

# Trust Aware Recommendation Systems



# Recommendation Systems

- Solution to information overload problem.
- Try to find items such as books or movies that match best with users' preferences.
- Based on the approaches recommendation systems can be classified into two major groups:
  - Content-based Recommendation.
  - Collaborative Filtering.

# How They Work

## Content-based

- Use Items' features and characteristics to rank the items based on the user's preferences.

## Collaborative Filtering

- Rely on user's past behavior e.g., purchases or ratings, to find similar users or items and utilize this information in order to find items of interest to the user.

# Problems

## Cold Start

- A situation where the algorithm's effectiveness is very low because items' (or users') vector do not have enough rated items to find vectors similar to them.

## Attacks

- Designed to drive recommender system to act in a way that the attacker wishes.
- Either recommend some desired items or prevent recommendation of other items.
- Set of attack profiles each containing biased rating data associated with fictitious user.

Solution:

Trust Aware Recommendation  
Systems

Users are influenced by their trustworthy friends, and are more likely to accept recommendations made by them.



# Overview

- Trust-aware recommender systems employ trust information to enhance recommender systems.
- Merging trust information and recommender systems can improve the accuracy of the recommender system as well as users' experience.
- They are also capable of handling some challenges of classical recommender systems such as cold-start and responding to attacks.

# Deliverables

Implement some of the state-of-the-art trust aware recommendation systems and improve them thereafter.



### Co-factorization Methods:

SoRec: social recommendation using probabilistic matrix factorization. In Proceeding of the 17th ACM Conference on Information and Knowledge Management , pages 931–940. ACM, 2008. DOI: 10.1145/1458082.1458205.

LOCABAL: Exploiting local and global social context for recommendation. In Proceedings of the Twenty-Third International Joint Conference on Artificial Intelligence , pages 2712–2718. AAAI Press, 2013.

### Ensemble Methods:

STE: Learning to recommend with social trust ensemble. In Proceedings of the 32nd International ACM SIGIR Conference on Research and Development in Information Retrieval , pages 203–210. ACM, 2009. DOI: 10.1145/1571941.1571978.

### Regularization Methods:

SocialMF: A matrix factorization technique with trust propagation for recommendation in social networks. In Proceedings of the Fourth ACM Conference on Recommender Systems , pages 135–142. ACM, 2010. DOI: 10.1145/1864708.1864736

6nd responding to attacks.

# Thanks!

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