# Deep Latent Factor Models for Recommender Systems

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#### **Recommender system**

- Recommends items to users.
- E.g. in e-commerce: Users who like this also like this.

#### Data available to a Recommender system

- Users × Items binary interaction matrix, also called the rating matrix *R*.
- User attributes G: age, location, ...
- Item attributes *H*: price, image, description, ...

### **Current recommendation models**

• Use either the interactions or the attributes but almost never both of them.

## **Proposed recommendation model**

- Combines the interactions and the attributes.
- Recommends the item *j* to the user *i* with the highest approximated rating  $R_{ij}$ :

 $R_{ij} \approx u_i \cdot v_j$ 

- $u_i = U_i = latent representations of user i$
- $v_j = V_j = \text{latent representation of item } j$
- U, V factorize the *R* => Latent Factor Model (LFM)
- The latent representations are produced by the user/item autoencoders visible in the upper/lower half of the architecture image.





# Conclusion

- A novel state of the art LFM capable of combining both interaction and attribute information has been designed and implemented.
- Four state of the art LFM based on ANN were implemented, evaluated and compared to our proposed model on several standard datasets.
- The proposed model has reached similar or better results across multiple metrics.
- The robustness of the proposed model to different hyperparameters has been evaluated during multiple experiments.
- Our model has been deployed by Recombee which offers recommendation as a service.