



# 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

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

# Traditional CBF VS Sequential CBF



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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

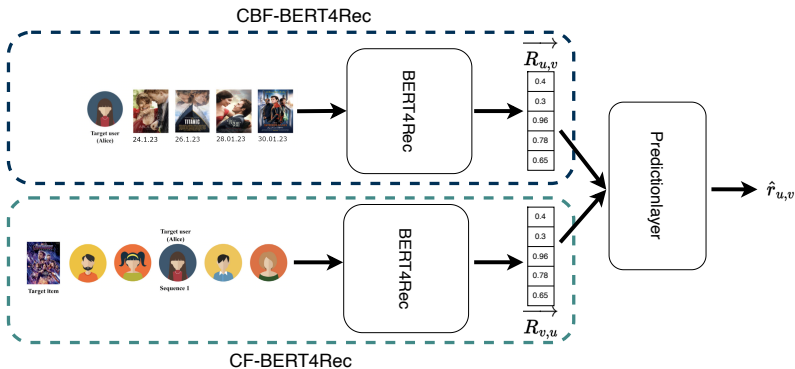


Figure 3: High level overview of HybridBERT4Recs Architecture. [1]

# The Setting

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# The Setting

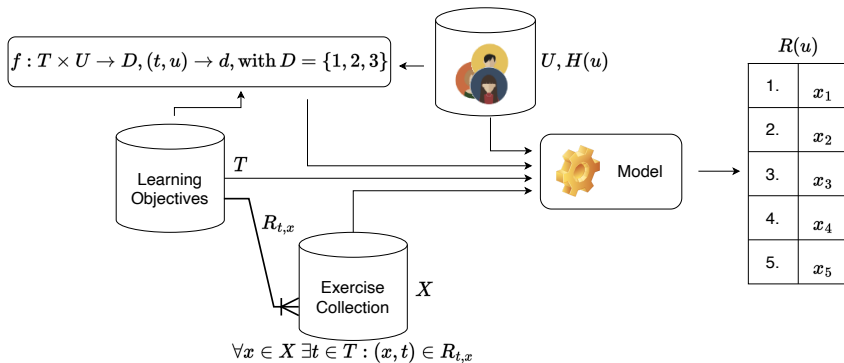


Figure 4: The Setting, consisting of a user collection  $U$  and their histories  $h(u)$ , a collection of learning objectives  $T$  and a collection of exercises  $X$ , which can be used to predict a ranking  $R(u)$  for a given user  $u$ .

# 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

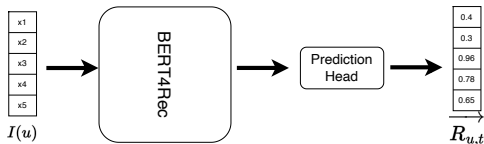


Figure 5: CBF-HybridBERT4Rec architecture and input in the described setting

# 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

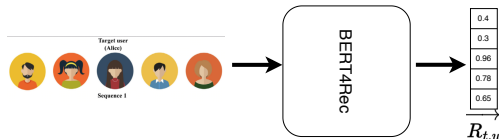


Figure 6: CF-HybridBERT4Rec architecture and input in the described setting

# Bringing It All Together

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## Algorithm 1 HybridBERT4Rec in an E-Learning Setting

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```
1: for all  $u_m \in U$  do
2:    $r_{x,u_m} = \text{cbf\_hybridbert4rec}(H(u_m))$ 
3:   for all  $(x, t) \in R_{t,x}$  do
4:      $r_{u,x} = \text{cf\_bert4rec}(u_m, t, x)$ 
5:      $\hat{r}_{u,x} = \text{prediction\_layer}(r_{x,u}, r_{u,x})$ 
6:   end for
7: end for
```

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- Yields a rating  $\hat{r}_{u,x}$  for each exercise and for each user
- Construct an overall rating of exercises by sorting the ratings
- Construct a topic specific rating by filtering for a topic and sorting the ratings

# 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).