

important. Consequently, Docear does not only know which papers a researcher currently is reading but also which parts of an article he considers particularly important. This allows highly personalized recommendations.

We are experimenting with different variations of content based filtering for Docear. All variations have in common that from a user's mind map collection a user model is build and the user model is matched with Docear's Digital Library containing around 1.8 million research articles from various disciplines (Figure 2). The Docear desktop software synchronizes all mind maps of a user every couple of minutes with Docear's server. When a user requests recommendations, Docear sends the request to Docear's Digital Library. This creates a user model and returns ten recommendations which are all accessible in full-text. Some users have reservations against having their mind maps transferred to Docear's server and having them analyzed. Therefore, Docear allows users to turn recommendation off which about 2/3 of users do (when users turn recommendations off, they still can use Docear to manage their literature).

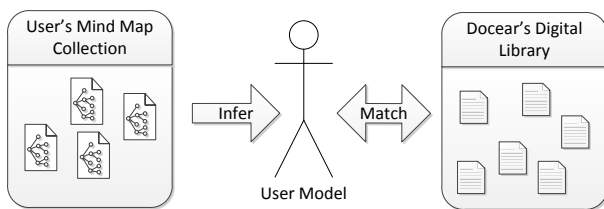


Figure 2: Basic recommendation model

Currently, we are experimenting with different variations of content based filtering. For instance, we are varying the number of mind maps and nodes that are analyzed for creating a user model (e.g. all mind maps the user created vs. only the last edited mind map), the user model size and the type of TF-IDF being used. Based on an evaluation with 938 users and 32,790 delivered recommendations, the average click through rate (CTR) is 6.03%. However, for specific variations such as when stop-words are removed, TF-IDF is applied instead of term frequency only, and only user requested recommendations are analyzed, click through rates increases to 10.31% (Figure 4), and in some specific scenarios even more. In comparison, a typical click through rate in advertisement is 0.5% [6]. Research paper recommender systems typically achieve CTRs of around 8% [7]. Further details on the recommendation algorithms will be published in a forthcoming paper.

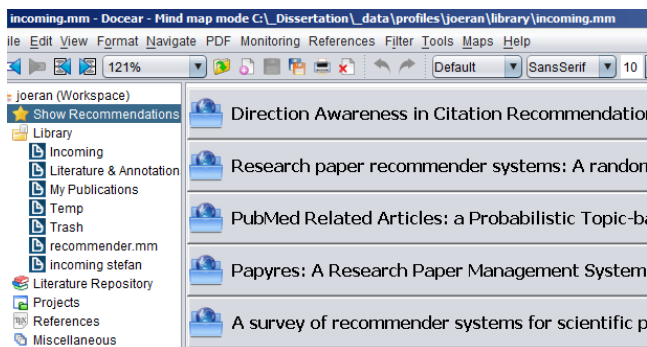


Figure 3: Recommendations in Docear (screenshot)

4. OUTLOOK

Although Docear's recommender system is in a rather early development stage, it already performs quite well. We are confident

that the performance can still be improved. For instance, as soon as user numbers of Docear rises, collaborative filtering might become an interesting addition to content based filtering (right now, there are too few similar users to perform CF). An extension of Docear's digital library should also improve the recommender's performance. With 1.8 million articles available, there are only few relevant articles to choose from for many researchers. When more papers are added to the digital library, there will be more potentially relevant recommendations. Finally, we plan on extending the recommender to not only recommend research papers but other items relevant to researchers. For instance, journals and conferences, research grants, and for students we would like to recommend university programs.

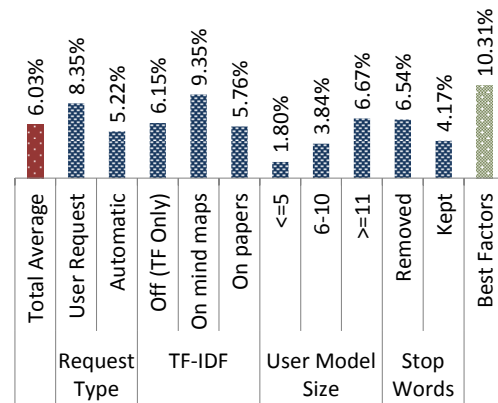


Figure 4: Results of content based filtering variations

5. REFERENCES

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