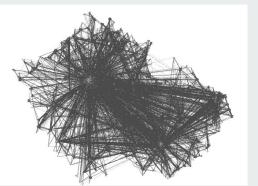


## **Recommender Systems**



STS 351: Minds and Machines
Research Presentation
Kimberly Prince, Chadin Youssef, Tyreek Griffith



## Overview | Introduction to Recommender Systems

#### What and Why:

- ☐ Algorithms used to filter predictability of information based on user preferences
- Most commercially utilized application for machine learning
- Dominated by big businesses

#### Design approaches used:

- Content based filtering relies on the contents of the targeted object
- □ Collaborative filtering relies on users responses to the targeted object

## **Collaborative Filtering Explored**

**How it works:** Collect & analyze large sum of information on users' behaviors and preferences; predictive model based on what other users with similar feature set like

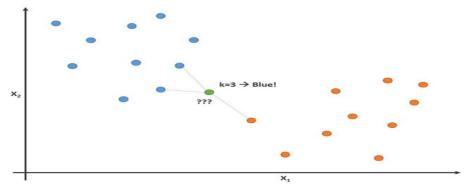
Does not rely on machine analyzable content

People who agree in the past will agree in the future and they will like similar

things

#### Types of Collaborative Filtering Algo's:

- User-User collaborative filtering
- Item-Item collaborative filtering
   (Items calculated using people's ratings



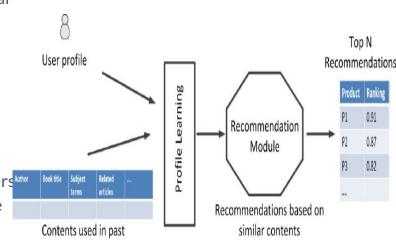
## **Content Filtering Explored**

How it works: Based on a description of the item AND a profile of the user. Keywords are used to describe items and user profiles are built to indicate which item a user likes. Algorithms recommended items similar to items user liked in the past.

To create a user profile the system focuses on:

- Model of user preferences
- History of user interactions with recommender system

Item profile is created and system creates content based profile of users based on weighted vector of item features. The weights determine the importance of each feature to the user.



## **Brief | Connection to Western Philosophy**

Why is there such a strong connection or attraction to user preferences or recommendations?

Agency

Qualia

Aristotle: Telos and Rational

Thus, inviting technology and devices that service these human habits, behaviors, and desires!

## Goals | Recommender Systems in Work

#### **Primary Models**

- Prediction version of problem
- Ranking version of problem

#### **Operational and Technical Goals**

- → Relevance
- Novelty
- Serendipity
- ☐ Increasing recommendation diversity

## Research Literature | Recommender Systems

What are popular research topics in the academic disciplines on recommender systems?

Communication technologies (Social Systems or Networks) or Business Transactions

What are scientific magazines, public online publications, and sites saying about recommender systems?

- ☐ Harvard Business Review Indicator of Company Leaders
- ☐ Companies Distributing Services Webtunix Services
- ☐ Pew Research Center ex. YouTube Users

What are some issues or controversies that come with recommender systems?

- Consumer Use and Trust | Target Example
- Data Mining

# Users and Items | Large-scale Recommender Systems

#### **Hybrid Recommender Systems**

- Uses a combination of content based filters and collaborative filtering
- ☐ Is dependant on the features of the targeted object as well as the user's preferences
- ☐ Typically used in big businesses

#### **Prediction is ranking**

Given the user and context - ranks in order of the likelihood the user is to interact with

## Data Collection | How Do We Collect Data?

Core data is found through events:

Implicit:

Online - page view, app interaction

Commerce - cart, purchase, return

Media - preview, watch, listen

- ☐ Explicit: Ratings, Reviews
- ☐ Intent: Search query
- ☐ Social: Likes, shares, follows

Content is attached to each event

## Rec. Systems in Use | The Start Up Problem

"Cold start" - unknown data about a user or an item

#### **New Items:**

- No interaction data
- Uses baselines or item content

#### **New Users:**

- No historical interaction data
- ☐ Limited context data
- ☐ Unable to use collaborative filtering models

Must turn to Deep Learning!

## Flaws in Recommender Systems

Sparsity

Scalability

**Privacy Protection** 

Over-specialisation

"Gray sheep" Problem

Shilling Attacks

## Future | Innovation and Improvements on Recommender Systems

Researchers and technologists goals for recommender systems in the future

- ☐ Dimensionality reduction techniques
- Improve privacy
- Prevention of large variances in data

Where | How do we think Recommender Systems will be implemented in the future:

- Kimberly
- ☐ Chadin
- □ Tyreek

## **Recommender Systems | Question Section**

**Questions from the Student Audience** 



#### Top Three Questions for the Student Audience

- Should the government monitor and be alert to user history from recommender systems? (Reviewing popular search terms, file downloads, etc.)
- Do you think it is ethical for companies and businesses to collect and use data from recommender systems?
- ☐ Where do you think recommender systems will be in the future?

### **Thank You for Your Attention**

## Recommender Systems | Bibliography

#### **MLA Citations**

Bakhtiyari, Kaveh. "The Future Directions of Recommender Systems." *Medium*, Futurist One, 23 June 2018, medium.com/futuristone/the-future-directions-of-recomme nder-systems-72beae7c2dd2.

Sharma, Richa & Singh, Rahul. (2016). Evolution of Recommender Systems from Ancient Times to Modern Era: A Survey. Indian Journal of Science and Technology. 9. 10.17485/ijst/2016/v9i20/88005

Scassellati, Brian. "Recommender Systems." *YouTube*, Harvard University, 10 Nov. 2015, www.youtube.com/watch?v=Eeq1DEeWUjA.

#### **MLA Citations**

Aggarwal, Charu C. *Recommender Systems: The Textbook.* Cham: Springer, 2016., 2016

Schrage, Michael. *Great Digital Companies Build Great Recommendation Engines*, Harvard Business Review - Technology, 1 Aug. 2017, hbr.org/2017/08/great-digital-companies-build-great-recommendation-engines.

Smith, Aaron. 7 Things We've Learned About Computer Algorithms, Pew Research Center - Fact Tan, News in the Numbers, 13 Feb. 2019, www.pewresearch.org/fact-tank/2019/02/13/7-things-weve-learn ed-about-computer-algorithms/.

## Recommender Systems | Bibliography

#### **MLA Citations**

Al Powered Recommendation as a Service, Webtunix, 2019, www.webtunix.com/ai-powered-recommendation-as-a-service.

Anandhan, A. (. 1. )., et al. "Social Media Recommender Systems: Review and Open Research Issues." *IEEE Access*, vol. 6, pp. 15608–15628.