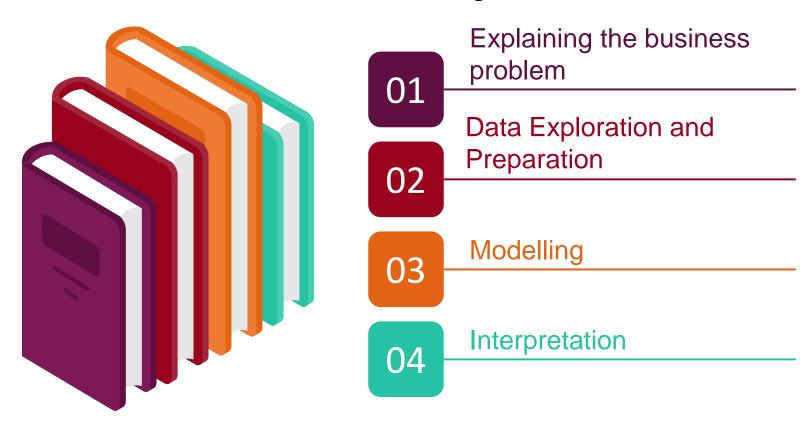
Giving the best book recommendations in a nutshell.





Business Problem



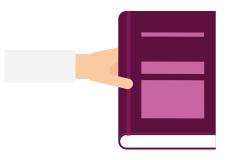
In a world in constant change and rapid growth, books have been losing their value. Therefore, it is important to apply strategies to increase their attractiveness (or, in a financial view, increase sales), to do that, a machine learning algorithm will be created to predict books you would actually might want to read.

Goal: Building a ranked list of book ratings

Target: Unknown ratings



3 Datasets



Users

- User-ID
- Location
- Age



Ratings

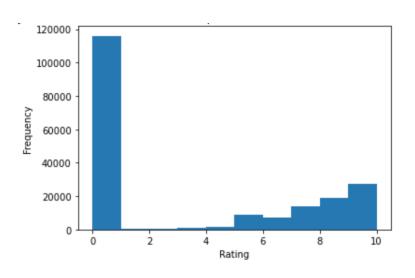
- User-ID
- ISBN
- · Book-Rating

Books

- ISBN
- Title
- Author
- Year of publication
- Publisher
- Image URLs (S, M, L)

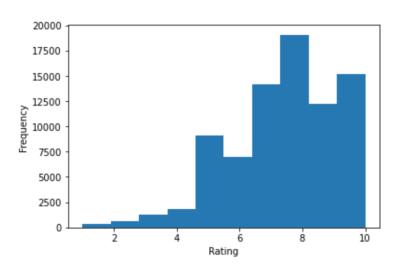


O1 Overall Ratings distribution

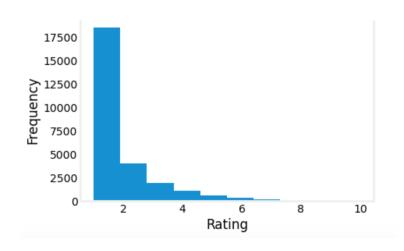


- 196,842 observations
- 116,036 zeros (users that bought books but gave no rating)
- Cannot use implicit and explicit ratings on modelling
- Drop the zeros and get 80,806 observations

Ratings distribution without implicit ratings

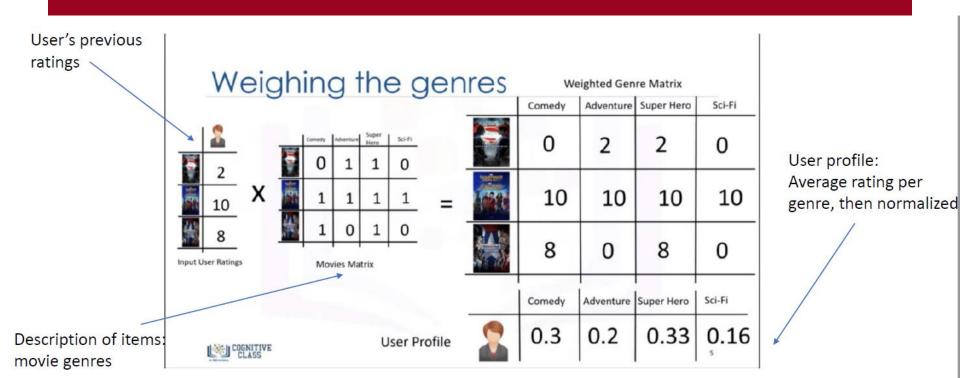


Ratings by user distribution

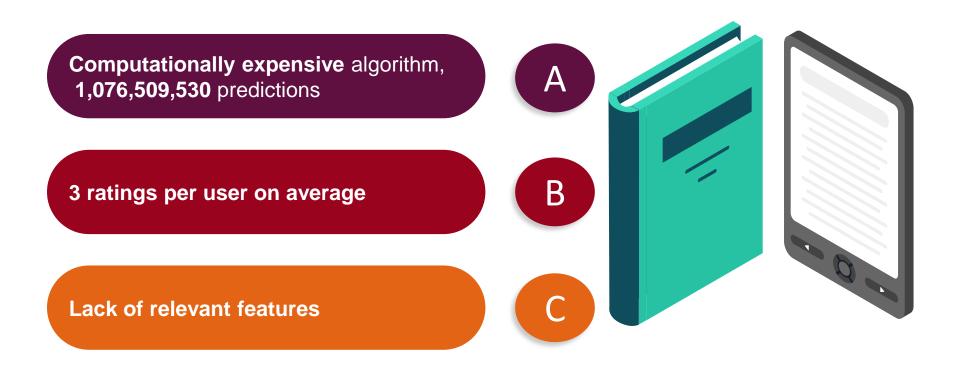


Modeling

Content-based Recommender System



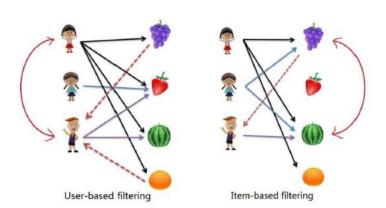
Content-based recommender system - Problems



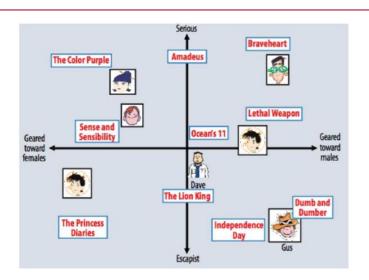
Modeling

Collaborative Filtering

Memory-based



Model-based



Results

01

Model-based Collaborative Filtering

02

RMSE

In-sample: 1.733

Out-of-sample: 1.707

03

Top 10 books

	Book	Predicted Rating
469	Harry Potter and the Order of the Phoenix (Boo	8.73
393	Harry Potter and the Sorcerer's Stone (Harry P	8.65
109	The Time Traveler's Wife	8.58
648	The Curious Incident of the Dog in the Night-T	8.57
1094	Charlotte's Web	8.54
478	Name Der Rose	8.45
2324	Illuminati.	8.40
67	Jane Eyre	8.39
260	The Mount: A Novel	8.38
380	Sister of My Heart	8.37

Interpretability

What are the underlying factors? We don't know

To get an idea: Plot books / users in feature space,

