

Hybrid Rating Prediction for Book Recommendation System

Learning Objective/ Goal

- 1) To predict book ratings for the given 'train.csv' using different approaches.
- 2) To use the book's metadata, learn and build a memory-based recommendation system listed in the [article](#)

Source code: <https://github.com/sudha-vijayakumar/CMPE256-BOOK-RECOMMENDATION-SYSTEM.git>

Input files (7)

- **input/metadata/** [goodreads_books_children.json.gz](#), [goodreads_reviews_children.json.gz](#)
- **input/** train.csv, test.csv
- **processed_data/** books.csv, reviews.csv, books_with_review.csv (will be generated at step-1,2)

Output (3)

- **output/** predictions-svd-75.csv, predictions_Fast_ai.csv, final_submission.csv, book_with_reviews.html, User_Ratings.html

HOW TO RUN | ML Pipeline (Run the notebooks in the below order)

No.	STEP	Filename
1	Read metadata as JSON	1_JSON_To_CSV.ipynb
2	Merge reviews with books	2_Merge_reviews_books_user_data.ipynb
3	EDA	3_EDA_using_Pandas.ipynb
4	Model Training <ol style="list-style-type: none"> 1) Identifying the best model for the train data <ol style="list-style-type: none"> a) AutoSurprise - SVD b) Fastai - Matrix Factorization 2) Train & make predictions for the test.csv using Surprise - SVD[Matrix Factorization] 	4.0_Auto_Surprise_BestModel.ipynb Run-time: limited to 6hours 4.1_Surprise-SVD_Book_Rating_Prediction.ipynb Run-time: 1hour 4.2_Fastai_Book_Rating_Prediction.ipynb Run-time: 50mins Submission: [0.7 * 4.1_(prediction)] + [0.3 * 4.2_(prediction)] 6_submission.ipynb
5	Recommendation System Collabarative filter - Memory-based - PearsonR	5_Recommendation_using_Colab_Filtering.ipynb

Learning outcome

- 1) Experimented with different approaches on how to predict ratings for the given dataset.
- 2) Built basic recommendation systems using the practical approaches taught in this course(Collaborative filtering using Pearson similarity).

Future work [To try HGN baseline for sequential recommendation](#) (Rank_1) [Source](#)