## **Evaluation Metrics for Recommender Systems**

Installing and Importing Important Libraries like Surprise and recmetrics apart from Numpy and Pandas

```
import matplotlib.pyplot as plt
import numpy as np
import pandas as pd
!pip install surprise
import surprise
from surprise import SVD, Dataset, Reader
from surprise.model_selection import train_test_split
!pip install recmetrics
import recmetrics
```

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Importing CSV files for analysis. This data contains user's ratings of movies, as well as movie genre tag.

```
ratings = pd.read_csv('/content/ratings.csv')
ratings = ratings.query('rating >=3')
ratings.reset_index(drop=True, inplace=True)
ratings.head()
        userId movieId rating timestamp
                                                    ılı.
     0
             1
                    296
                             5.0 1147880044
     1
                             3.5 1147868817
             1
                    306
     2
             1
                    307
                             5.0 1147868828
     3
                             5.0 1147878820
             1
                    665
                             3.5 1147868510
     4
             1
                    899
#only consider ratings from users who have rated over n movies
n=1000
users = ratings["userId"].value_counts()
users = users[users>n].index.tolist()
ratings = ratings.query('userId in @users')
print(ratings.shape)
ratings.head(3)
     (81874, 4)
            userId movieId rating timestamp
                                                       16
     19897
               187
                          1
                                 3.5 1277374478
     19898
                          2
                                 3.5 1277374864
               187
     19899
                                 3.0 1277839361
               187
                          3
rated_movies = ratings["movieId"].tolist()
movies = pd.read_csv('/content/movies.csv')
movies = movies.query('movieId in @rated_movies')
movies.set_index("movieId", inplace=True, drop=True)
movies = movies["genres"].str.split("|", expand=True)
movies.reset index(inplace=True)
movies = pd.melt(movies, id_vars='movieId', value_vars=[0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
movies.drop_duplicates("movieId", inplace=True)
movies.set_index('movieId', inplace=True)
movies = pd.get_dummies(movies.value)
movies.head()
                                                                       Film-
    e Animation Children Comedy Crime Documentary Drama Fantasy
                                                                              Horror Musical Myster
                                                                         Noir
               0
                         0
                                 0
                                        0
                                                     0
                                                            0
                                                                     0
                                                                            0
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    1
               0
                                        0
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    1
                         0
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    )
                         0
                                 1
                                                            0
                                                                            0
                                                                                    0
```

## Long Tail Plot

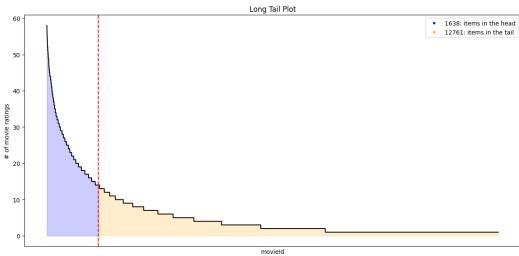
D

D

This plot is used to explore popularity patterns in user-item interaction data such as clicks, ratings, or purchases. Typically, only a small percentage of items have a high volume of interactions, and this is referred to as the "head". Most items are in the "long tail", but they only make up a small percentage of interactions.

/usr/local/lib/python3.10/dist-packages/seaborn/\_decorators.py:36: FutureWarning: Pass the fol warnings.warn(

/usr/local/lib/python3.10/dist-packages/recmetrics/plots.py:60: FutureWarning: The frame.appen head = head.append(tail.head(1))

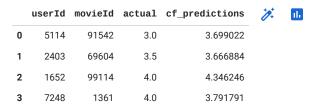


```
reader = Reader(rating_scale=(0, 5))
data = surprise.Dataset.load_from_df(ratings[['userId', 'movieId', 'rating']], reader)
trainset, testset = train_test_split(data, test_size=0.25)
```

SVD reduces the dimensionality reduction and gets us important latent features which can almost approximate all values in the utility matrix (including null values). Already available values in the utility matrix can be used to evaluate the predictions and tune the parameters/weights in the latent features (if matrix factorization used) or help us to pick the number of latent features from the SVD decomposition. Once number of latent features to be picked are decided, we can populate the null values and use these predicted ratings of a user for movie to recommend the movie with highest predicted rating to that particular user.

Making Predictions based on the model created above

```
test = algo.test(testset)
test = pd.DataFrame(test)
test.drop("details", inplace=True, axis=1)
test.columns = ['userId', 'movieId', 'actual', 'cf_predictions']
test.head()
```



Evaluate model with MSE and RMSE

print("MSE: ", recmetrics.mse(test.actual, test.cf\_predictions))
print("RMSE: ", recmetrics.rmse(test.actual, test.cf\_predictions))

MSE: 0.30261220628201635 RMSE: 0.5501019962534369

Double-click (or enter) to edit