



HybridBERT4Rec: A Hybrid Recommender System Based on BERT

Sequential Content-Based and Collaborative Filtering

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Recap: Sequential Modelling & HybridBERT4Rec

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Traditional CBF VS Sequential CBF



Target user
(Alice)



Figure 1: Example history for Alice in traditional CBF [1]

- models **general** user preference

Traditional CBF VS Sequential CBF



Target user
(Alice)



Figure 1: Example history for Alice in traditional CBF [1]

- models **general** user preference
- **BUT:** User preferences change over time! [2]

Traditional CBF VS Sequential CBF



Target user
(Alice)



Figure 1: Example history for Alice in traditional CBF [1]

- models **general** user preference
- BUT:** User preferences change over time! [2]



Target user
(Alice)



24.1.23



26.1.23



28.01.23



30.01.23

Figure 2: Example history for Alice in sequential CBF [1]

- Considers the **order** of historical interactions
- Allows the modelling of “temporary spikes” of interests, as well as the general preferences [2]

HybridBERT4Rec Architecture

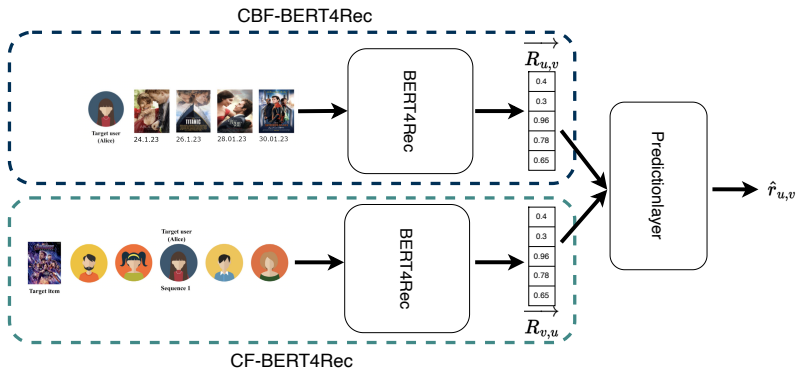


Figure 3: High level overview of HybridBERT4Recs Architecture. [1]

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The Setting

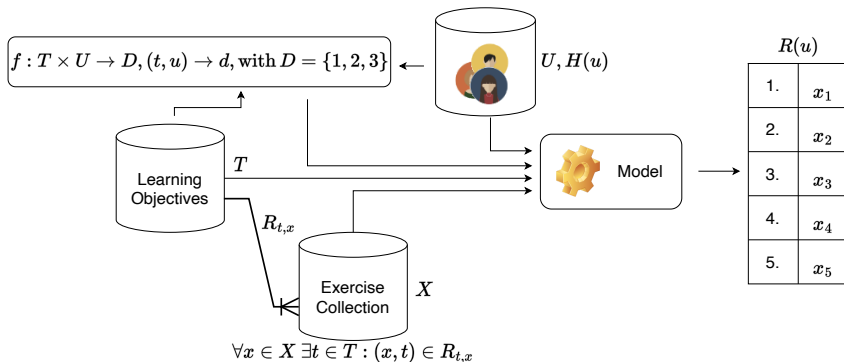


Figure 4: The Setting, consisting of a user collection U and their histories $h(u)$, a collection of learning objectives T and a collection of exercises X , which can be used to predict a ranking $R(u)$ for a given user u .

Model Adaption

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CBF-HybridBERT4Rec

- $H(u) := (\{(x_i, t_j, s_k) | (x_i, t_j) \in R_{t,x}\}, \leq)$, with $s_{k-1} \leq s_k$
- $I(u) := (\{x_i | (x_i, t_j, s_k) \in H(u)\}, \leq)$
- $\overrightarrow{R_{u,t}}$: the interaction probability distribution of all items with the user u over the target item

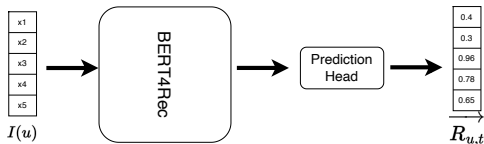


Figure 5: CBF-HybridBERT4Rec architecture and input in the described setting

CF-HybridBERT4Rec

- $u \in N \iff d_{u,t} = d_{u_m,t} \wedge (x, t) \in \{(x, t) | (x, t, s_k) \in H(u)\}$, with $U_m \in U, U \in U, t \in T, N$ being the set of neighbors for target (masked) user u_m and learning objective t
- $\vec{R}_{t,u}$: a user-similarity probability distribution of all users over the target (masked) user

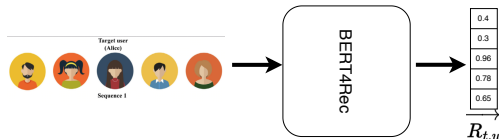


Figure 6: CF-HybridBERT4Rec architecture and input in the described setting

Evaluation

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References

- [1] Chanapa Channarong et al. “HybridBERT4Rec: A Hybrid (Content-Based Filtering and Collaborative Filtering) Recommender System Based on BERT”. In: *IEEE Access* 10 (2022), pp. 56193–56206. ISSN: 2169-3536. DOI: 10.1109/ACCESS.2022.3177610. (Visited on 11/02/2023).
- [2] Shoujin Wang et al. “Sequential Recommender Systems: Challenges, Progress and Prospects”. In: (2019), pp. 6332–6338. (Visited on 11/02/2023).