

Lesson 16

Recommenders

Recommender Definition and Need

- Refers to a machine learning (ML) tool that enables recommendations after extractions of features in the item sets from the multiple datasets
- A recommender's need arises because a customer finds it difficult to search the right product due to information overload.

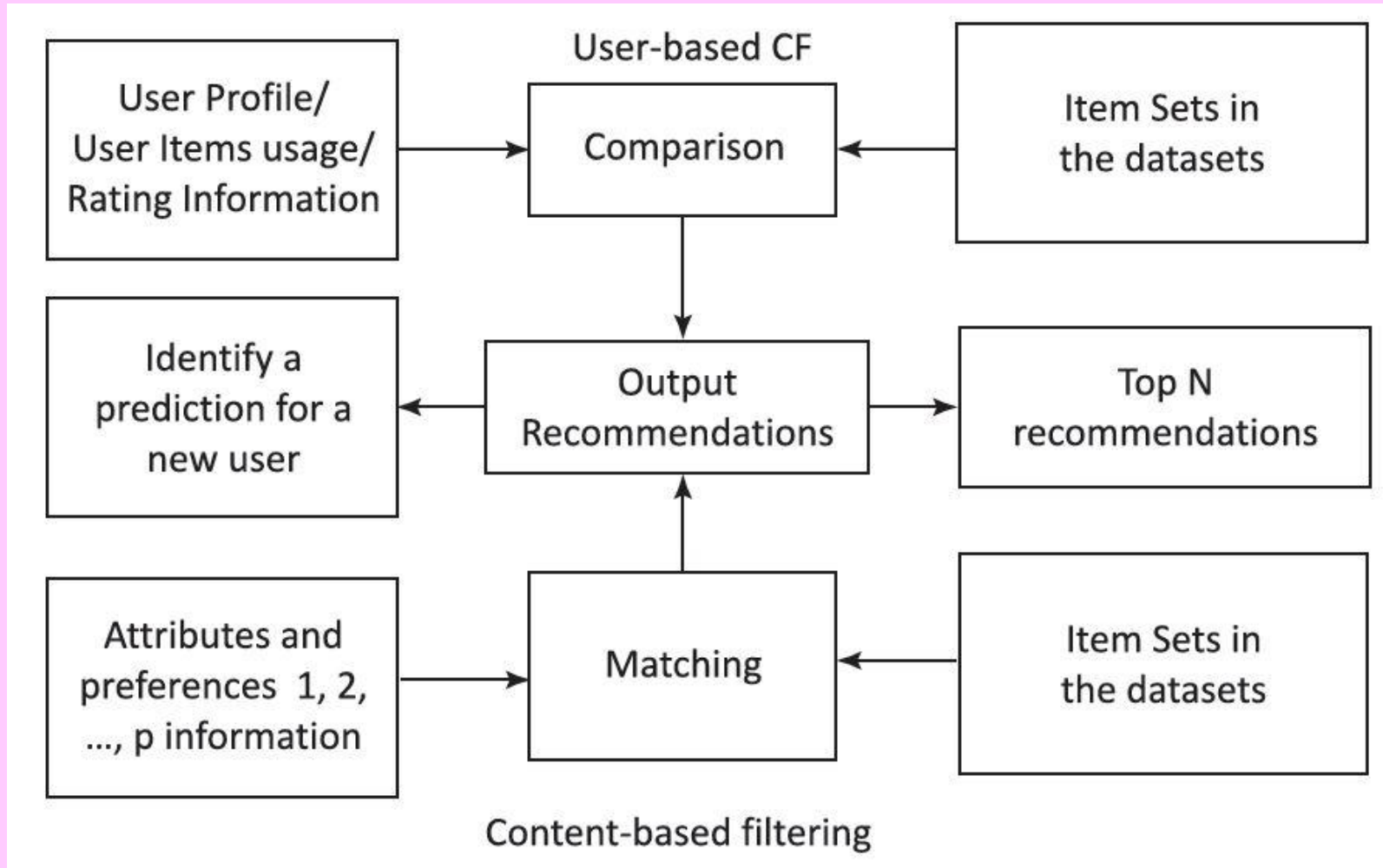
Recommender Examples

- A product recommender, such as book recommender
- Enabler for the selection of the right product among the high recommendation products.

Recommenders

- User-based collaborative filtering (CF)
- Content-based filtering (CBF)
recommenders)

Recommender CF and CBF Methods



CF Algorithm

- Makes recommendations by calculating the similarities in item sets in-between different items in the datasets
- Algorithm predicts the likeliness of an item that it has not rated based on a set of historical preference judgments from a community of users
- Relies on the behavior of users

Recommending Method using the CF

- Group forms for similar users which means in neighbourhoods
- Similar means not distant
- Recommender is an application of near-neighbour search method

Recommending Method using the CF

- The already rated items are significant in searching for a neighbour
- Once a neighbour of a user is found, different algorithms used
- Combine the preference of neighbours to generate recommendations

Top-N Recommendations

- Recommends a set of N top-ranked items that will be of interest to a certain user
- Techniques analyze the user-item matrix to discover relations between different users or items and make Top- N recommendations

User-Item Matrix Input Data Example

- Input Data: User, Item, Rating
- For example, (1, 206, 4.0), (2, 256, 5.0), (3, 106, 3.0), (4, 56, 2.0), (5, 206, 4.5), (1, 206, 4.0), (1, 206, 4.0), ...
- Data dynamically built ups the Matrix and updates the rating in the existing matrix

CF Recommendation Approaches

- Model-based collaborative filtering –
Based on ML techniques (Section 6.8.2)
- Memory-based collaborative filtering –
Based on similarity between users and items

S

s= 1

User-Based CF Algorithm for Top-N Recommendations

- Identify the k most similar users (nearest neighbours) to the active user using the Pearson correlation or vector space model
- User as a vector in the v -dimensional item space

Item-based Collaborative Filtering

- Item-based similarity between items is more static compared to dynamic Users based CF
- Enables pre-computation which improves online performance
- An item-based similarity is more meaningful

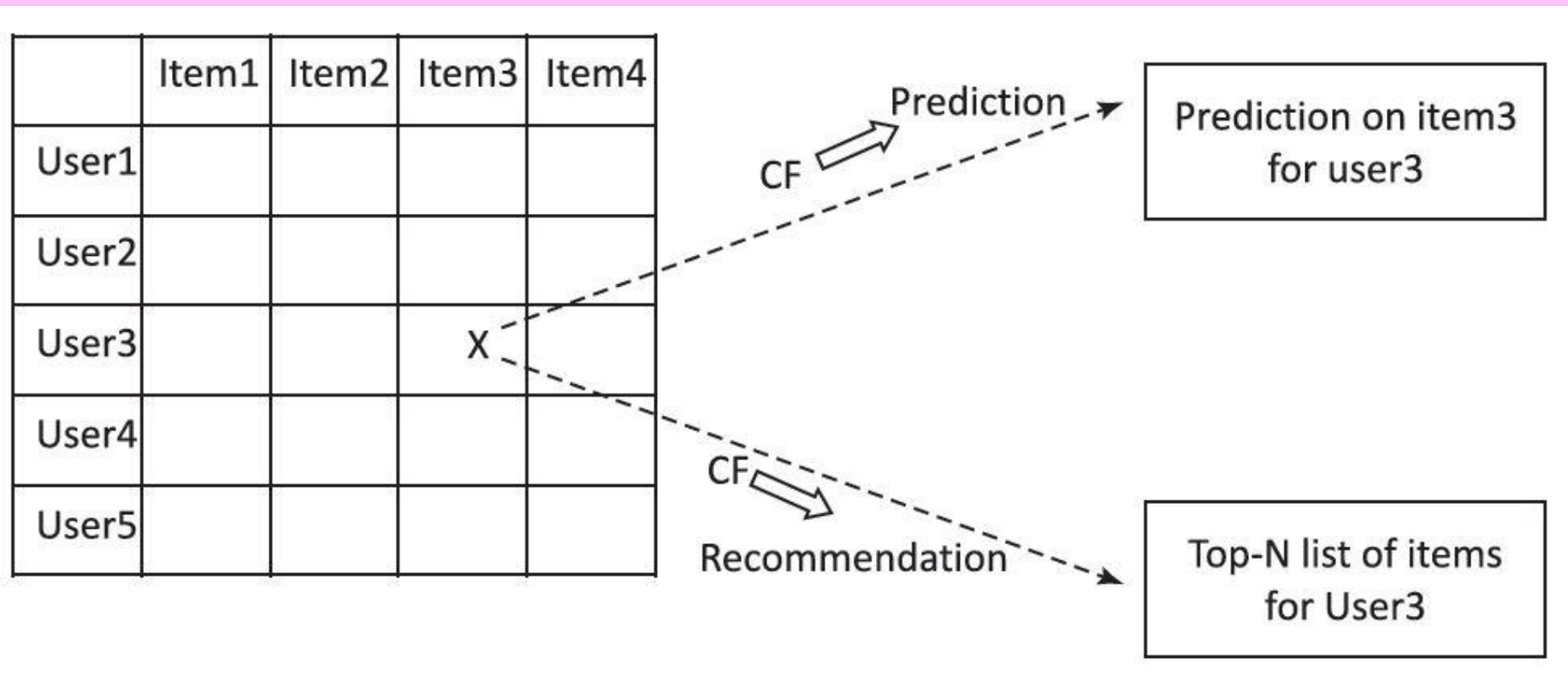
The similarity between two users of items

- Measured by treating each user as a vector of rating frequencies and computing the cosine of the angle formed by the frequency vectors

Using an $m \times n$ user-item matrix, P

- Similarity between two items j and i defined as the cosine of the angles between v -dimensional vectors corresponding to the i -th and j -th columns of matrix P .
- [Equation (6.23b)]

Figure 6.20 Prediction on an item and top 10 recommendation list from a user-item matrix



Model building for recommender approach

- A machine learning approach
- Recommend a set of items using pre-computed models
- Use the association rules, regression, clustering, neural network, Bayesian classifiers, decision trees and matrix completion techniques

Content Based Filtering (CBF) Recommendation

- Built around the attributes and preferences of a given item
- Evaluates the contents or items on the preferences of others with a similar point of view (Figure 6.19)

Content Based Filtering (CBF) Recommendation

- When anyone buys a product online, most websites automatically recommend other products that she/he may be interested to buy

Knowledge-based Filtering (KBF) Recommender

- Built on explicit knowledge about user preferences, the number of items of specific attributes and a criterion function for the recommendation
- Positive recommendation criterion for the given context.

KBF Advantage

- Advantage of KBF recommender is that it applies in scenarios where CF and CBF have difficulties in uses, for example, due to cold start problem
- Applies in complex domains, such as house recommender

KBF Disadvantage

- Problem associated with acquiring difficulties for knowledge
- A definitive recommendation in explicit form difficult with partial knowledge

Hybrid Filtering

- Refers to combination of the CF and CBF

Recommender Applications

- E-commerce, Product recommender – Amazon
- Digital Libraries, Pages/ documents/ books recommendation
- Enterprise Activity Intelligence
- Domain expert recommender, ...
- Personal Assistance, for example, for Museum guidance, ...

Summary

We learnt:

- Collaborative filtering Recommender
- Content based recommender
- User-Item Matrix for predicting
- Top-N recommendations
- Knowledge based Recommender
- Applications

End of Lesson 16 on **Recommenders**