Recommender System

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Dataset - BookCrossing

Book rating datasets (3 tables: books, users, raitings) to build a recommendation engine



Merged data sets

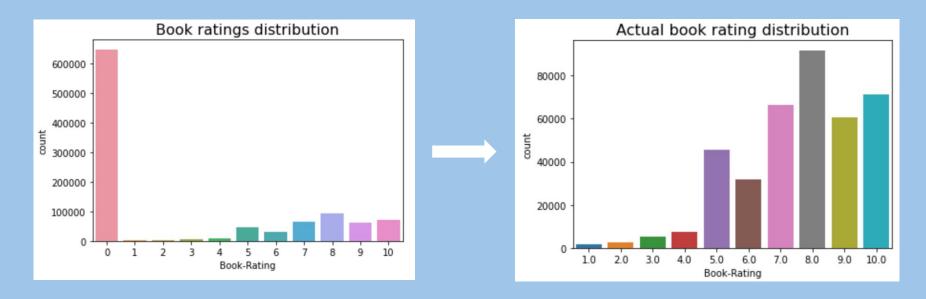
- Number of books: 270151
- Number of users: 92106
- Average book rating: 7.63

Variables

- ISBN
- Book-Title
- Book-Author
- Year-Of-Publication
- Publisher
- Image-URL-S
- Image-URL-M
- Image-URL-L
- User-ID
- Location
- Age
- Book-Rating

Missing or null values

- No mentioning of what 0 value means for rating
- Judging from the distribution => most likely 0 means missing value => drop these entries



Features engineering

- 1. Dropping Books with no review
- 2. Dropping images features
- 3. Imputing missing values (Age 30% of missing values)
- 4. Extract feature country from the user location
- 5. Extract sample of size n=8000

Dataset size

Ratings | Users | Books (inner join)

Drop null reviews

Most books have very few reviews & most users gave very few reviews, so we drop those with frequencies < 50

=> 1031136 entries

=> 383842 entries (very large for our resources)

=> 13137 entries (1079 users, 507 books, 13137 non-null reviews)

Recommendation tasks

Content Based

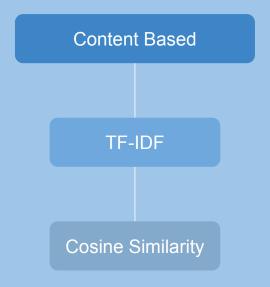
To build a recommender that suggests books based on similar book titles

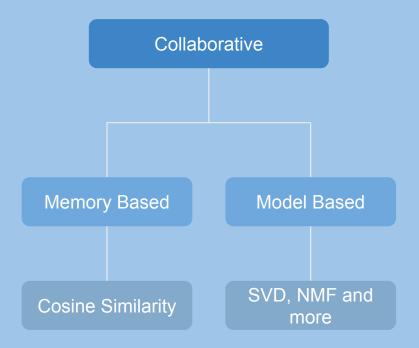
Collaborative: Memory Based

To build the recommender that suggest the books and other users based on the book grading

Collaborative Filtering: Model based

Recommender System





Test User



Age: 47

Location: Manchester, New

Hampshire, USA

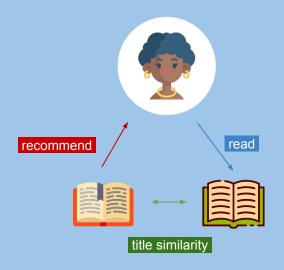


Age: 37

Location: Woodbridge,

Virgina, USA

Content Based



If a user has finished reading a book, which further readings can we recommend to this user based on the book title similarity?

Content Based - Feature Selection

Book Features

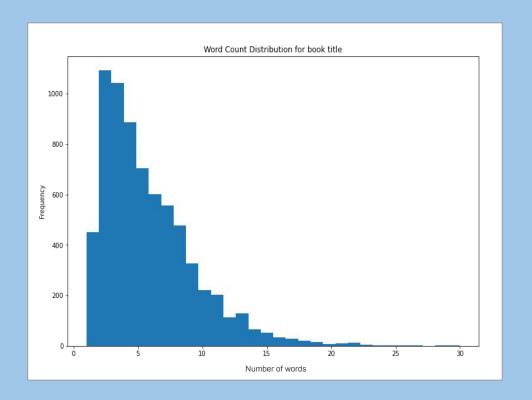
ISBN

Book-Title

Book-Author

Year-Of-Publication

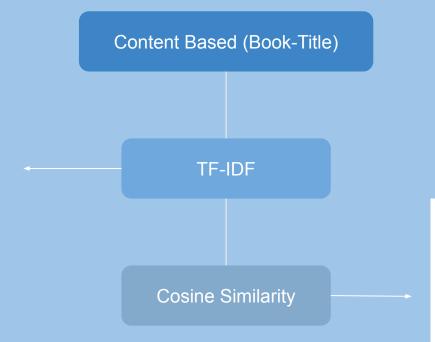
Publisher



Content Based

Scikit-learn: TfidfVectorizer

Transformation to convert title strings into vectors



Scikit-learn: cosine_similarity

Is a similarity metrics

Can be found by taking the dot product of the title vectors

Content Based - Test Case I



Initial Title: Her father house

Recommendation 1:

Ramona and Her Father (Avon Camelot Books (Paperback)) by Beverly Cleary with 0.267 similarity score

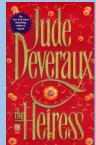
Recommendation 2:

The Heiress by Jude Deveraux with 0.0 similarity score

Percommendation 3:

Tangled Webs (Forgotten Realms: Starlight and Shadows, Book 2) by Elaine Cunningham with 0.0 similarity score







Content Based - Test Case II



Initial Title: Reunion (Star Trek The Next Generation)

Recommendation 1:

Crossover (Star Trek The Next Generation) by Michael Jan Friedman with 0.713 similarity score

Recommendation 2:

Intellivore (Star Trek: The Next Generation) by Diane Duane with 0.713 similarity score

Recommendation 3:

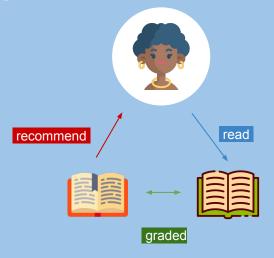
Imzadi (Star Trek: The Next Generation) by Peter David with 0.713 similarity score





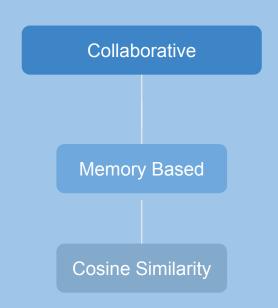


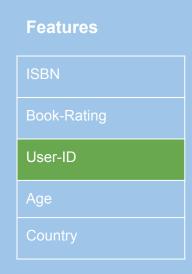
Collaborative: Memory Based

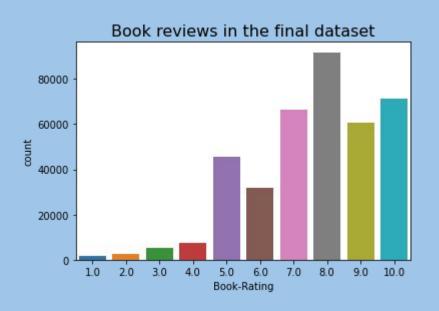


If a user has finished reading a book, which further readings can we recommend to this user based on the **book grading**?

Collaborative: Memory Based



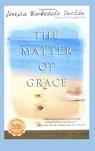


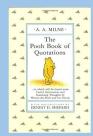


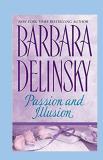
Memory Based - Test Case I



Rating based recommender







Memory Based - Similar users





3 user
Age: 27
Location: Lafayette,
indiana, USA



Age: 51 Location: Manchester, New Hampshire, USA



1 user Age: 39 Location: Burlington, Ontario, Canada

Memory Based - Test Case II



Rating based recommender







Memory Based - Similar users

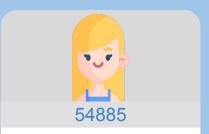




3 user Age: 46

Location: Oceanside,

California, USA



2 user Age: 30

Location: Astoria, New

York, USA



1 user Age: nan

Location: Houston,

Texas, USA

Collaborative Filtering Model based

Model-based CF: what it does

Name	Avengers	Star wars	Thor	Spider-man	Iron Man
Alex	4	2	?	5	4
Bob	5	3	4	?	3
Tom	3	?	4	4	3

Model-based CF

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Model-based Collaborative

Filtering

Matrix factorization algorithms

Single value decomposition (SVD)
Non-negative Matrix Factorization (NMF)
Slope One
Co Clustering

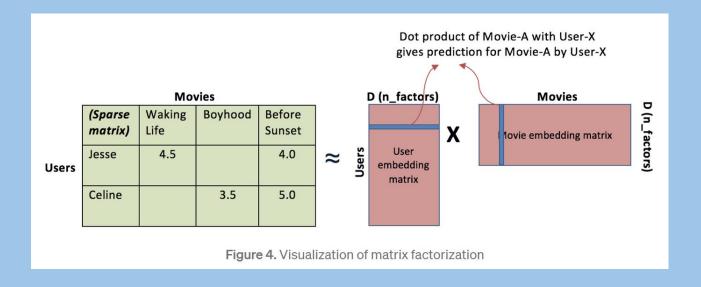
Deep Learning

Neural Networks

Model-based CF: matrix factorization

The idea: preference of user can be determined by a small number of hidden factors.

We can call these factors as **Embeddings**.



Models comparison

Surprise library -> http://surpriselib.com

3-fold cross-validation

Reporting RMSI

	RMSE (test)	Time (train)	Time (test)
Algorithm			
SVD	1.545178	0.650762	0.047195
SVDpp	1.546650	3.752617	0.183756
BaselineOnly	1.559600	0.018354	0.022904
CoClustering	1.656513	0.333316	0.035182
KNNWithMeans	1.682558	0.066995	0.195750
KNNBaseline	1.699507	0.068997	0.240620
KNNWithZScore	1.711343	0.096441	0.222861
KNNBasic	1.803811	0.046665	0.196623
SlopeOne	1.803843	0.037321	0.110750
NormalPredictor	2.338401	0.015554	0.036952
NMF	2.612697	0.798814	0.039230 24

SVD prediction

random_sample[['User-ID', 'ISBN', 'Book-Rating']]							
	User-ID int64	ISBN object	Book-Rating float64				
2215	184339	0061073628	9				
2582	214786	0679774386	nan				



user: 184339 item: 0061073628 r_ui = 9.00 est = 7.91 user: 214786 item: 0679774386 r_ui = nan est = 7.84