

CF Tutorial

Some Notes on Collaborative Filtering

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Introduction

- Want to predict user ratings for movies which they haven't rated.
- This can be achieved with [Collaborative Filtering](#).
- CF is not *one* algorithm, it's a broad set of different techniques.
- Here: Model-based approach using [Alternating Least Squares](#).

Users	Movie 1	Movie 2	Movie 3	...
User 1	?	5.9	2.6	...
User 2	1.4	5.8	?	...
User 3	1.5	5.8	?	...
...



Model-based Collaborative Filtering

- Have a *sparse* and usually large matrix $\mathbf{R} \in \mathbb{R}^{u \times v}$ of user ratings.
- u - Number of users (rows)
- v - Number of movies (columns)
- R_{ij} - Rating of user i for movie j .
- Example:

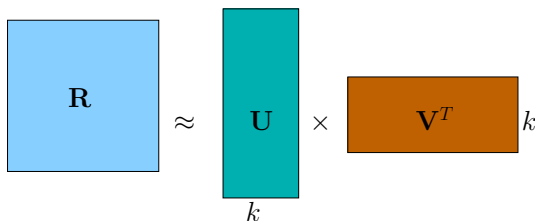
$$\mathbf{R} = \begin{pmatrix} 2.2 & 2.7 & 7.7 & 2.9 & 3.3 \\ ? & ? & 2.6 & 1.2 & 8.9 \\ 7.0 & ? & 3.5 & 0.7 & 2.1 \\ 9.1 & 0.6 & ? & 1.8 & ? \\ ? & ? & 7.4 & 3.1 & 5.9 \end{pmatrix}$$



Alternating Least Squares: Intuition

- Model-based: Try to uncover **latent factors** that model the data.
- Can be achieved through approximate matrix decomposition:

$$\mathbf{R} \approx \mathbf{U} \times \mathbf{V}^T$$



- The k columns in \mathbf{U} and \mathbf{V}^T correspond to the latent, i.e. *unobserved* factors.
- **ALS**: Approximate \mathbf{U} and \mathbf{V} through linear regression.



FIXME: Add links

