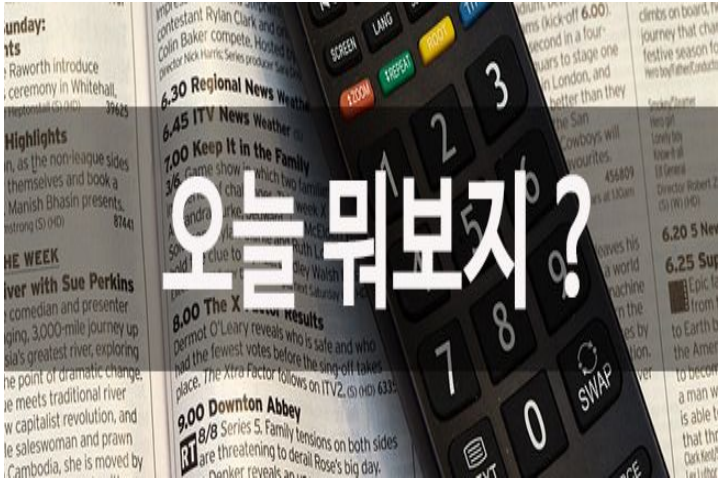
VOD RECOMMENDATION SYSTEM

설재완

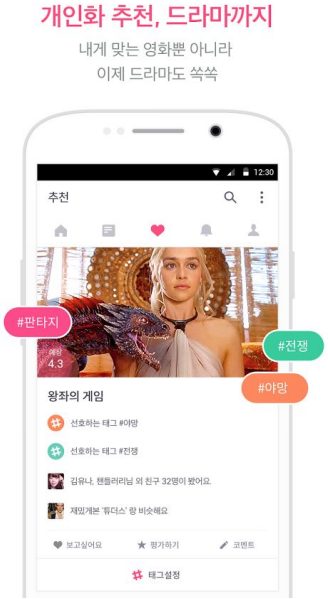
Goal



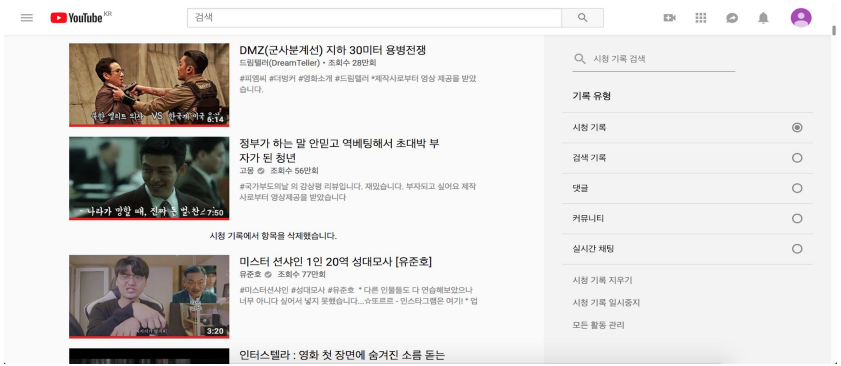
How about others

NETFLIX

WATCHA PLAY

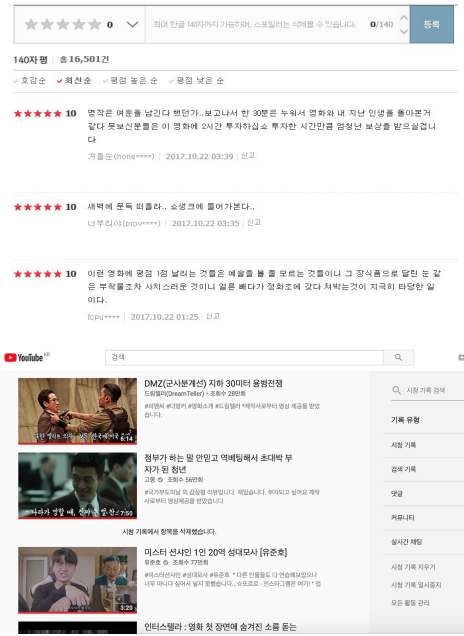


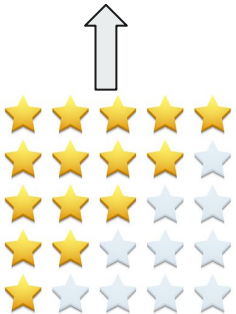
No rating but history

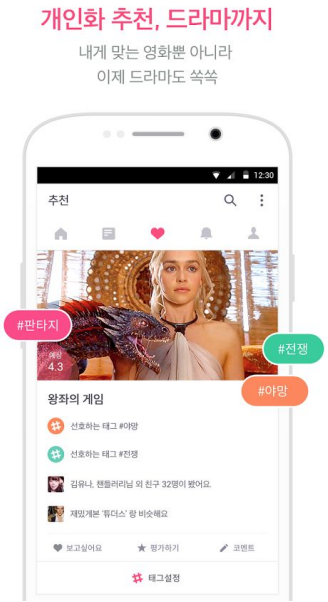


	USER_ID	SERIES_ID	ASSET_ID	DURATION	EVENT_TIME
23451632	303428	1122	22022	3128	2017-09-30 23:58:57
23451633	5059	90	5233	52	2017-09-30 23:58:57
23451634	1444	22	588	722	2017-09-30 23:58:59
23451635	459223	592	24531	613	2017-09-30 23:58:59
23451636	459214	150	3677	1761	2017-09-30 23:59:00

Conversion From History To Rating



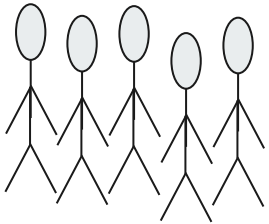




How to convert

29:37 / 59:57

45:11 / 59:57



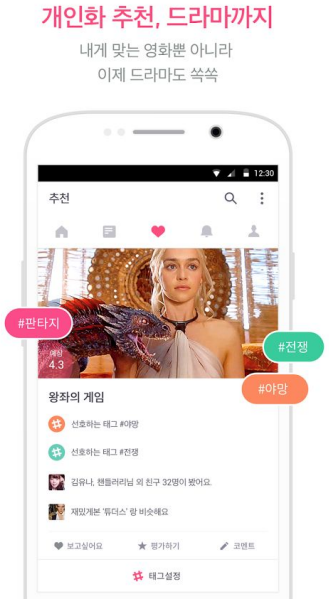
mean watching time = 30

watching time = 60

Recommendation Overview

Watch History ⇨ **Rating** ⇨ **Rating Based Recommendation** ⇨

ex)
collaborative filtering
matrix factorization



history based recommendation

Implement - Rating Conversion

사람	영상	시청시간
갑	벡터맨	10
을	벡터맨	40
갑	벡터맨	10
갑	스칼라맨	30
병	스칼라맨	10

absolute time

사람	영상	시청시간
갑	벡터맨	20
갑	스칼라맨	30
을	벡터맨	40
병	스칼라맨	10

relative time

	벡터맨	스칼라맨
갑	20 / 40 = 0.5	30 / 30 = 1
을	40 / 40 = 1	
병		10 / 30 = 0.33

user - asset matrix

Rating based recommendation ex)1

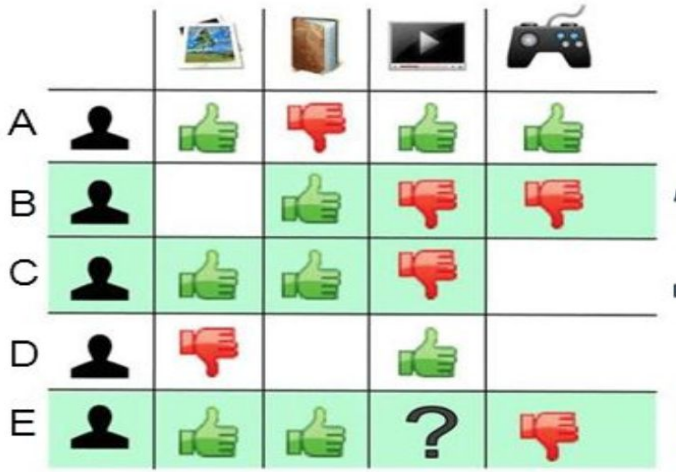
		Items					
		1	*2*	...	*i*	...	*m*
Users	*1*	5	3		1	2	
	2		2				4
	:			5			
	u	3	4		2	1	
	:					4	
	n			3	2		

baseline model

- **consider each user and asset**
- **rui = mean + bu + bi**
- **also can be applied to another model**

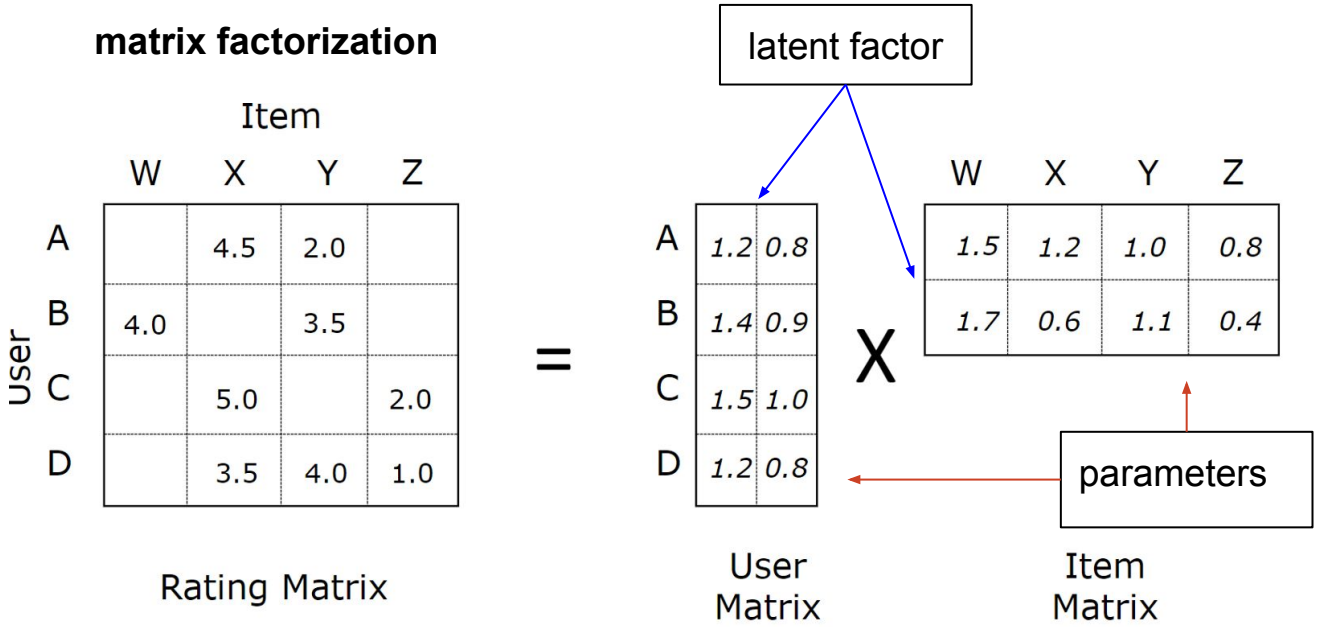
Rating based recommendation ex)2

collaborative filtering



- **Similarity**
  - **between item**
  - **between user**

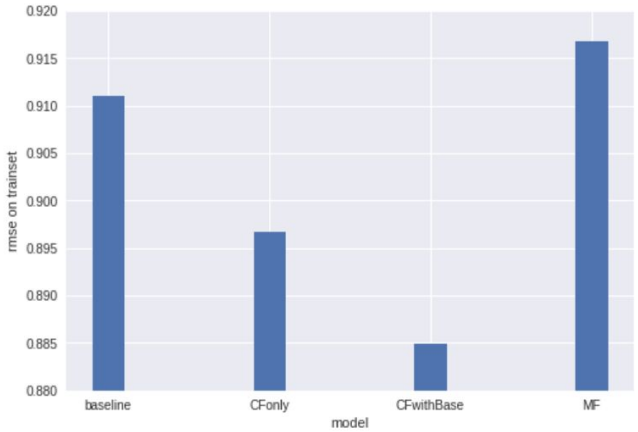
Rating based recommendation ex)3



Experiment

- 21,000,000 records
- 555,000 users
- 55,000 assets
- Compare and select model between
  - baseline
  - collaborative filtering only(CFonly)
  - collaborative filtering with baseline(CFwithBase)
  - matrix factorization(MF)
- Tain set: watch history of July, August
- Test set: watch history of September

Results



Final accuracy using CFwithBase: 47.24%

Limitation and future work

- If there were validation set, results will be better
- If there were good optimization technique(optimizer in ML/DL framework, fine tunning, etc), results will be better(It might not be trained well)
- If we can consider watching pattern(interval between each asset, consecutive assets, etc), results will be better
- If we use ‘genre’ and ‘series’ data well, results will be better
- If we pre-processed raw data well(there were so many histories of ‘뽀로로’), results will be better

Conclusion

- sutdy and implement basic ML concepts
  - use only numpy
  - gradient descnt, early stopping
- study and implement recommendation algorithm
  - collaborative filtering
  - matrix factorization
- read paper and implement
- problem solving life cycle
  - recognition -> background study -> design and implement -> evaluation -> result analysis

Thanks!

BACKUP SLIDES

Rating Conversion

- Let arr[i, j] be rating of user i at asset j (after converting history to rating)
- for jdx in range(len(asset)):
- ratings = arr[:, j] # all ratings for asset j
- median = sorted(ratings)[len(ratings) / 2] # median
- arr[:, jdx] /= median # divide by median

Rating Conversion

- Relative time
  - t = (total watching time) / (running time)
  - can over-estimate for cases that short running time
- Absolute time
  - t = (total wathcing time)
  - can over-estimate for cases that long running time
- So, let's use (total watching time) / (mean or median watching time)
  - can consider other people(relative time) and pure wathcing time(absolute time) at once
  - median is better when extreme value exists