# Item-Item Collaborative Filtering via Graph Regularization

By: Melle Koper (M.M.M.KoperookGeschrevenJansen@student.tudelft.nl) Supervisors: Maosheng Yang and Elvin Isufi



# Background

- Recommender System  $\rightarrow$  Predict how a user would rate an item
- Simple Methods > Machine Learning [1]
- Graph Regularization predict data from noisy graph signals [2]

# **Research Question**

- How do Item-Item graph Regularizations techniques compare against a baseline?
  - Tikhonov Collaborative Filtering
  - Sobolev Collaborative Filtering

### Collaborative Filtering

- Similarity Matrix
  - Pearson Correlation
- K Nearest Neighbors
- Prediction = Weighted Average of Neighboring items

## Graph Regularization





Tikhonov Sobolev  $X^* = (I + \mu L)^{-1} Y$  $\boldsymbol{X^*} = (\boldsymbol{I} + \boldsymbol{\mu} * (\boldsymbol{L} + \boldsymbol{\epsilon} * \boldsymbol{I})^{\beta})^{-1} \boldsymbol{Y}$ 

L = Degree Matrix - Similarity Matrix

# Methodology

#### Setup

Similarity Matrix

0

0

Test Data = Moviel ens100k 80-20 Train Test split Hyper Parameter Tuning = GridSearch Compare on best mean results over 5 rounds

## Metrics

- Root Mean Squared Error
- Precision
- Recall
- Normalized Discounted Cumulative Gain

### Hyper Parameter Search + Results



Table 1: Best mean results achieved over 5 rounds

# Conclusion + Future Work

- Graph Regularization outperforms the baseline
- Expand on hyperparameter testing for Sobolev
- Use more sophisticated techniques for creating the similarity matrix.

#### References:

[1] M. Ferrari Dacrema, P. Cremonesi, and D. Jannach, "Are we really making much progress? a worrving analysis of recent neural recommendation approaches." 09

2019.

[2] E. Isufi, B. Das, A. Natali, M. Sabbaqi, and M. Yang, "Graph filters for processing and learning from network data," tech. rep., Delft University of Technology, 2021.