

FYP JC2104

Presentation

Term 2

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1. Introduction
2. Recommendation System
3. Implementation
4. Experiments and Results
5. Conclusion

Introduction

Recommendation System



NETFLIX

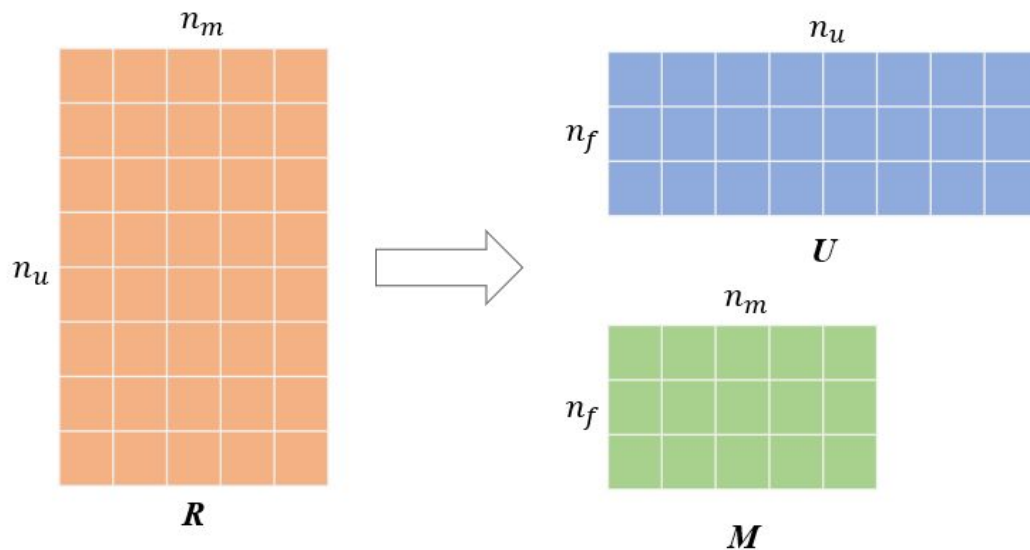
Recommendation System

- Collaborative filtering

		<i>item id</i>				
		1	2	3	4	5
<i>user id</i>	1	3	?	1	1	?
	2	?	?	?	?	4
	3	1	?	5	?	?
	4	?	?	?	?	?
	5	?	2	?	?	5
	6	5	?	?	?	?
	7	?	?	?	?	4
	8	?	?	?	?	?

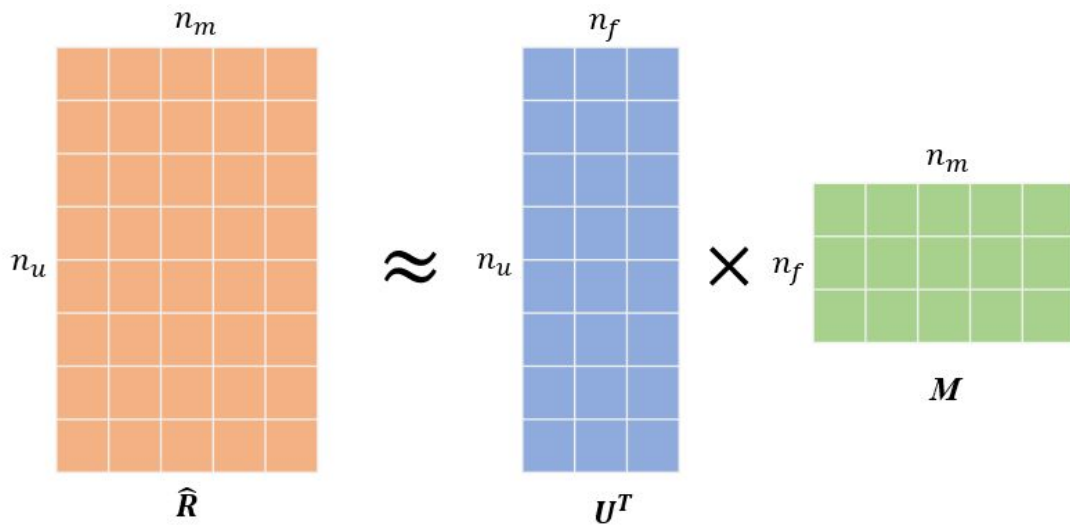
Recommendation System

- Latent factor model



Recommendation System

- Latent factor model



$$\hat{r}_{um} = p_u^T \cdot q_m$$

Recommendation System

- Loss function for Optimizing feature vectors U and M

$$f(U, M) = \min_{U, M} \sum_{r_{um} \in R} (r_{um} - p_u^T q_m)^2$$

$$f(U, M) = \min_{U, M} \sum_{r_{ui} \in R} (r_{ui} - p_u q_i)^2 + \lambda \left(\sum_u n_{user_u} \|p_u\|^2 + \sum_i n_{movie_m} \|q_i\|^2 \right)$$

- Alternating Least Square

$$p_u = \left(\sum_{r_{um} \in r_{u*}} q_m q_m^T + \lambda n_{user_u} I_{nf} \right)^{-1} \sum_{r_{um} \in r_{u*}} q_m r_{um}^T = (M_u M_u^T + \lambda n_{user_u} I_{nf})^{-1} (M_u R_{u_m})$$

$$q_m = \left(\sum_{r_{um} \in r_{*m}} p_u p_u^T + \lambda n_{movie_m} I_{nf} \right)^{-1} \sum_{r_{um} \in r_{*m}} p_u r_{um} = (U_m U_m^T + \lambda n_{movie_m} I_{nf})^{-1} (U_m R_{m_u})$$

Recommendation System

- Netflix Prize Dataset

100,480,507 ratings (1 - 5); 1.2%

480,189 users

17,770 movies

1:

1488844,3,2005-09-06

822109,5,2005-05-13

885013,4,2005-10-19

30878,4,2005-12-26

823519,3,2004-05-03

893988,3,2005-11-17

124105,4,2004-08-05

1248029,3,2004-04-22

1842128,4,2004-05-09

2238063,3,2005-05-11

1503895,4,2005-05-19

Implementation

- Data Reading
- Data Transformation
- Optimization
- Prediction and Recommendation

Implementation

- Data Reading
 - `n_job_processes`, `parallelism`, `nf`, `lambda`, `iterations`, `seed`, and data file path

`run_als.py`

`ALS.CC`

Submit job to Ursa Cluster

Read data files

Create `R_users`

Create `user_id_map`

Switch `user_id` to index

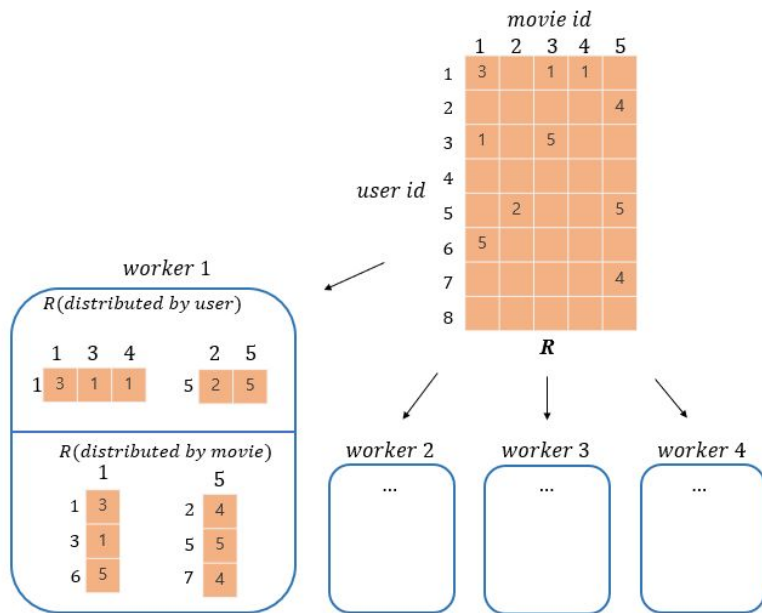
Create `R_movies`

Create `movie_feature_matrix`

Create `user_feature_matrix`

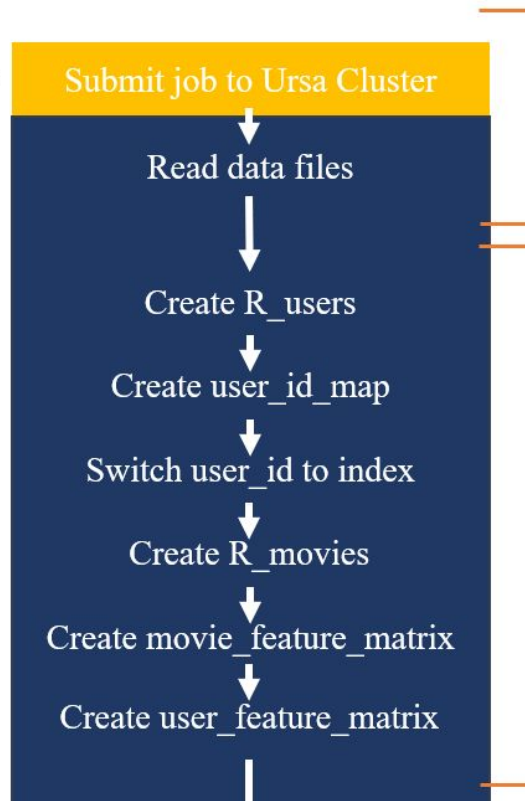
Implementation

- Data Transformation



run_als.py

ALS.CC



Implementation

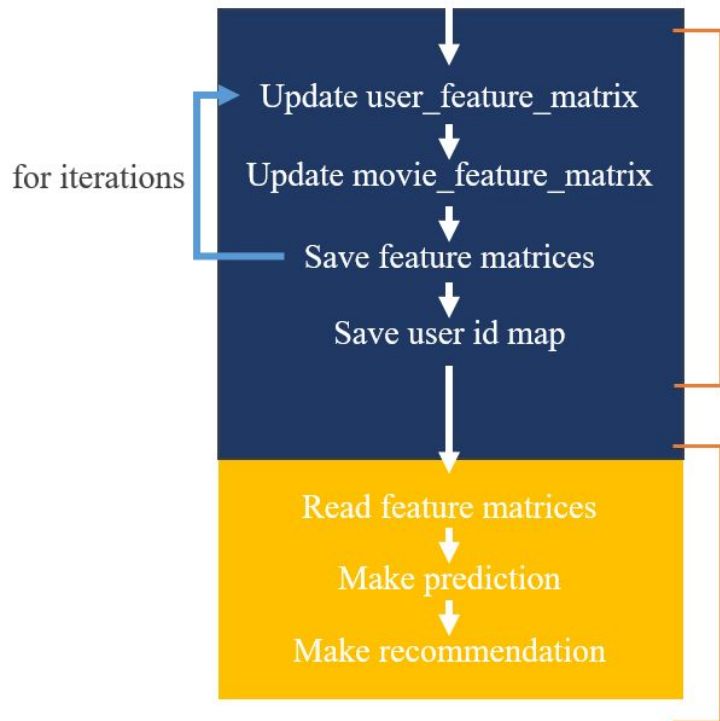
- Optimization

$$p_u = (M_u M_u^T + \lambda n_{user_u} I_{nf})^{-1} (M_u R_{u_m})$$

$$q_m = (U_m U_m^T + \lambda n_{movies_m} I_{nf})^{-1} (U_m R_{m_u})$$

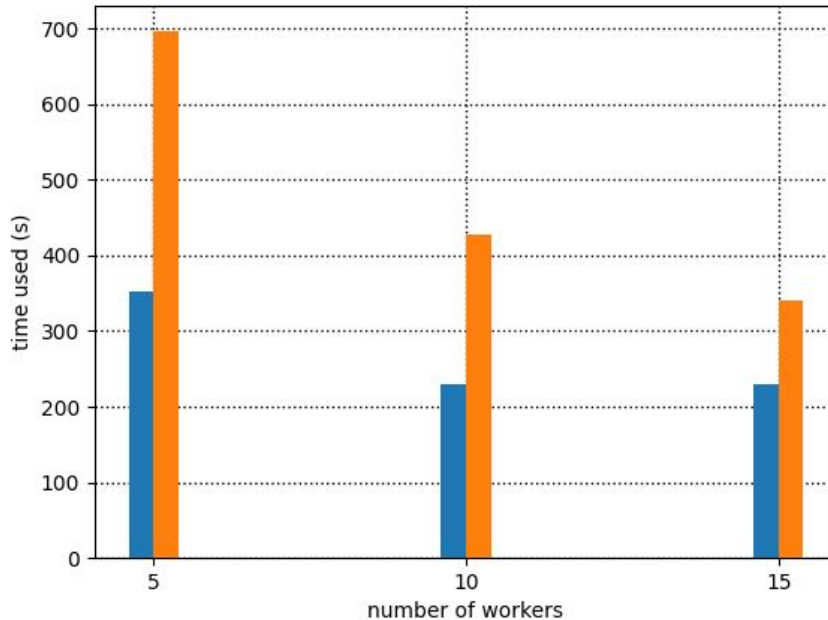
- Prediction and Recommendation

$$\hat{r}_{um} = p_u^T \cdot q_m$$



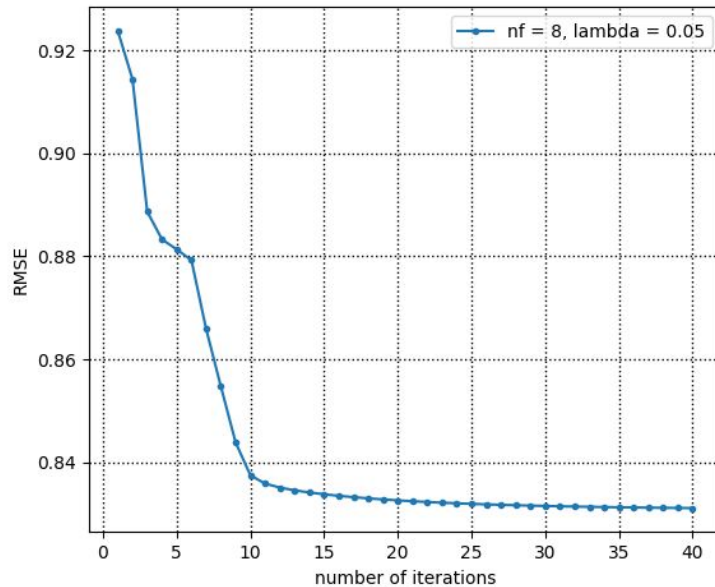
Experiments and Results

- Number of workers($\lambda=0.05$, iterations=20)



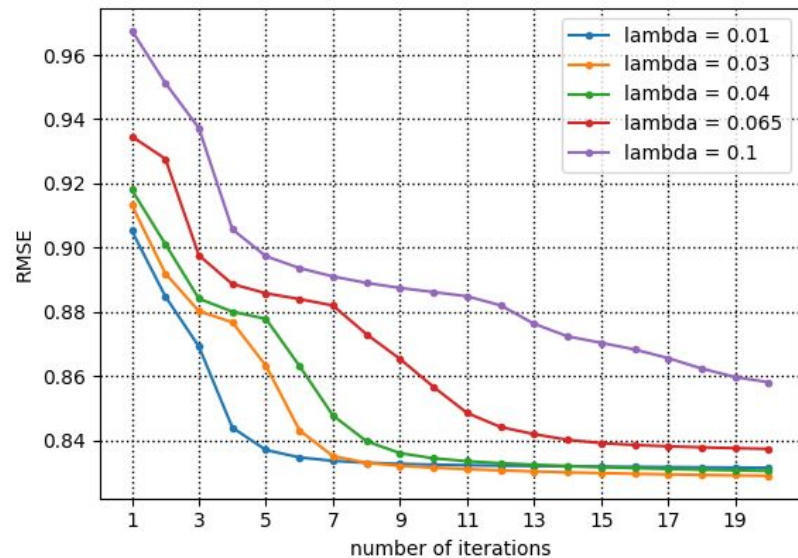
Experiments and Results

- Number of iterations



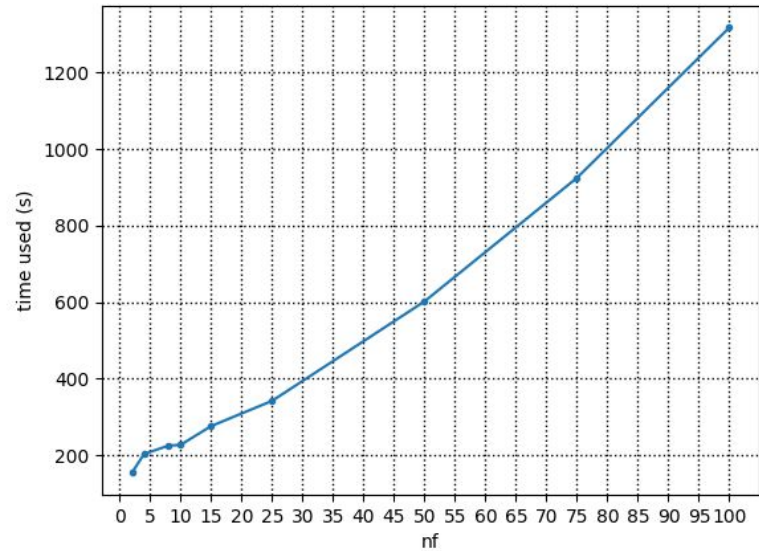
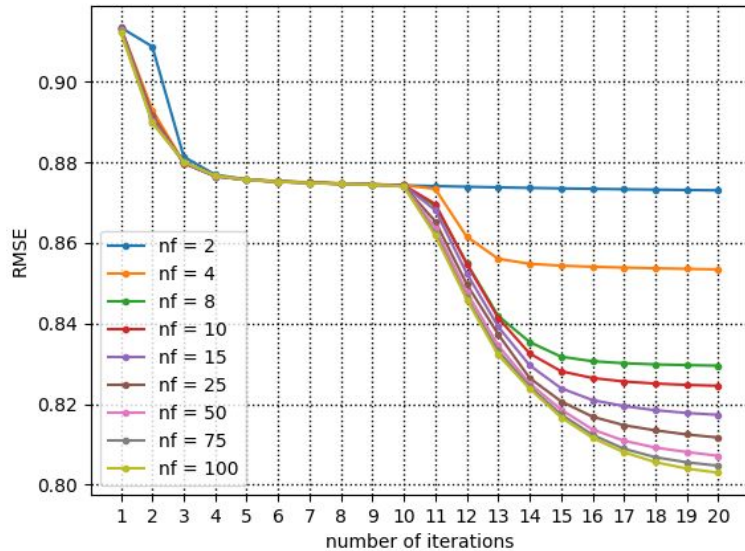
Experiments and Results

- Lambda (nf=8, iterations=20)



Experiments and Results

- $\text{nf} (\text{lambda}=0.03, \text{iterations}=20)$



Experiments and Results

- $\text{nf} = 200, \text{lambda} = 0.03, \text{iterations} = 25$
 - RMSE: 0.794831436337
 - time used: 5426531ms (about 1.5 hour)

Experiments and Results

- User ID: 1583391

- 10 5-Rated Watched of Movies

- Independence Day([Action](#), [Adventure](#), [Science Fiction](#))
- Jurassic Park([Adventure](#), [Science Fiction](#))
- Spider-Man([Fantasy](#), [Action](#))
- The Bourne Identity([Thriller](#), [Action](#), [Drama](#))
- The Italian Job([Action](#), [Crime](#))
- The Matrix([Action](#), [Science Fiction](#))
- The Patriot([Action](#), [Thriller](#))
- Stargate SG-1: Season 5, 6, 7([Sci-Fi & Fantasy](#), [Action & Adventure](#), [Mystery](#))
- Saving Private Ryan([Drama](#), [History](#), [War](#))
- Band of Brothers([Drama](#), [War & Politics](#))

- Top 7 Recommended Movies

1. Braveheart([Action](#), [Drama](#), [History](#), [War](#))
2. Lord of the Rings: The Return of the King: Extended Edition([Adventure](#), [Fantasy](#), [Action](#))
3. The Lord of the Rings: The Fellowship of the Ring: Extended Edition([Adventure](#), [Fantasy](#), [Action](#))
4. Lord of the Rings: The Two Towers: Extended Edition([Adventure](#), [Fantasy](#), [Action](#))
5. Gladiator: Extended Edition([Action](#), [Drama](#), [Adventure](#))
6. Lord of the Rings: The Fellowship of the Ring([Adventure](#), [Fantasy](#), [Action](#))
7. Stargate SG-1: Season 4([Sci-Fi & Fantasy](#), [Action & Adventure](#), [Mystery](#))

Conclusion

Thank you!