

Problem

Users: Get information interested in

Content Provider: Make contents stand out from the crowd

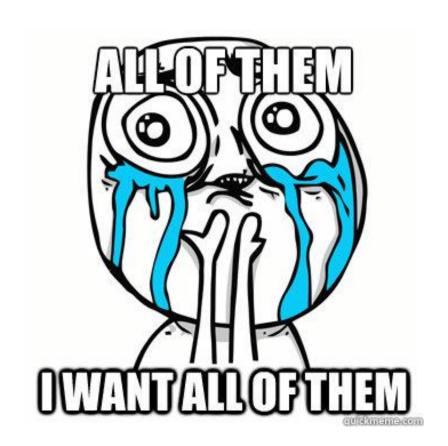
Predict "ratings" a user would give to an item or "interactions" between them

Challenges

How do we win the cold-start problem? How do we improve the recommendation quality? How do we make recommendations based on multiple factors?

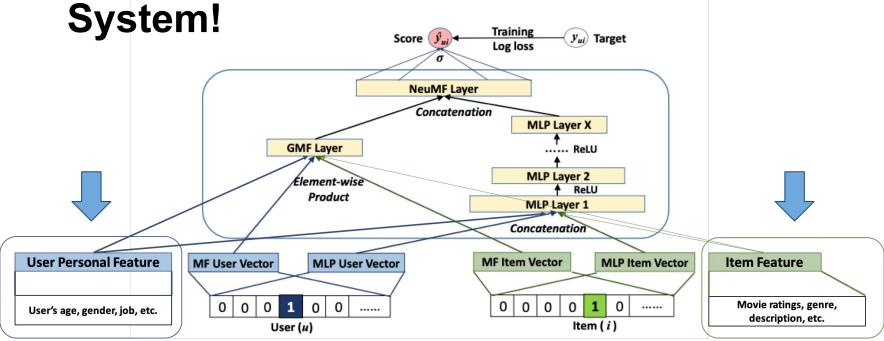
Famous Recommendation Approaches

Recommendation algorithms	Advantages	Disadvantages 1. Limited by the features extraction methods; 2. New user problem; 3. The training of classifier needs massive data; 4.Poor scalability.		
Content based	 Recommendation result is intuitive and easy to interpret; No need for users? Access history data; No new item problem and no sparsity problem; Supported by the mature technology of classification learning. 			
Collaborative filtering	 No need for professional knowledge; Performance improving as the increasing of the user number; Automatic; Easy to find user's new interesting point; Complex unstructured item can be processed. eg. Music, Video, etc. 	1. Sparsity problem; 2. Poor scalability; 3. New user and new item problem; 4. The recommendation quality limited by the history data set		



Proposed Approach

CB-CF Hybrid Neural Recommender System!



Data

MovieLens

- > 20 million ratings
- > 465,000 tag applications
- > 27,000 movie
- > 138,000 users

tagld	genome_tag	movield	relevance	title	genres	userld	tag	tags_timestamp	rating	rating_timestamp
1	7	1	0.025	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1644	Watched	1417736680	3.5	1421135243
2	007 (series)	1	0.025	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	computer animatio	1183903155	4	1038421883
3	18th century	1	0.05775	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	Disney animated f	1183933307	4	1038421883
4	1920s	1	0.09675	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	Pixar animation	1183934770	4	1038421883
5	1930s	1	0.14675	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	1741	ΤΓήΒ©a Leoni does	1245093573	4	1038421883
6	1950s	1	0.217	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	2299	Pixar	1403306852	3	1403306845
7	1960s	1	0.067	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	3596	animation	1290312028	1	1290312052
8	1970s	1	0.26275	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	3596	family	1290312032	1	1290312052
9	1980s	1	0.262	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	3596	Tom Hanks	1290312036	1	1290312052
10	19th century	1	0.032	Toy Story (1995)	Adventure Animation Children Comedy Fantasy	6233	Pixar	1387079572	4.5	1390443274

Evaluation Methodology

Hit Ratio (HR) on top-K list

$$HR@K = \frac{Number of Hits@K}{|GroundTruth|}$$

Normalized Discounted Cumulative Gain (NDCG) on top-K list

Normalizer

graded relevance at position i
$$NDCG@k = Z_K \sum_{i=1}^K \frac{2^{r_i} - 1}{log_2(i+1)}$$