ADA Assignment 4

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- MCS202215

0. Imports

```
In [ ]: # for file handling
        import os, sys
        # for data manipulation
        import numpy as np
        import pandas as pd
        # for data visualization
        import matplotlib.pyplot as plt
        import seaborn as sns
        # sklearn surprise imports
        from surprise import Dataset, Reader, SVD
        from surprise.model_selection import train_test_split
        from surprise import accuracy
        from surprise.dataset import DatasetAutoFolds
In [ ]: # ignore warnings
        import warnings
        warnings.filterwarnings('ignore')
```

1. Data Loading

```
In []: data_dir_path = os.path.join('data', 'raw', 'ml-25m')

# read the ratings.csv file as pandas dataframe
df_ratings = pd.read_csv(os.path.join(data_dir_path, 'ratings.csv'))

reader = Reader(rating_scale=(0.5, 5))
data = Dataset.load_from_df(df_ratings[['userId', 'movieId', 'rating']], reader)

df_ratings.head()
```

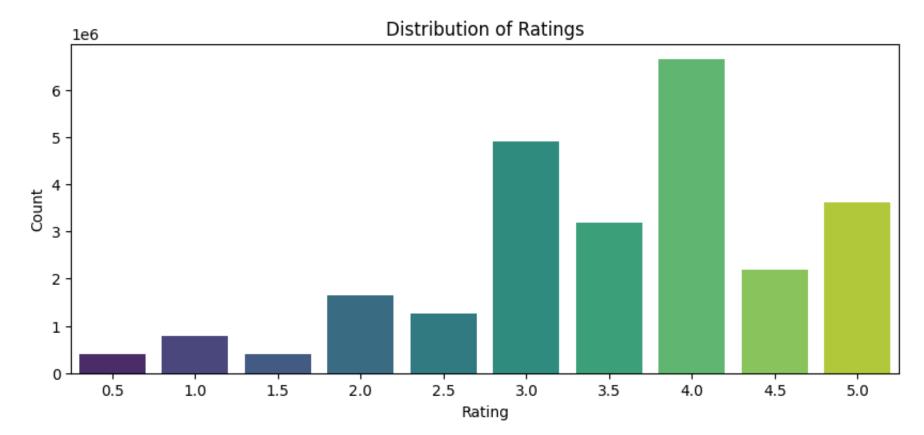
Out[]:		userId	movield	rating	timestamp
	0	1	296	5.0	1147880044
	1	1	306	3.5	1147868817
	2	1	307	5.0	1147868828
	3	1	665	5.0	1147878820
	4	1	899	3.5	1147868510

2. EDA

```
In []: # print the shape of the dataframe
    print('Shape of the dataframe: ', df_ratings.shape)

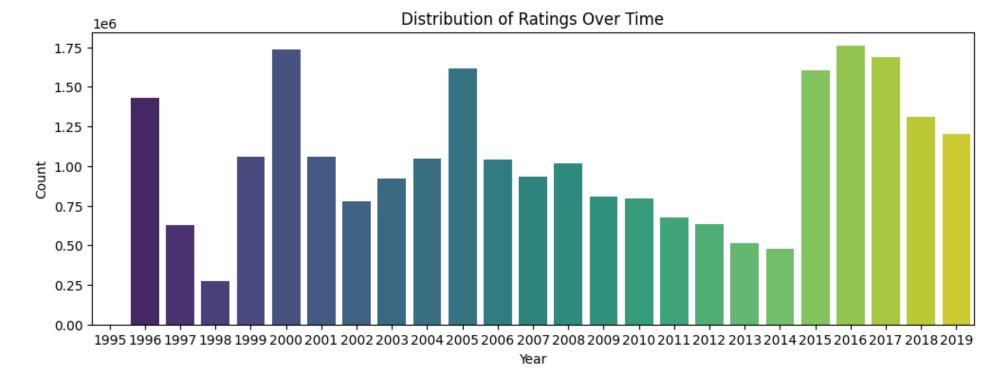
Shape of the dataframe: (25000095, 4)

In []: # distribution of ratings
    plt.figure(figsize=(10, 4))
    sns.countplot(x='rating', data=df_ratings, palette='viridis')
    plt.title('Distribution of Ratings')
    plt.xlabel('Rating')
    plt.ylabel('Count')
    plt.show()
```



```
In []: # distribution of ratings over time
    df_ratings['timestamp'] = pd.to_datetime(df_ratings['timestamp'], unit='s')
    df_ratings['year'] = df_ratings['timestamp'].dt.year

plt.figure(figsize=(12, 4))
    sns.countplot(x='year', data=df_ratings, palette='viridis')
    plt.title('Distribution of Ratings Over Time')
    plt.xlabel('Year')
    plt.ylabel('Count')
    plt.show()
```



3. Train Test Split

```
In []: # split the data into training and testing sets
trainset, testset = train_test_split(data, test_size=0.2, random_state=42)
```

4. Model Fitting (SVD)

```
In []: # train the SVD model
    model = SVD()
    model.fit(trainset)
```

5. Model Evaluation

```
In []: # make predictions on the test data
testset_pred = model.test(testset)

In []: # Calculate RMSE
rmse = accuracy.rmse(testset_pred)
print(f'RMSE: {rmse}')
```

The SVD model has a relatively low RMSE of 1.7839, which indicates the model is performing well.

```
In [ ]: # function to get top N movie recommendations for a user
        def get_top_n(predictions, n=10):
            top n = \{\}
            for uid, iid, true_r, est, _ in predictions:
                if uid not in top_n:
                    top_n[uid] = []
                top_n[uid].append((iid, est))
            # Sort the predictions for each user and get the top N
            for uid, user_ratings in top_n.items():
                user_ratings.sort(key=lambda x: x[1], reverse=True)
                top_n[uid] = user_ratings[:n]
            return top_n
In [ ]: # Get top N recommendations for each user
        top_n = get_top_n(testset_pred, n=10)
        # Evaluate Precision and Recall
        precision_sum = 0
        recall_sum = 0
        for uid, user_ratings in top_n.items():
            # Get the movies in the test set for the user
            actual_movies = [item[0] for item in testset if item[0] == uid]
            # Get the recommended movies for the user
            recommended_movies = [item[0] for item in user_ratings]
            # Calculate Precision and Recall
            precision = len(set(actual_movies) & set(recommended_movies)) / len(recommended_movies)
            recall = len(set(actual_movies) & set(recommended_movies)) / len(actual_movies)
            precision_sum += precision
            recall_sum += recall
        # Calculate average Precision and Recall
        average_precision = precision_sum / len(top_n)
        average_recall = recall_sum / len(top_n)
        print(f'Average Precision: {average precision}')
        print(f'Average Recall: {average_recall}')
       Average Precision: 0.7329
```

Average Recall: 0.7401

Model is ready to be used, as it obtains a decent average precision and recall on the test dataset.