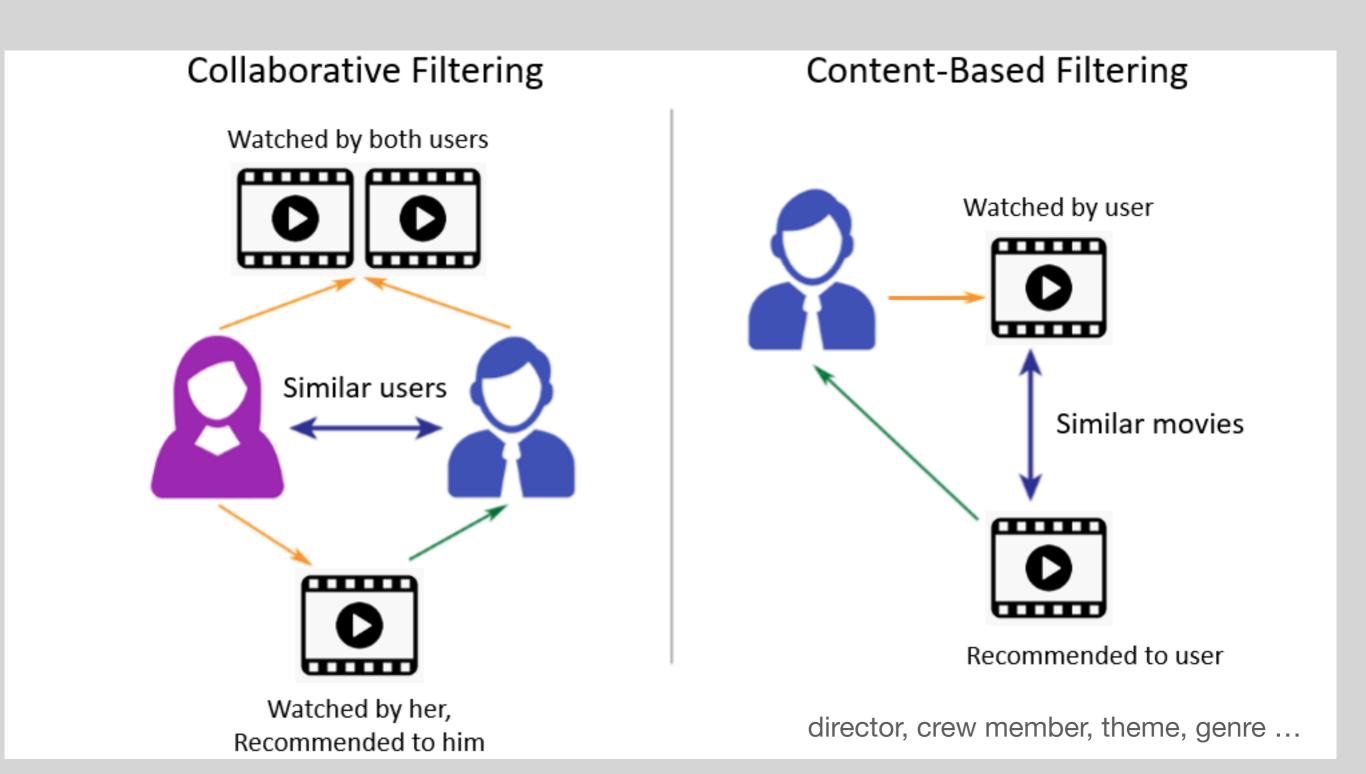
A Movie Recommender System

based on collaborative-based method





Data

- 62,000 Movies
- 163,000 users
- 25 million 5-star rating records (0.5 lowest, 5.0 highest with step 0.5) between 1995 and 2019

Constraints

- Also recorded in IMDB dataset (drop 3.6%)
- Movies have at least 10 reviewers

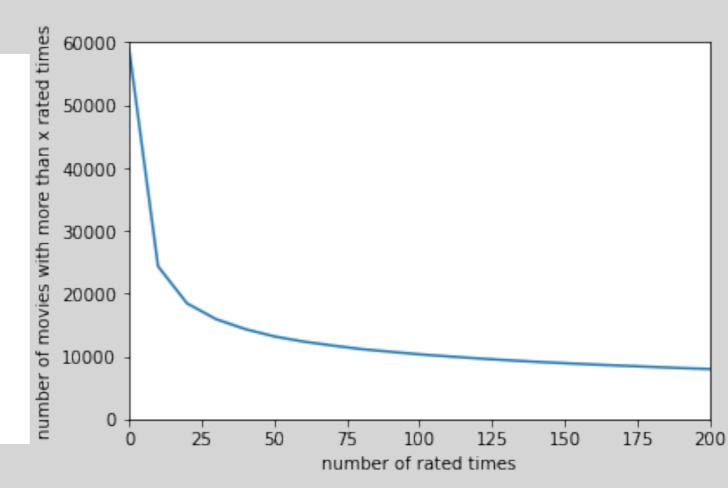
24,430 movies left in the dataset

Data Manipulation and Analysis

genres	title	ovield	ı
Adventure Animation Children Comedy Fantasy	Toy Story (1995)	1	0
Adventure Children Fantasy	Jumanji (1995)	2	1
Comedy Romance	Grumpier Old Men (1995)	3	2
Comedy Drama Romance	Waiting to Exhale (1995)	4	3
Comedy	Father of the Bride Part II (1995)	5	4

Movies

	userld	movield	rating	timestamp
0	1	296	5.0	1147880044
1	1	306	3.5	1147868817
2	1	307	5.0	1147868828
3	1	665	5.0	1147878820
4	1	899	3.5	1147868510

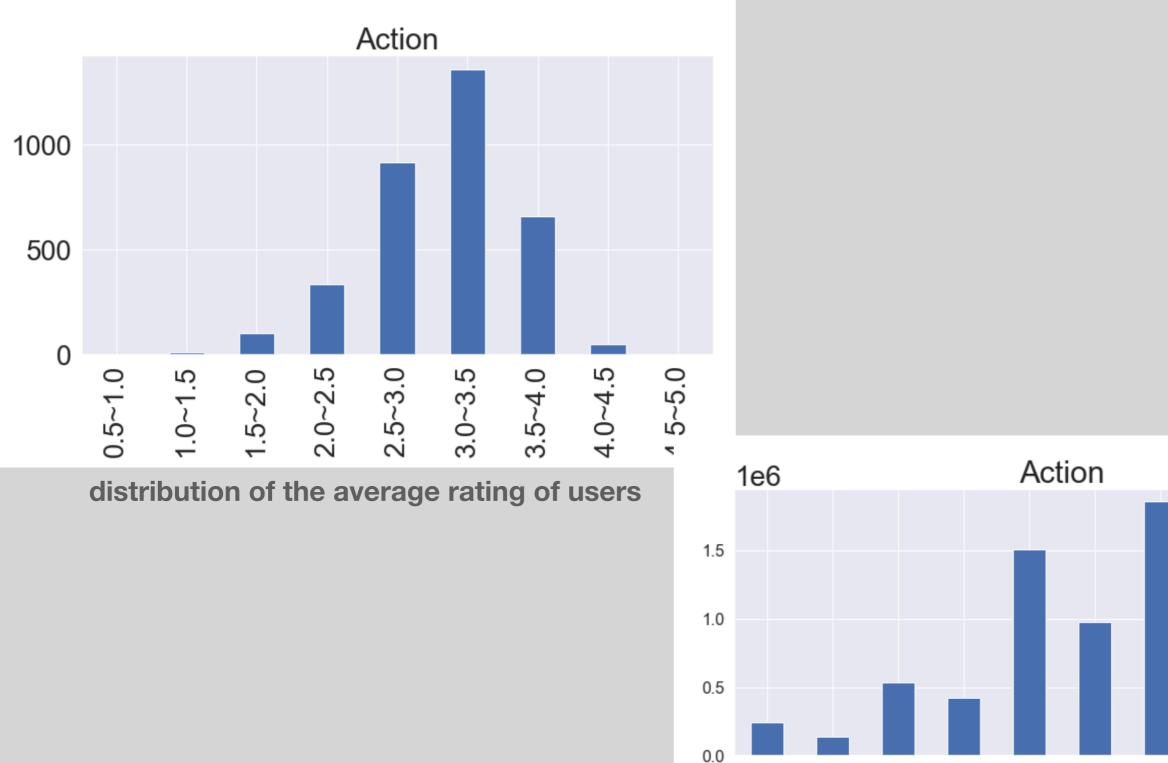


expand genres in boolean columns, add rating numbers, average rates for movies

movield	title	rating numbers	average rate		Adventure	Animation	Children	Comedy	Crime	Documentary	
1	Toy Story (1995)	57309	3.893708	False	True	True	True	True	False	False	
2	Jumanji (1995)	24228	3.251527	False	True	False	True	False	False	False	

Add statistics for each user

	userld	Movie Watched	Highest Rate	Lowest Rate	Average Rate
0	1	68	5.0	0.5	3.860294
1	2	184	5.0	0.5	3.630435
2	3	656	5.0	2.0	3.697409
3	4	239	5.0	0.5	3.380753
4	5	101	5.0	2.0	3.752475



distribution of the average rating of movies

number of rated 3.0

number of rated 3.5

number of rated 4.0

number of rated 4.5

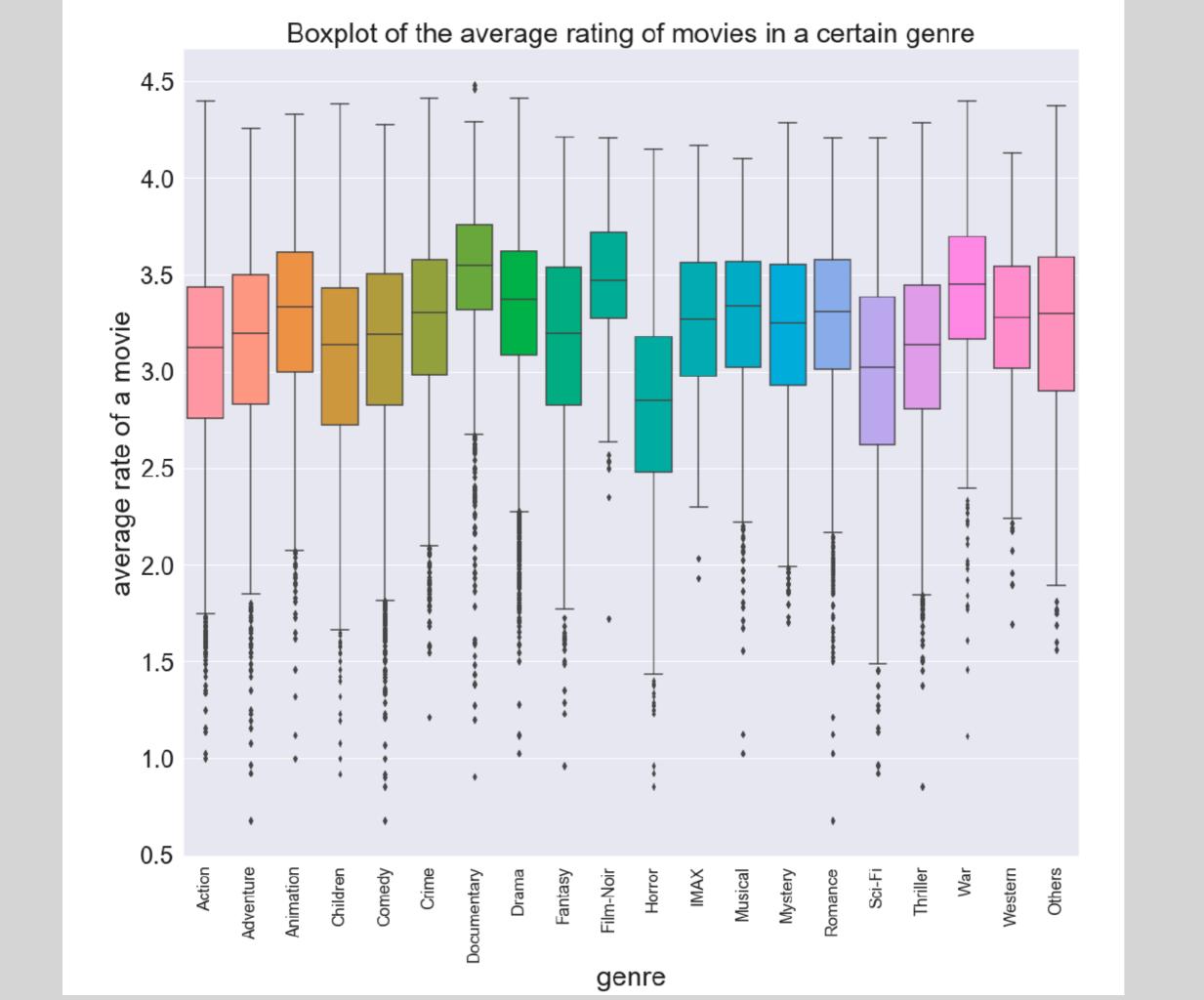
number of rated 5.0

number of rated 1.0

number of rated 2.0

number of rated 1.5

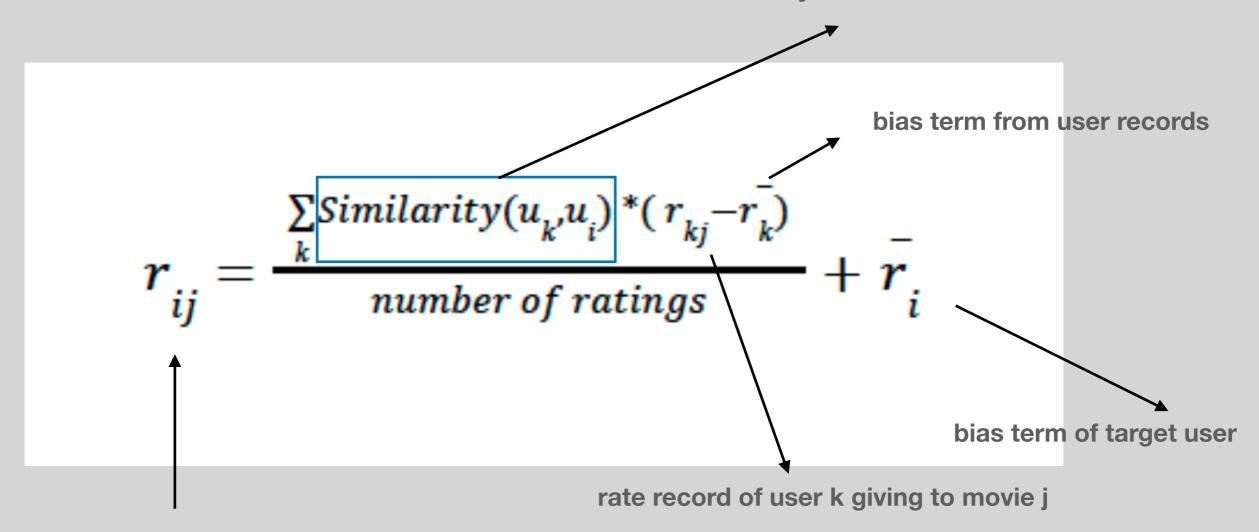
number of rated 2.5



Model

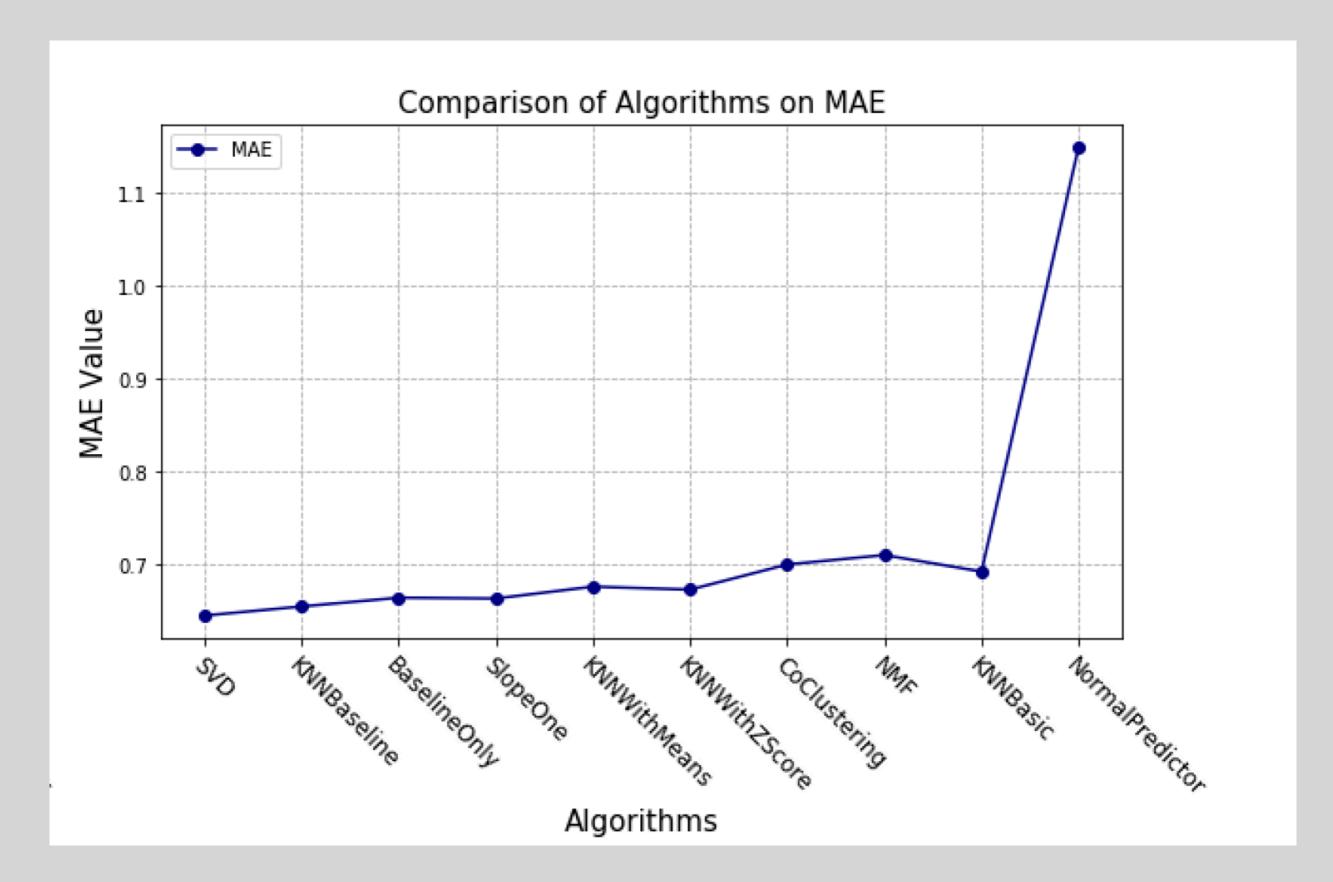
collaborative-based method

(Euclidean distance, Pearson's coefficient and cosine similarity) similarity function between user k and i



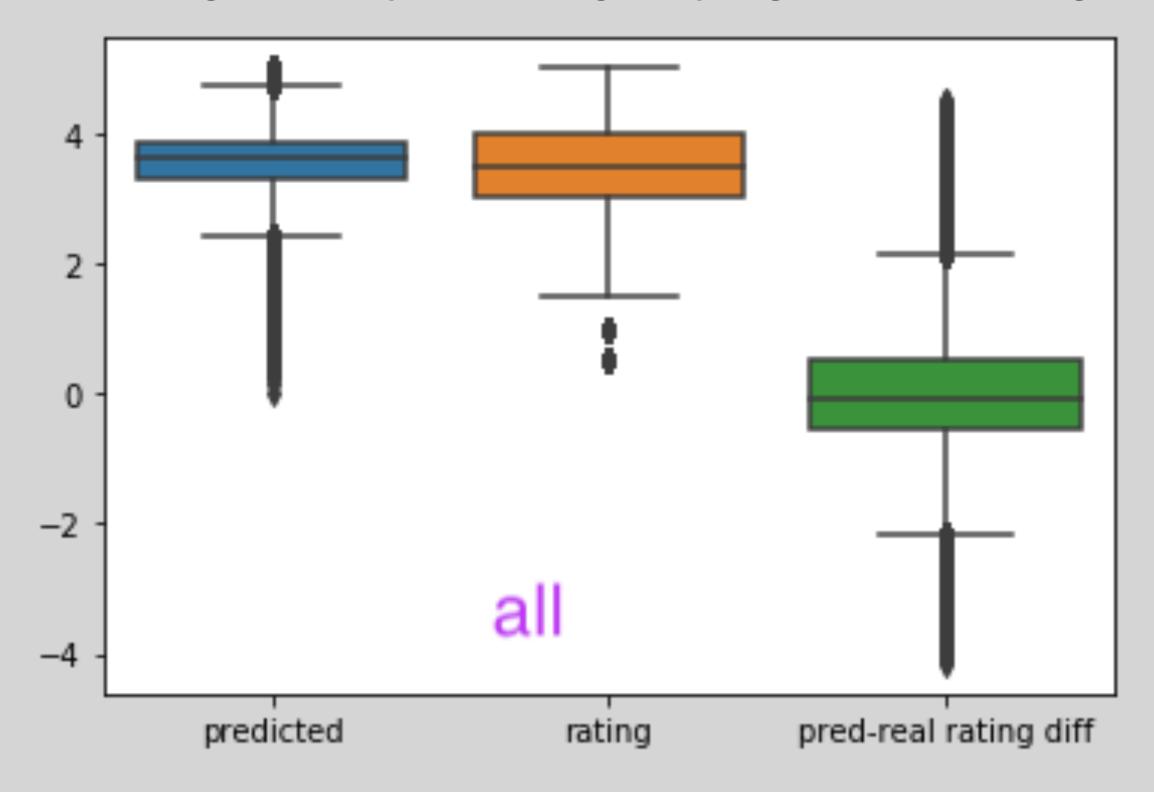
target: predicted rate of user i to movie j

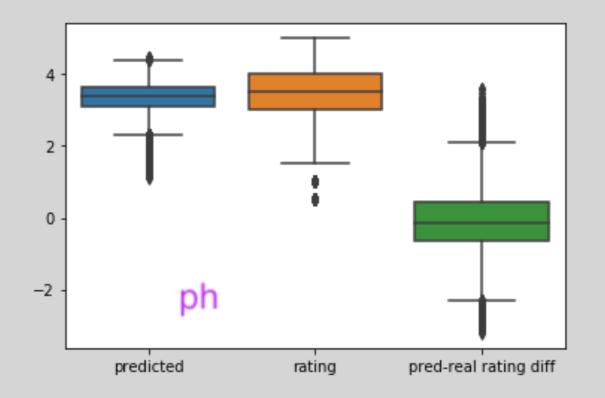
the problem of scarcity and sparsity: matrix factorization

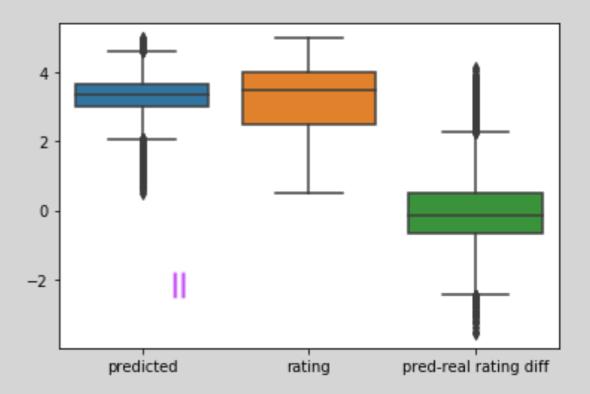


Performance Analysis

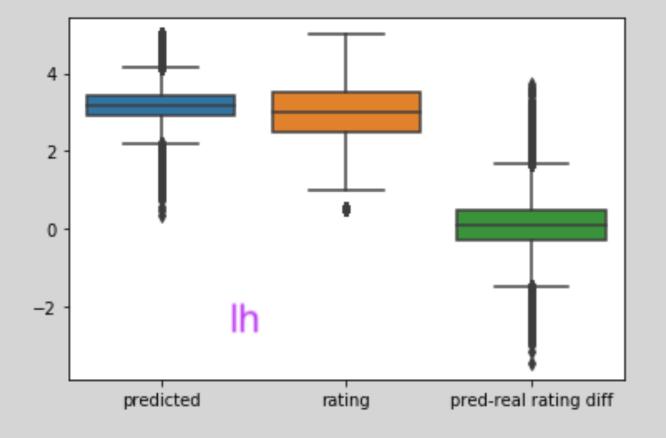
overall box diagrams of the predicted ratings comparing with the actual ratings

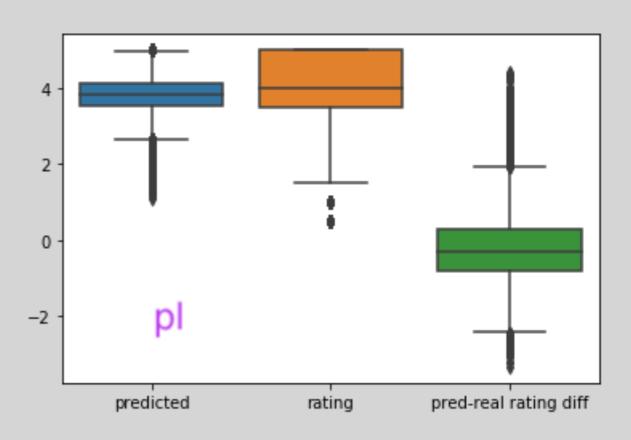




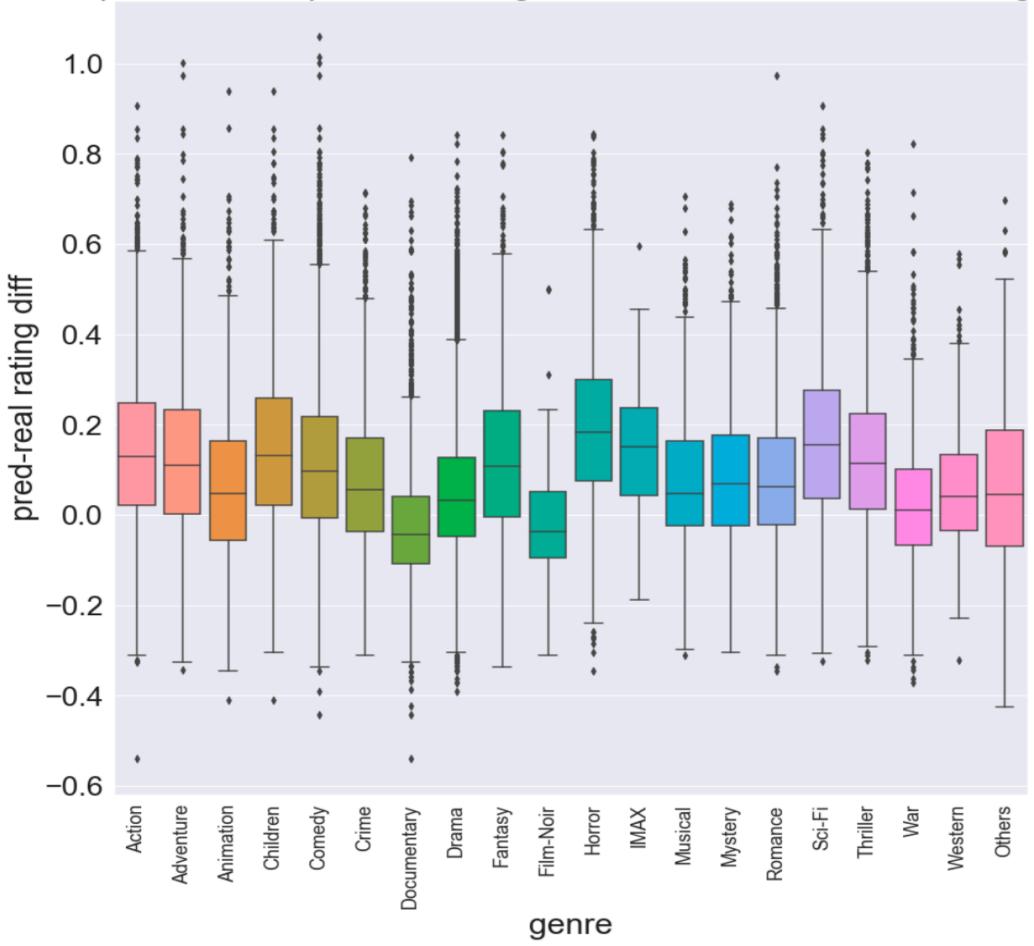


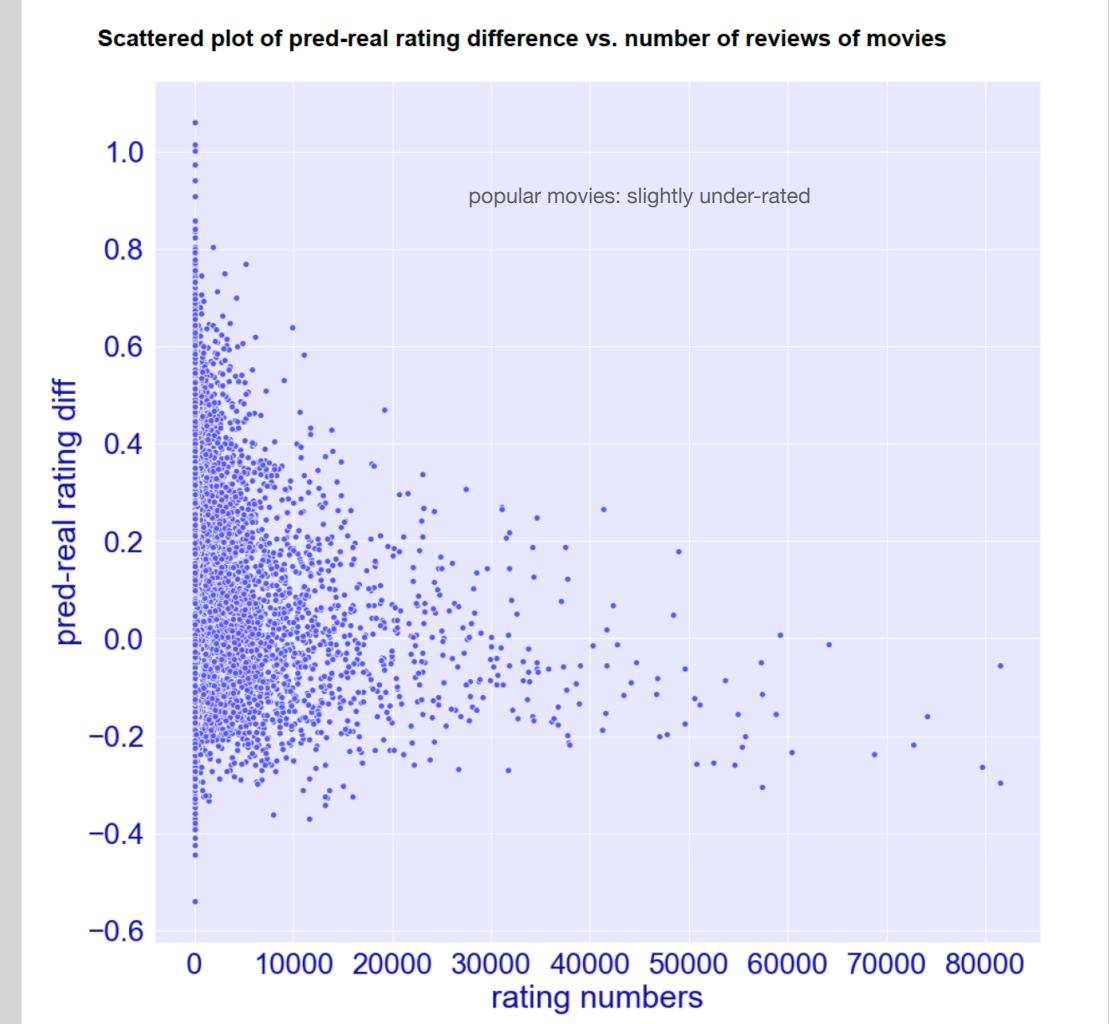
different cases: p(l)h(l) means popular (less-known) movies and heavy (light) users



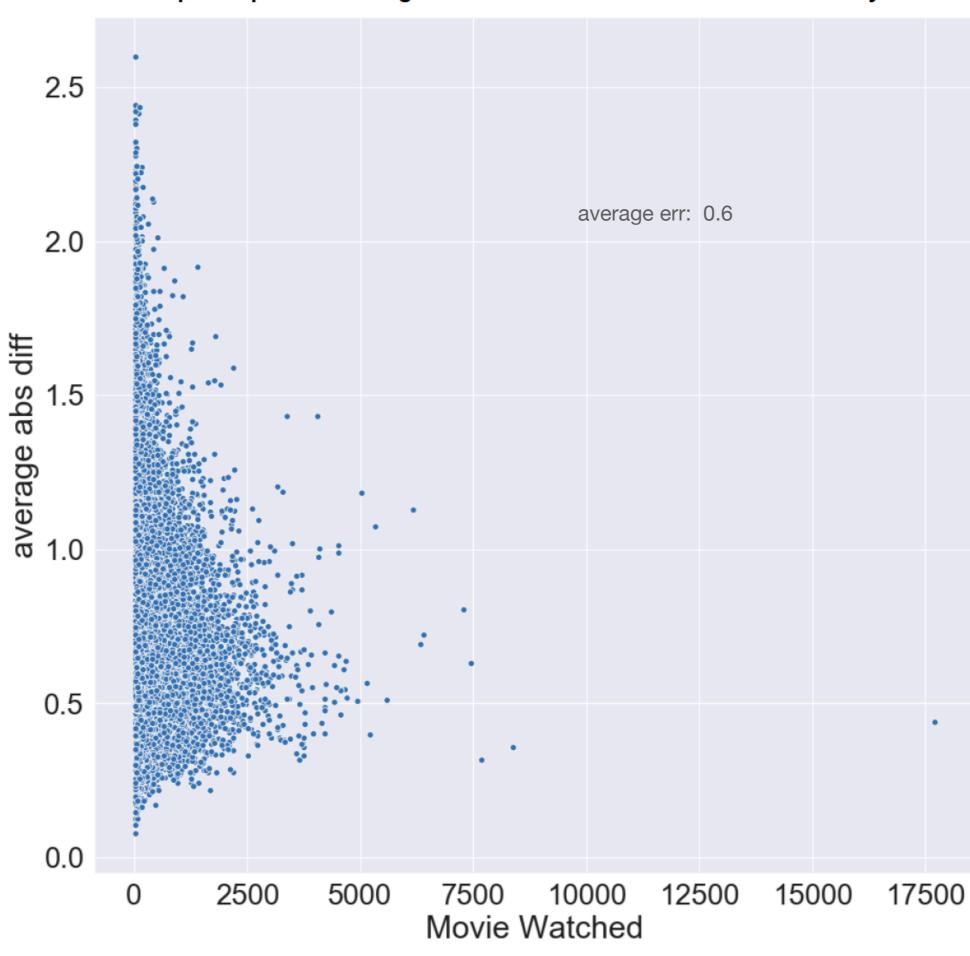


Boxplot of the the pred-real rating difference of movies in a certain genre





Scattered plot of pred-real rating difference vs. number of movies watched by users



Examples of concrete predictions

	title	rating numbers	average rate	average pred rate	pred-real rating diff	abs
108	Braveheart (1995)	59184	4.002273	4.008778	0.006505	0.006505
475	Jurassic Park (1993)	64144	3.679175	3.666863	-0.012312	0.012312

Two Popular Movies that have Low Prediction Error

	title	rating numbers	average rate	average pred rate	pred-real rating diff
22668	What Men Do! (2013)	11	1.545455	2.605522	1.060068
19804	The Coed and the Zombie Stoner (2014)	12	1.416667	2.430434	1.013767

Two Most Overrated Movies

	title	rating numbers	average rate	average pred rate	pred-real rating diff
20670	BaadAsssss Cinema (2002)	10	4.15	3.610038	-0.539962
22274	George Carlin: What Am I Doing in New Jersey?	10	3.75	3.307036	-0.442964

Two Most Underrated Movies

Summary

- A collaborative-based method movie recommendation system is built by SVD algorithm, with MAE 0.68
- less-known movies tend to be over-predicted, while popular movies tend to be slightly under-predicted
- for almost every genre of movies, 75% of the median prediction error is within the range of -0.1 to 0.3, within the smallest rating step.
- for a user who has a large number of review records, typically our recommendation system has the average error around 0.6.

