

notes and demo

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paper-rec` 📃 🤖 💙

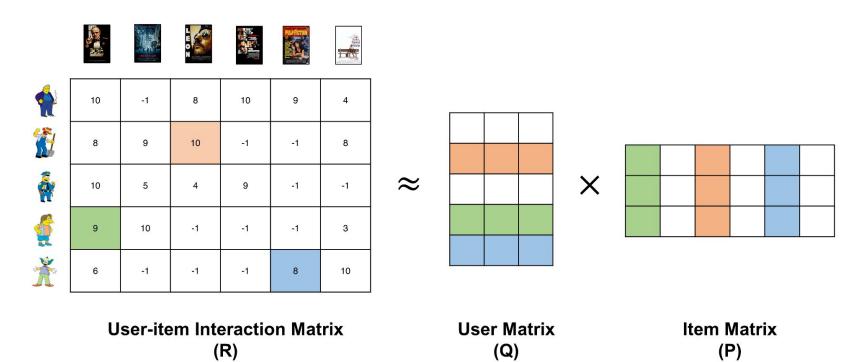




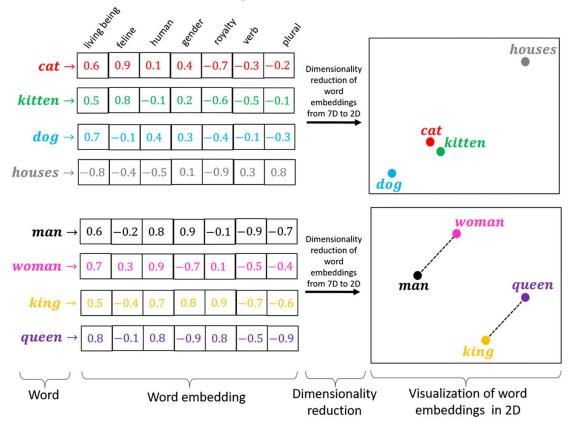


- In recommender systems, the idea is to model users and items using feature vectors (i.e., embeddings) that represent their characteristics, and then compute scores based on operations on those feature vectors (e.g., dot product), the computed score is used to create a ranked list of personalized items for a given user
- In the case of **items**, such (latent) features represent properties of the objects, e.g., for movies they may correspond to metadata such as genre, cast, director, or to content features extracted from the video and audio feed
- For users the embedding represents their preferences based on their history, e.g., what users have bought, read, or listened to before helps to capture their preferences over time

Embeddings Using Collaborative Filtering

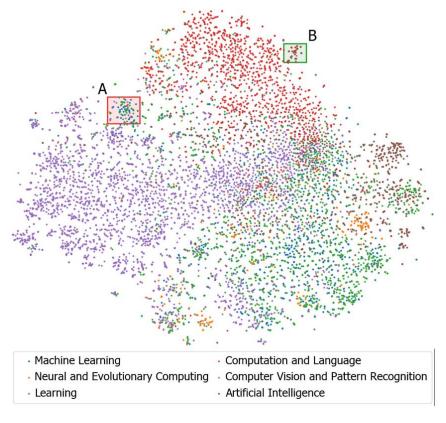


Content-based Embeddings



Content-based Embeddings

Embeddings for arXiv papers (6 ML categories)



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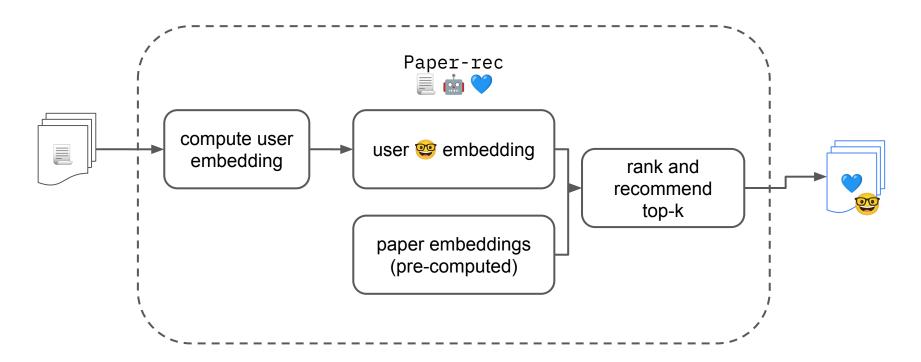


- What paper in ML/AI should I read next? It is difficult to choose from all the great research publications published daily. The goal of paper-rec is to offer a personalized selection of papers from the latest scientific contributions available
- Data: arXiv RSS feeds of latest contributions in particular:
 - <u>Artificial Intelligen</u>ce
 - Computation and Language
 - Computer Vision and Pattern Recognition
 - Information Retrieval
 - Machine Learning (cs)
 - Machine Learning (stat)
- Use <u>sentence-transformers</u> as feature extractor

`paper-rec`: representing items and users

- Represent each paper (item) using an embedding computed using a SentenceTransfomer from the sentences in the title and abstract of the paper
- For the users: represent their preferences as an embedding computed on extracts of text from articles they like, i.e., apply a SentenceTransfomer on sentences extracted from text that users like

`paper-rec`



Conclusion and Outlook

• paper-rec recommends articles relevant to user interest

Limitations:

- Paper-rec follows a content-based approach, however, combining collaborative filtering based on the observed interactions, e.g., when user clicks on a recommendation, to fine tune the embeddings could yield a better performance. In the demo is not possible to keep track if the users click or not on a recommendation and to use this information
- Integrating a recsys library as part of the Hugging Face Hub Libraries outlined here
 https://huggingface.co/docs/hub/libraries
 would be an interesting exercise and something we
 would like to explore with more time. Currently, the integration is to support downstream
 capabilities, but not the tasks associated with the libraries integration in the Hub

Conclusion and Outlook ctd.

- Interesting to explore: experiments on the embeddings extracted by the sentence transformer vs BERT using a recsys as downstream task, i.e., measure recommendation quality
- recsys ≠ NLP, CV, Audio tasks

Output

- `paper-rec` recommender library using huggingface_hub library for downstream support: https://github.com/bluebalam/paper-rec
- Dataset : https://huggingface.co/bluebalam/paper-rec
- Demo : https://huggingface.co/spaces/bluebalam/paper-rec