Topic: Synthesis and Evaluation of various Recommender System approaches against a Movie database dataset

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# Abstract

The idea of this thesis is to implement and evaluate multiple recommender system categories against the publicly available MovieLens dataset. Primary objective is to see how using various implementations and algorithms affects the recommendations and end-user experience.

We will be using both the Collaborative filtering algorithms (both neighborhood and model-based) and Content-based recommender system implementations.

For evaluation metrics, the main focus will be on comparison between predicted and actual outcomes. Nonetheless, an attempt will be made in order to evaluate predicted users experience.

The outcome of this thesis should be a clear overview of which implementation brings the best results and what steps could be taken to further enhance the outcome.

# Recommender Systems

## Introduction

Recommender systems have evolved as a natural response to ever growing amount of information available to wide audiences of users. With the rise of Internet and general availability of it to common users, it has become necessary to help users navigate the content and steer them towards the new options they might have not anticipated in the first place.

Strictly technically speaking, Recommender systems represent a subclass of Information filtering systems whose main purpose is to predict the “ratings” or “preferences” a user would give to an item [1].

## History

There does not seem to be an official information on when the recommender systems have been mentioned for the first time.

One of the earliest mentions of Recommender systems as such dates to early 1990’s where the official term has been mentioned in a technical report written by Jussi Karlgren at Columbia University [2].

Since then, especially with the rise of Cloud computing and Big data, when processing of vast amounts of data became possible, recommender systems have become an important aspect and indispensable commodity of all successful businesses.

It is especially important to mention the efforts of GroupLens working group [3]. GroupLens is a group of scientists from the department of computer science and engineering in University of Minnesota. This group has pushed forward the efforts towards both publishing various research papers on recommender systems and publishing the freely available datasets to be analyzed and used for educational and research purposes. Some of their featured projects include MovieLens – a web site that helps people find movies to watch, Cyclopath – an editable map where anyone can find maps and routes for riding bicycle and LensKit – open source toolkit for building, researching and studying recommender systems. MovieLens is also the dataset that this thesis is using for evaluation of results.

## Overview

It has been stated before that one of the main driving factors for introduction of recommender systems was development of Internet and World wide web. Once the data became publicly available and many users gained access to it, there was a need to help users browse the content and steer their attention towards the items that are curated for them.

Another highly important catalyst that has driven this development is the ease with which users are able to express their preferences. Namely, today, users are able to demonstrate their liking or disliking of a certain product by a single click of mouse. This one second of users feedback multiplied by number of users and amount of items they interact with, results in enormous volumes of data which are then used to recommend even more fine-grained items. This loop repeats itself.

Charu C. Aggarwal in his Recommender systems book [4] takes a Netflix as an example. He suggests to take content providers, such as Netflix, as an example. *“In such cases”*, he further states, *“users are able to easily provide feedback with a simple click of a mouse. A typical methodology to provide feedback is in the form of ratings, in which users select numerical values from a specific evaluation system (e.g., five-star rating system) that specify their likes and dislikes of various items.”.*

The aforementioned approaches for feedback collection are usually referred to as “Explicit ratings” or “Explicit feedback”. This naming stems from the fact that the rating or the feedback of the item being recommended was explicitly specified by user. This also means that user is most likely consent to share his preferences with the service provider and is looking forward towards getting more curated content.

In contrast to explicit ratings, there is a group of so called “Implicit ratings” or “Implicit feedback”. Namely, this is the kind of feedback that can be derived based on the actions and behavior of user in question. Perfect example of such scenario is Amazon.com. As Aggarwal [4] states, a simple act of a user buying or browsing an item may be viewed as an endorsement for that item. Another example would be a YouTube.com. Having a user watch the video from beginning to an end is a perfect example of a positive feedback where users expresses his interest towards the item being watched. On the other hand, user skimming through the video being played is an indirect act of providing negative feedback. This negative feedback should be treated with caution as it can happen that user is not interested only at that moment, but might be looking forward to interacting with same item on another occasion. It is up to the designer of system to make sure that the feedback is evaluated properly and all relevant things are being taken into consideration.

## Netflix, Last.fm and Pandora

Write a bit about how recommender systems were/are used at these

## Recommender systems and Big data

Search for some work in this area and write a bit about it …

## Future of Recommender systems

Write a bit about how we see the future of Recommender systems, etc.

# Methodology

Write a bit about the data set that will be used. Show some data, etc.

Write about how the results will be evaluated.

# Bibliography

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