# CS-E4650 Methods of Data mining

# Exercise 4 / Autumn 2023

### 4.1 PageRank and HITS

Learning goal: The idea of PageRank and HITS algorithms.

Figure 1 shows the linkage structure of web pages and Table 1 lists the keywords that occur in the pages. The task is to evaluate PageRank and hubs and authority values of pages given a query. In this task, you can use any of the existing PageRank and HITS (simulation) tools (you can find also online calculators). Note that different tools may use different initialization or scaling but the top results should be the same. If the tool allows you to adjust the teleportation probability, use 0.10 or 0.15.

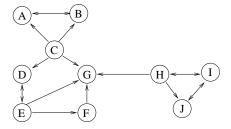


Figure 1: Linkage structure of web pages.

- a) Evaluate PageRank values for all pages. What would be the most reputable sources containing query words i) "PageRank" or ii) "teleportation"?
- b Construct the HITS graph (base set and its edges) for query "Page-Rank". Include pages in the root set, all pages pointed by the root pages and all pages pointing to the root pages. Then calculate the hubs and authority values. Which page is the best authority on the topic? What about the best hub?
- c) Repeat b) for query "teleportation".
- d) Compare the results with PageRank and HITS. Do they agree on the most reputable sources?

Table 1: Keywords that occur in pages A–J.

id	keywords
	· ·
A	authority, page, reputable, source
В	hub, page, link, good, source
С	PageRank, HITS, ranking, algorithm
D	reputable, page, link, PageRank
$\mathbf{E}$	reputation, visit, frequency, random, surfer
F	random, surfer, trap, dead-end
G	PageRank, teleportation, random, surfer, model
Η	teleportation, travel, planet
Ι	Star Trek, transporter, teleportation
J	beam, Scotty, transporter

### 4.2 Collaborative filtering for movie recommendations

Learning goal: How to use neighbourhood-based collaborative filtering in recommender systems; problems of adjusted cosine similarity.

Table 2 presents movie ratings by 6 users on 6 movies. The latex source of the table is available on the course page (mratingstable.tex). The ratings are between 1 (didn't like at all) to 5 (fantastic movie) and 0 means a missing rating (the user hasn't watched the movie). The users are notated  $u1, \ldots, u6$  and movies  $m1, \ldots, m6$ . The task is to apply recommender systems for rating prediction using neighbourhood-based collaborative filtering (see Aggarwal 18.5.2 and an example in the lecture).

- a) Calculate mean ratings per user. Use all non-missing ratings in the calculation. These are needed in parts b) and c).
- b) Calculate required pairwise similarities between users<sup>1</sup> using a modified Pearson correlation r ("Pearson" in Aggarwal Equation 18.12). Use the mean values calculated in part a. Remember that the correlation is calculated only over co-rated movies.
- c) Predict missing ratings using two nearest neighbours (K = 2) and an extra requirement that the similarity is  $r \ge 0.5$ . Tell if the movie is recommended to the user (if the user would like it more than average). Report if some prediction cannot be made (not enough sufficiently similar neighbours with required ratings).

<sup>&</sup>lt;sup>1</sup>Note: similarity between u2 and u3 is not needed, so 14 similarities.

d) Consider the item-based way of predicting the missing ratings of movies m3 and m4 with adjusted cosine similarity, as suggested in Aggarwal 18.5.2.2. Why it is not a good solution here? Suggest an alternative item-based solution that could be used instead (no need to calculate the actual predictions).

Table 2: Movie ratings (scale 1–5) by 6 users (u1-u6) on 6 movies (m1-m6). Special value 0 means a missing rating.

	O	. 0			
m1	m2	m3	m4	m5	m6
3	1	2	2	0	2
4	2	3	3	4	2
4	1	3	3	2	5
0	3	4	4	5	0
2	5	5	0	3	3
1	4	0	5	0	0
	3 4 4 0	3 1 4 2 4 1 0 3	$\begin{array}{c cccc} m1 & m2 & m3 \\ \hline 3 & 1 & 2 \\ 4 & 2 & 3 \\ 4 & 1 & 3 \\ 0 & 3 & 4 \\ \end{array}$	$\begin{array}{c ccccc} m1 & m2 & m3 & m4 \\ \hline 3 & 1 & 2 & 2 \\ 4 & 2 & 3 & 3 \\ 4 & 1 & 3 & 3 \\ 0 & 3 & 4 & 4 \\ 2 & 5 & 5 & 0 \\ \hline \end{array}$	$\begin{array}{c ccccccccccccccccccccccccccccccccccc$

#### 4.3 Distances between molecular structures

Learning goal: The concept of maximum common subgraph (MCS) and related distance measures.

Figure 2 shows an example of four molecular graph structures: Niacin (vitamin B1), Nicotine (active ingredient in tobacco), psilocin (active ingredient in "magic mushrooms"), and proline (amino acid). The node labels correspond to atoms (carbon, oxygen or nitrogen)<sup>2</sup>.

- a) Determine the nearest neighbour for each molecule using Union-normalized MCS distance (*Udist* in slides, see Aggarwal Eq. 17.2).
- b) Determine the nearest neighbour for each molecule using Max-normalized MCS distance (*Mdist* in slides, see Aggarwal Eq. 17.3).
- c) Under which conditions are *Udist* and *Mdist* equivalent? I.e., give conditions related to some graphs  $G_1$ ,  $G_2$  and  $MCS(G_1, G_2)$  such that  $Udist(G_1, G_2) = Mdist(G_1, G_2)$ .

<sup>&</sup>lt;sup>2</sup>For simplicity hydrogen atoms and double bonds between atoms are not presented.

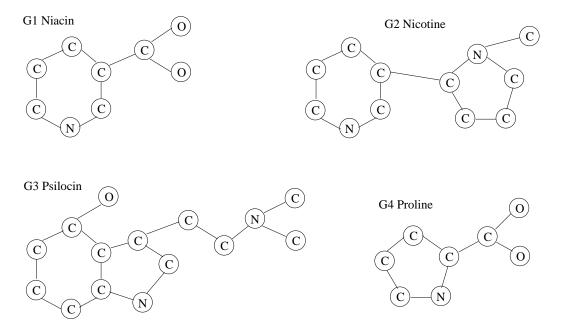


Figure 2: Four graphs corresponding to molecular structures.

# 4.4 Homework: This will be added later

This task is homework that is done in groups of 2–3 students. Note that you cannot do the task alone or in a larger group, so it is recommended to search a group now. The TAs can help to find collaborators.