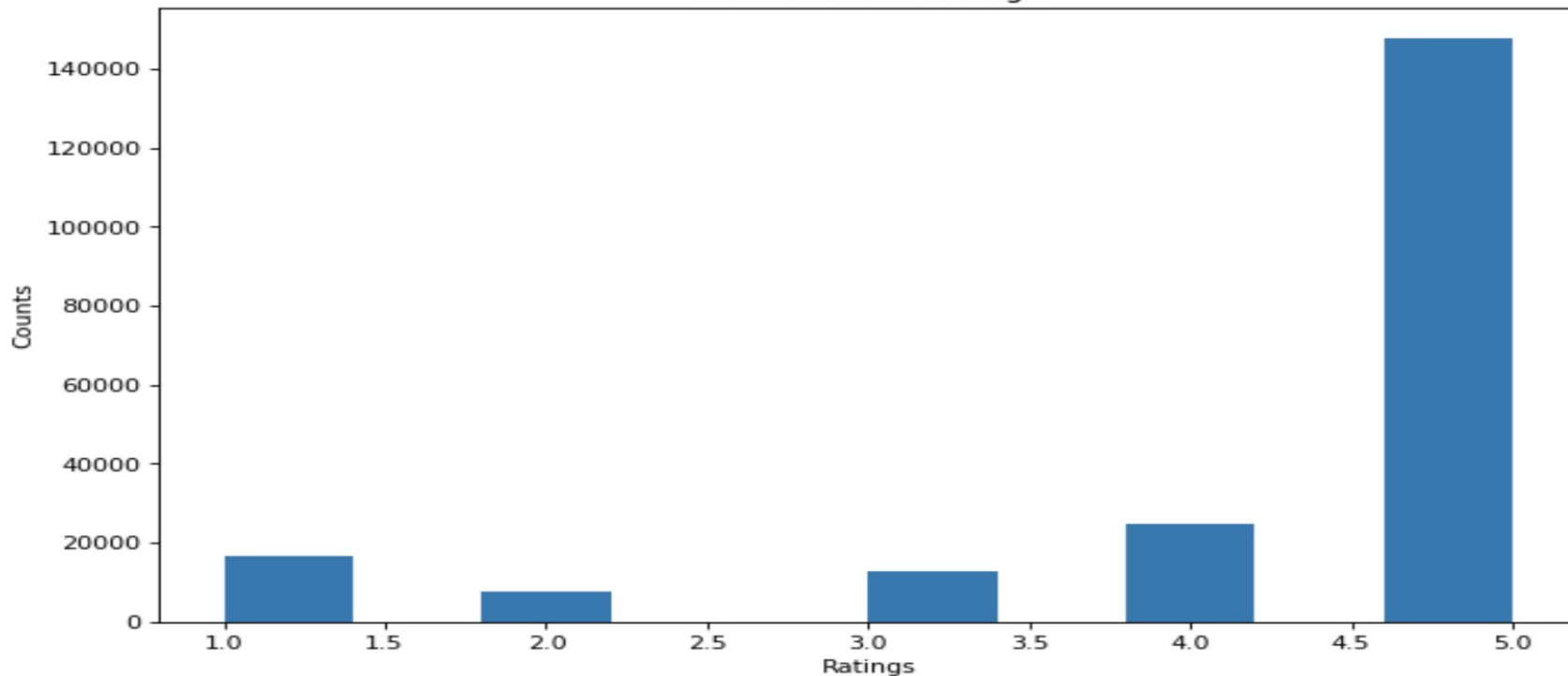
N= 38,864  
Mean: 17.2  
Max: 959  
25% : with 1 review



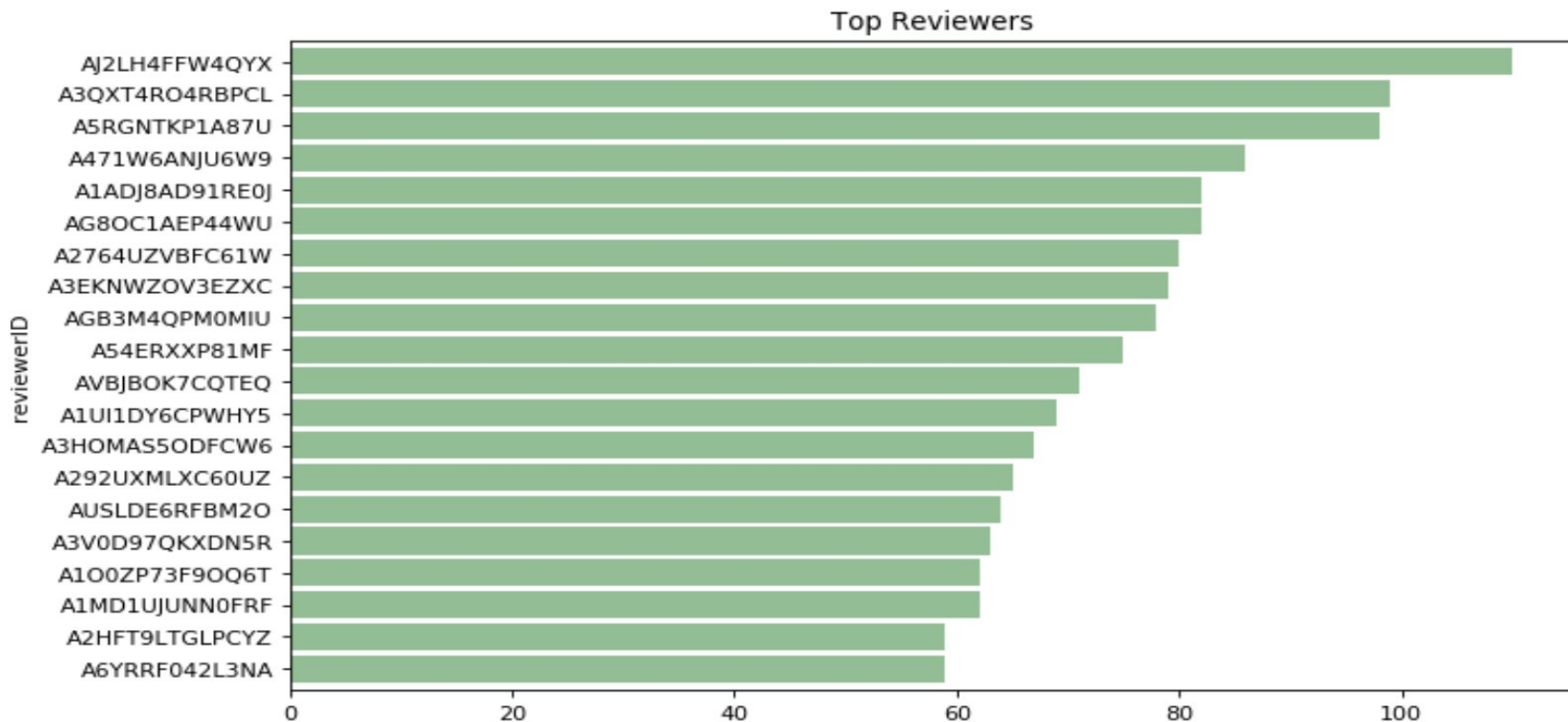
N=119,945  
Mean: 1.74  
Max: 110  
50%: with 1 review

# Data Exploration - Rating Distribution

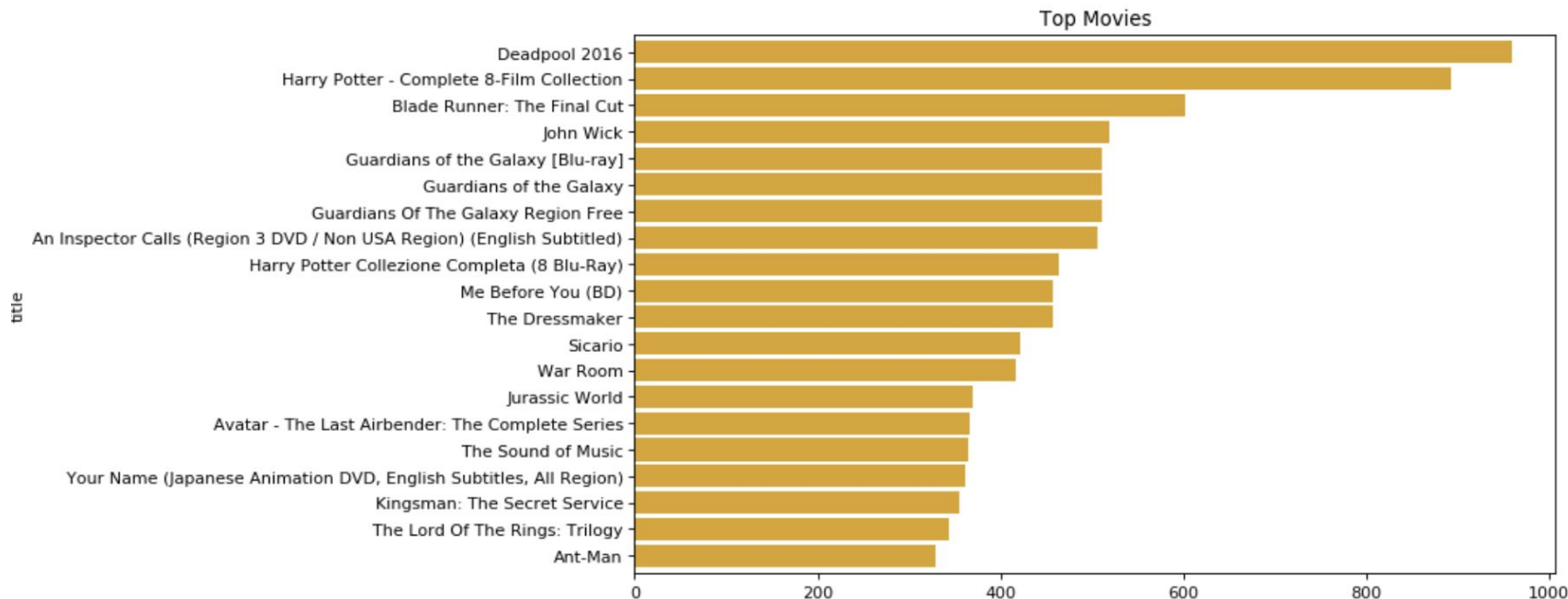
Distribution of Ratings



# Data Exploration - Top Reviewers by count of reviews



# Data Exploration -Top Movies by Count of Reviews



# Collaborative Filtering Using Neural Networks

## ★ Collaborative filtering

- based on users' rates
- recommend user A movies that users similar to A have watched & like

## ★ Keras embedding:

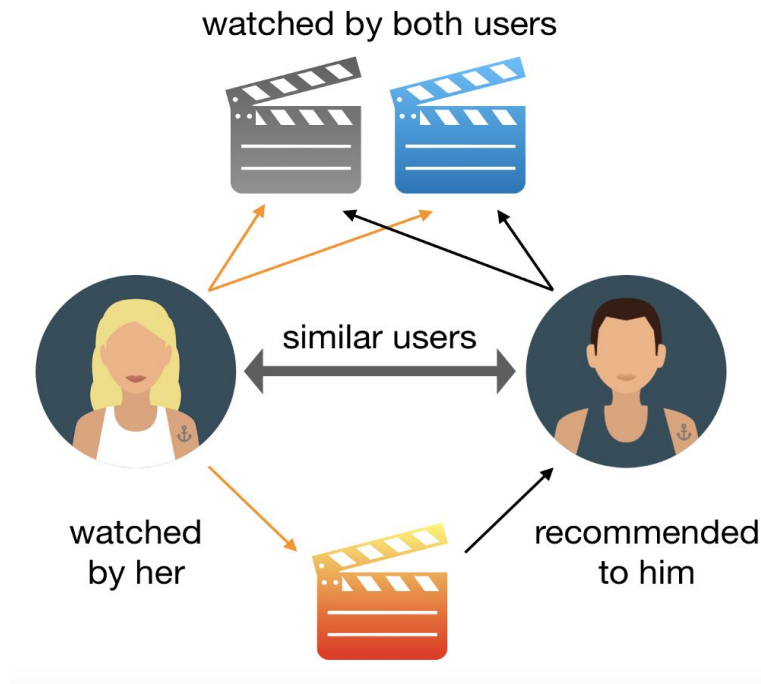
- split one matrix into two smaller matrix
  - high dimension → low dimensions

## ★ Neural networks:

- efficiently learn the underlying explanatory factors and useful representations

## ★ Evaluation metrics:

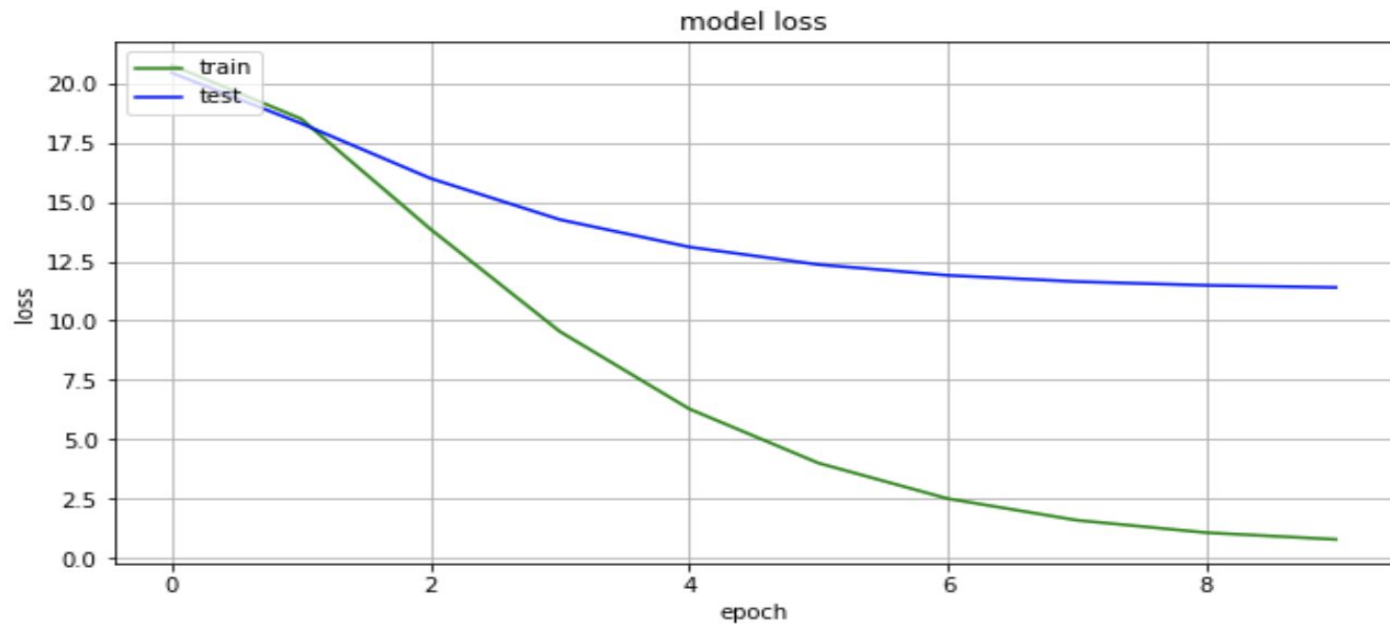
- Mean Absolute Error (MAE)





# Base Model (input $\rightarrow$ output)

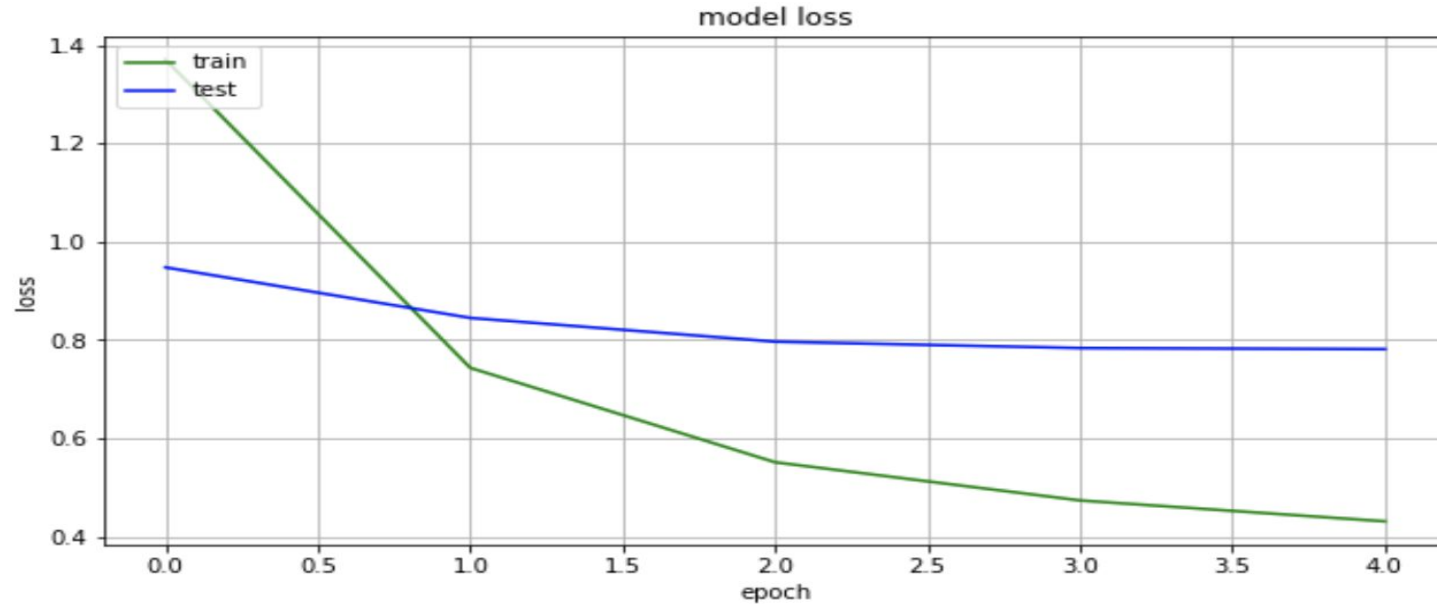
★ MAE = 2.4



Loss function: Mean squared error

# Final Model (input → hidden layers/dropout → output)

★ MAE = 0.43



Loss function: Mean squared error

# Cross Validation on Final Model

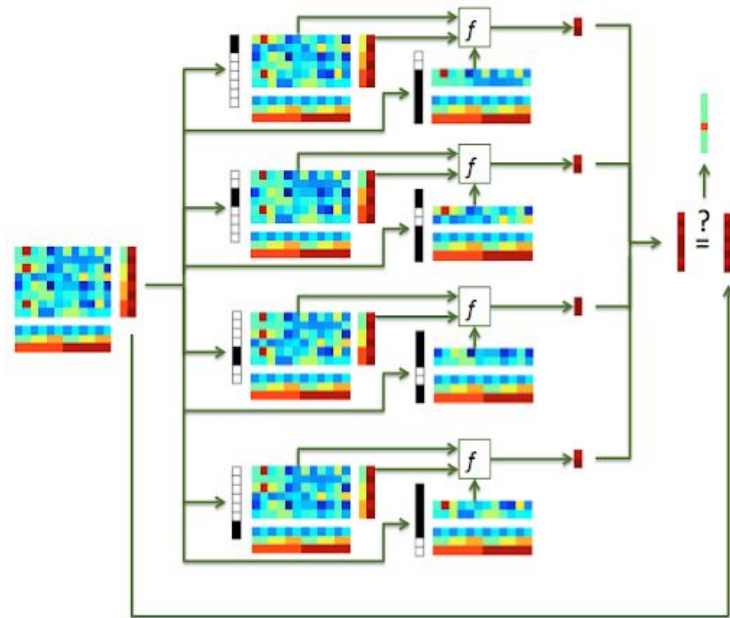
Metrics: Mean Absolute Error (MAE)

StratifiedKFold, n\_splits=5

- ★ Split 1: 0.41
- ★ Split 2: 0.40
- ★ Split 3: 0.38
- ★ Split 4: 0.37
- ★ Split 5: 0.38

Average of MAE: 0.39

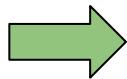
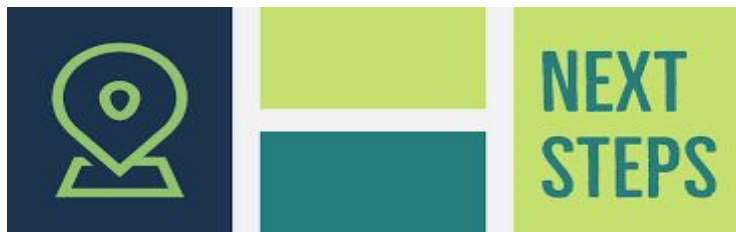
Standard Deviation of MAE: 0.0124



# Make Recommendation



GO!!!



1. Cluster users & weight on rates
2. Build a recommendation system based on sentiment analysis on review texts
3. Compare two models & the combination of the two
4. Visualize keras embedding
5. Improve deployment
6. Build a new environment

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