

CmpE493 - Assignment 2

A Movie Recommendation System

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1 Description

Building a recommendation system is a common task in many modern applications. The goal of a recommendation system is to identify relevant data for their users. These systems are generally based on two methods: collaborative filtering and content based filtering. While collaborative filtering exploits similar users' rates, content based filtering considers the contents of the items liked by the same user.

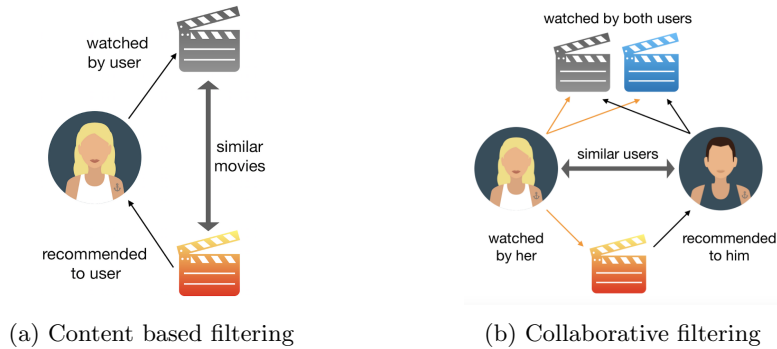


Figure 1: Recommendation systems ¹

In this assignment, you will build a simple movie recommendation system from scratch. Firstly, you will extract contents and recommendations of movies from IMDB. Then, to represent the contents of movies, you will implement the vector-space model based on TF-IDF weighting. After vectorizing all movies' contents using TF-IDF, you will make recommendations for a given movie by getting the most similar K movies based on cosine similarity. Finally, you will calculate the evaluation metrics considering the IMDB recommendations as the relevant (ground truth) movies. The IMDB ids for the movies and a notebook template that you must follow are available on Moodle. The libraries you are allowed to use for this assignment are *requests*² and the standard Python libraries.

¹Source: <https://bit.ly/2Anc0E4>

²Requests is a library to make HTTP requests.

2 Implementation

2.1 Extracting contents from IMDB

For a given IMDB ID, the web page for the corresponding movie can be found at <https://www.imdb.com/title/<imdb-id>> as shown in the yellow box in Figure 2.

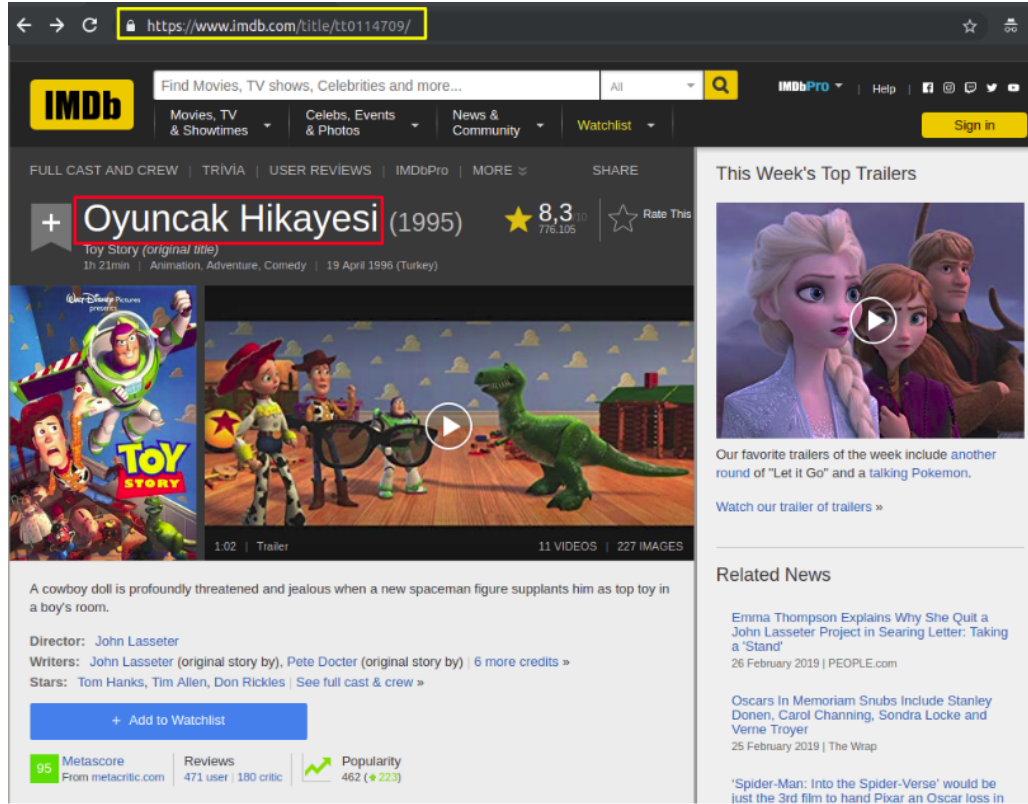


Figure 2: IMDB web page for Toy Story

The data which must be collected for each movie are:

1. Title of a movie (seen in a red box in Figure 2)
2. Storyline of a movie (seen in a red box in Figure 3)
3. IMDB ids of all recommended movies³ (seen in a red box in Figure 4)

³Not only the ones seen in the current panel, but also those in the next panels.

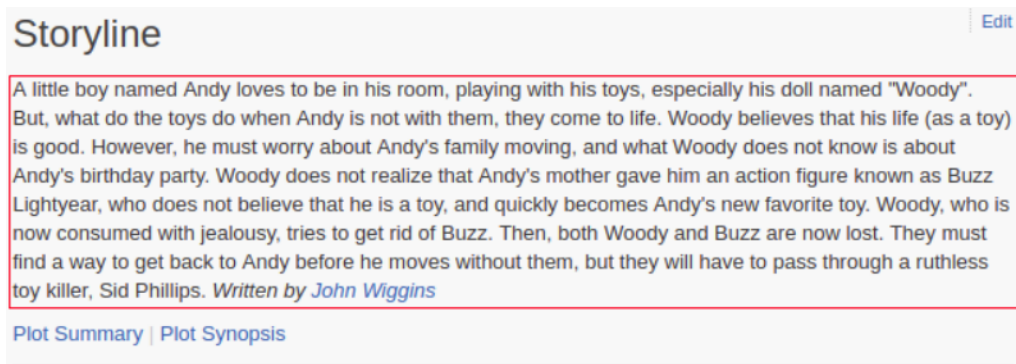


Figure 3: Storyline for Toy Story

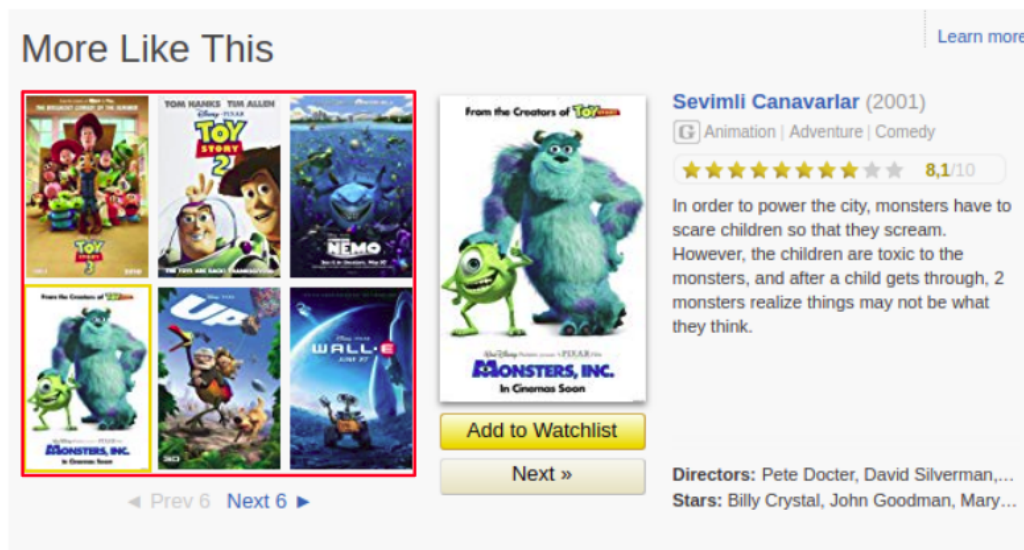


Figure 4: Recommendations for Toy Story

2.2 TF-IDF Model

To represent the storylines of movies in a vector space, you will implement the TF-IDF based vector space model. Firstly, you will process the storylines of movies, identify the vocabulary, as well as the term and inverse document frequencies. Then, you will get the words with highest TF-IDF scores⁴ and encode each movie's storyline by using the occurrences and scores of these words.

⁴You can either set a threshold for TF-IDF score or simply take the highest scoring N words. Explain clearly how you decide it in your notebook.

2.3 Recommendation

To recommend movies, a simple content based strategy will be followed. Given a movie, the system will recommend the most similar K movies based on cosine similarity.

2.4 Evaluation Metrics

To evaluate the system, precision, recall and F1 scores must be calculated for $K = 1, 2, 3, 10$ when recommendations are requested for a movie. The definitions of these metrics in this setting are as follows:

- **Precision** is the fraction of IMDB recommendations among your system's recommendations.
- **Recall** is the fraction of your system's recommendations among the IMDB recommendations.
- **F1 score** is the harmonic average of the precision and recall.

Figure 5 might help to understand what these metrics represents. The search results in this figure corresponds to the recommendations of your system, while the relevant documents corresponds to the IMDB recommendations.⁵

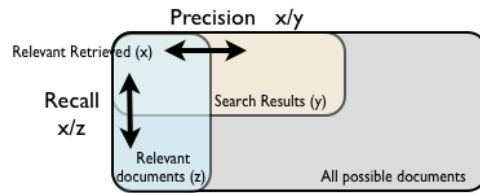


Figure 5: Precision and Recall⁶

⁵Nice post explaining metrics for recommendation: <http://sdsawtelle.github.io/blog/output/mean-average-precision-MAP-for-recommender-systems.html>

⁶Source: <http://aimotion.blogspot.com/2011/05/evaluating-recommender-systems.html>

3 Submission & Grading

You are expected to submit a single .ipynb file that is runnable. Name your notebook with your student id (e.g. 2018700100.ipynb) and do not write your name inside the notebook or anywhere else for the sake of blind review.

Note that this notebook will be your report as well, so explain your work in the related sections of the notebook.

1. IMDB Scraping: 30 pts
2. Recommendation: 35 pts
3. Evaluation Metrics: 20 pts
4. Notebook Organization & Explanations: 15 pts
5. Bonus⁷: 10 pts

Late Submission

You are allowed a total of 5 late days on homeworks with no late penalties applied. You can use these 5 days as you wish (e.g., 2 days for homework 1, 2 days for homework 2, and 1 day for homework 3). After using these 5 extra days, 10 points will be deducted for each late day.

⁷You are encouraged to improve the recommendation system by getting more valuable data while scraping and integrating them to the system or enhancing the models in the pipeline.