



BUILDING RECOMMENDATION ENGINES WITH PYSPARK

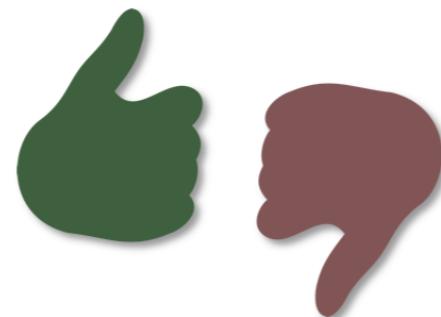
**Why learn how to build
recommendation engines?**

Jamen Long
Data Scientist

What recommendations look like



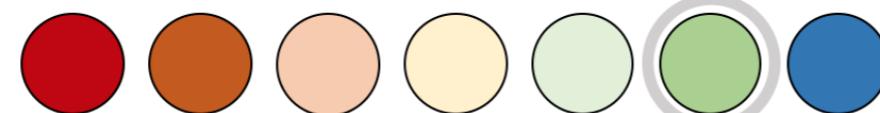
Learning about you



You rated this item:



Worst



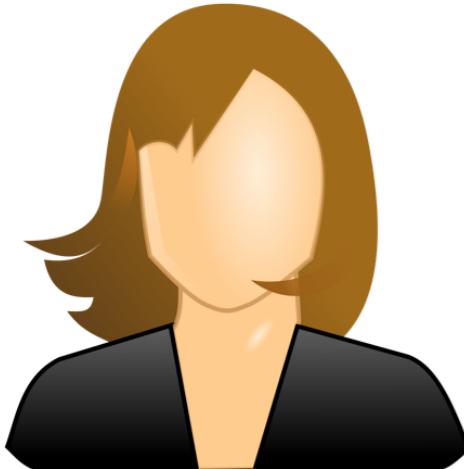
Best

How recommendation engines work

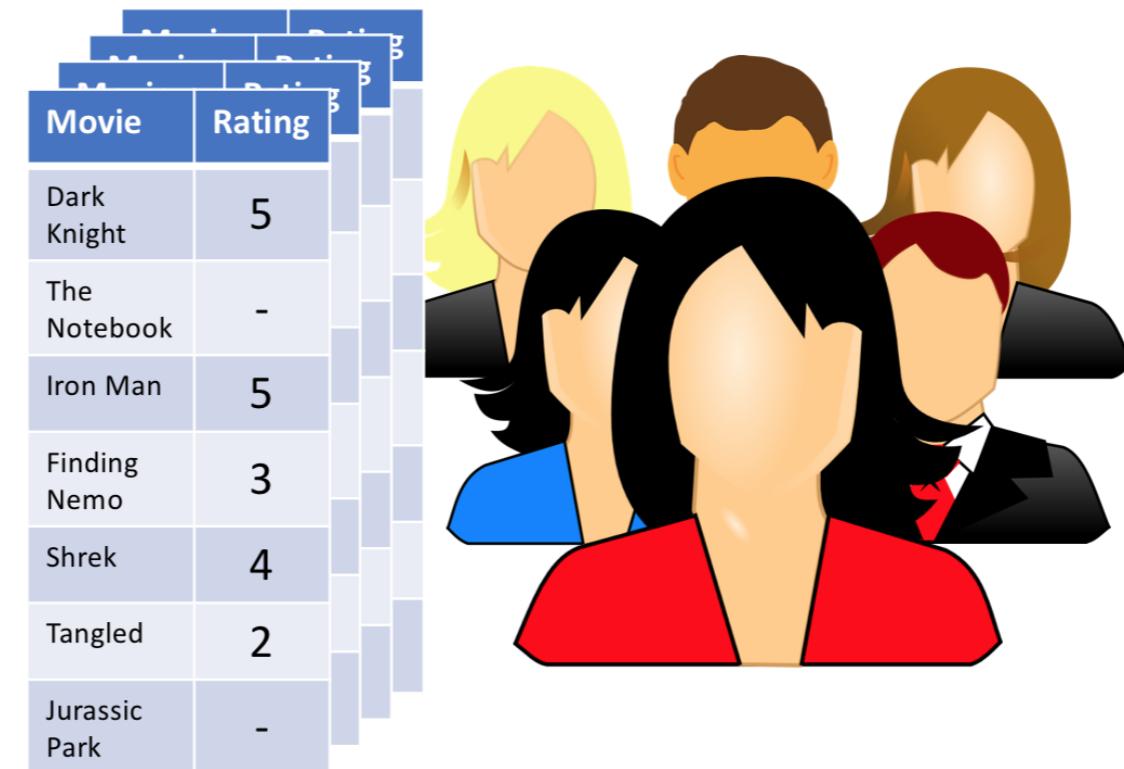


Movie	Rating
Dark Knight	5
The Notebook	-
Iron Man	4
Finding Nemo	3
Shrek	-
Tangled	1
Jurassic Park	4

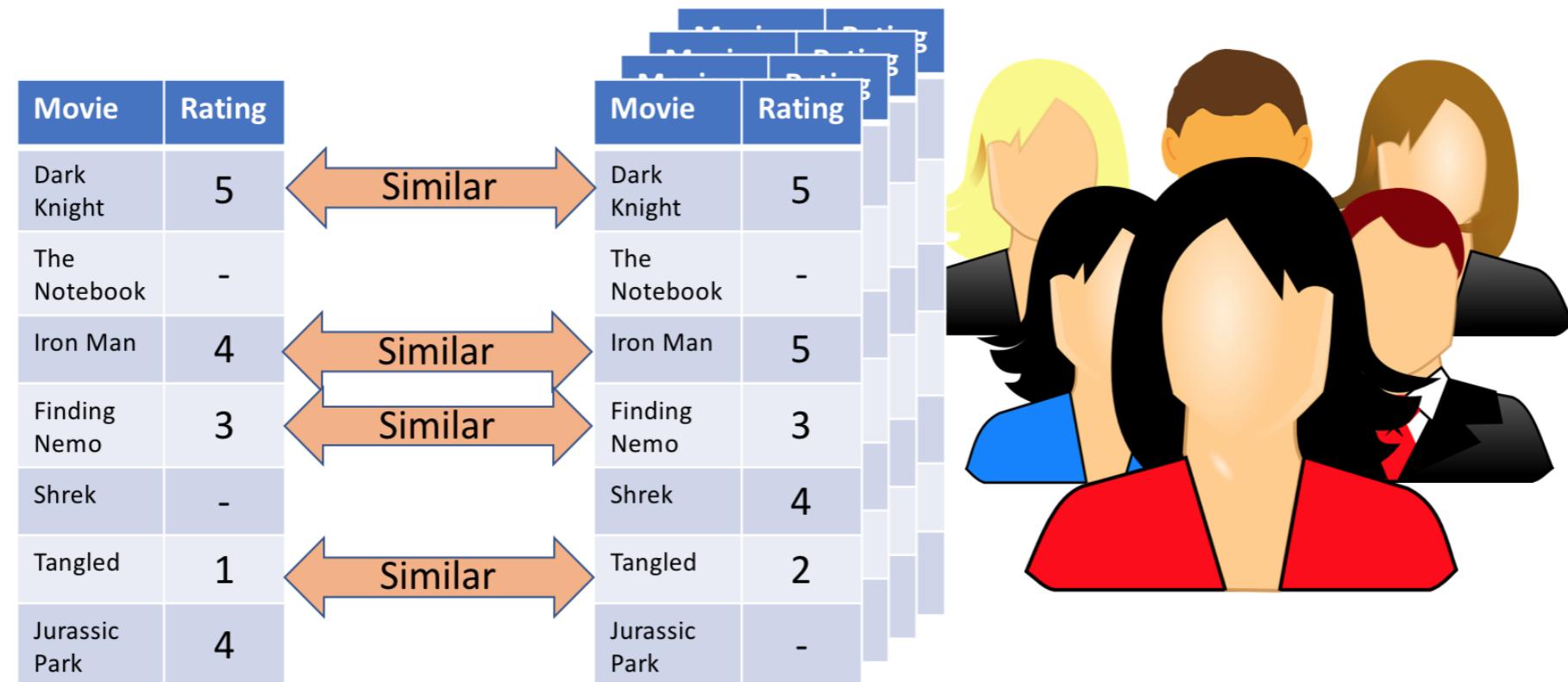
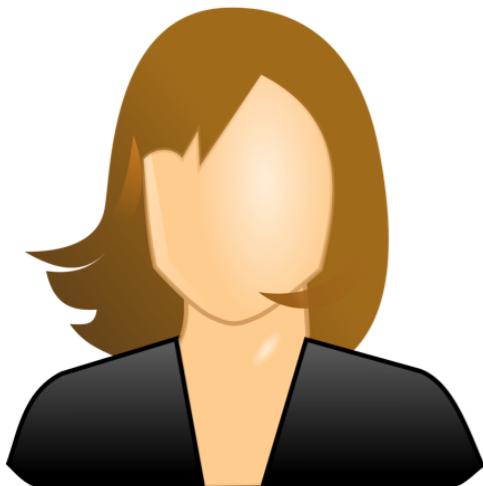
How recommendation engines work



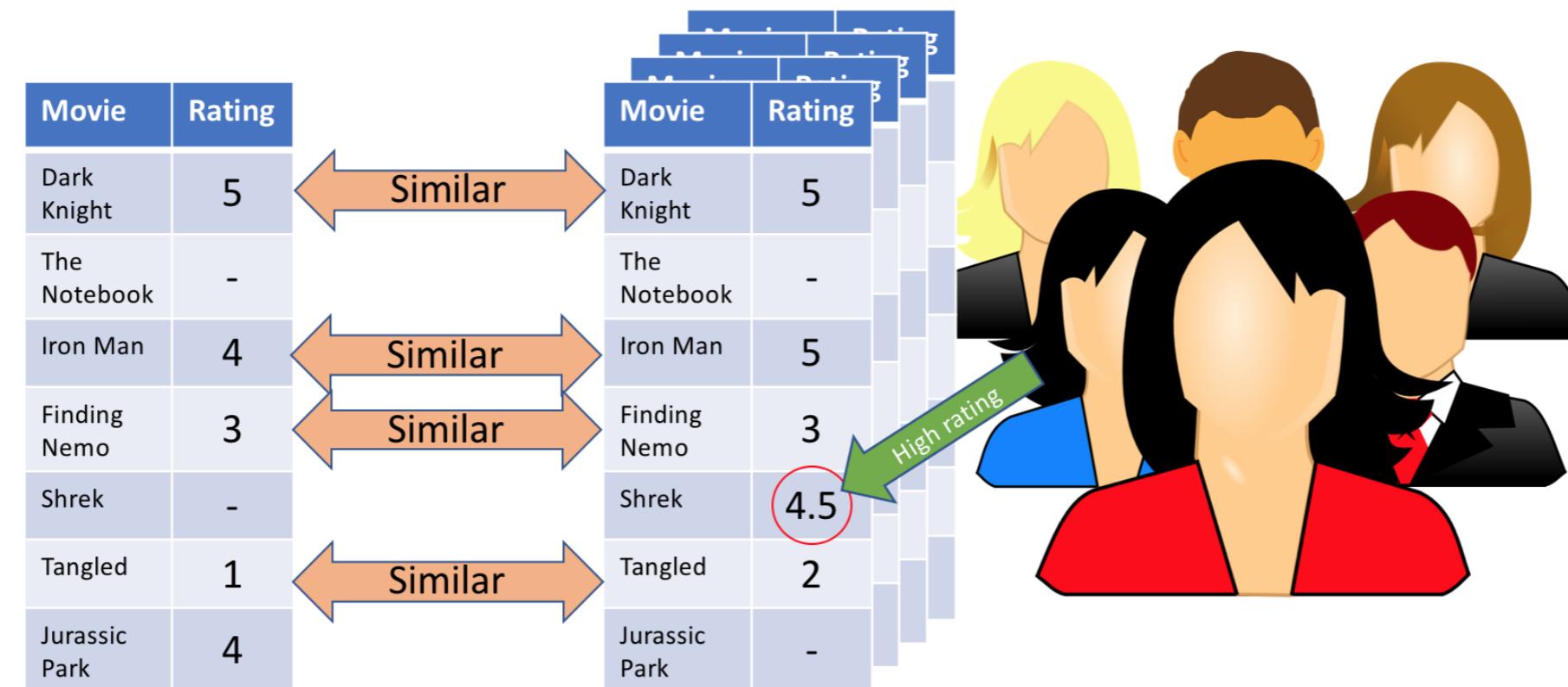
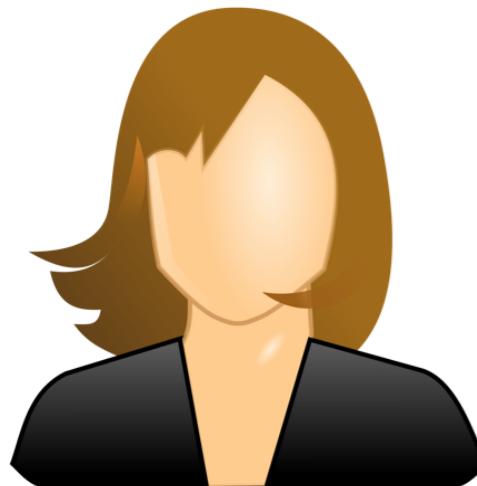
Movie	Rating
Dark Knight	5
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Iron Man	4
Finding Nemo	3
Shrek	-
Tangled	1
Jurassic Park	4



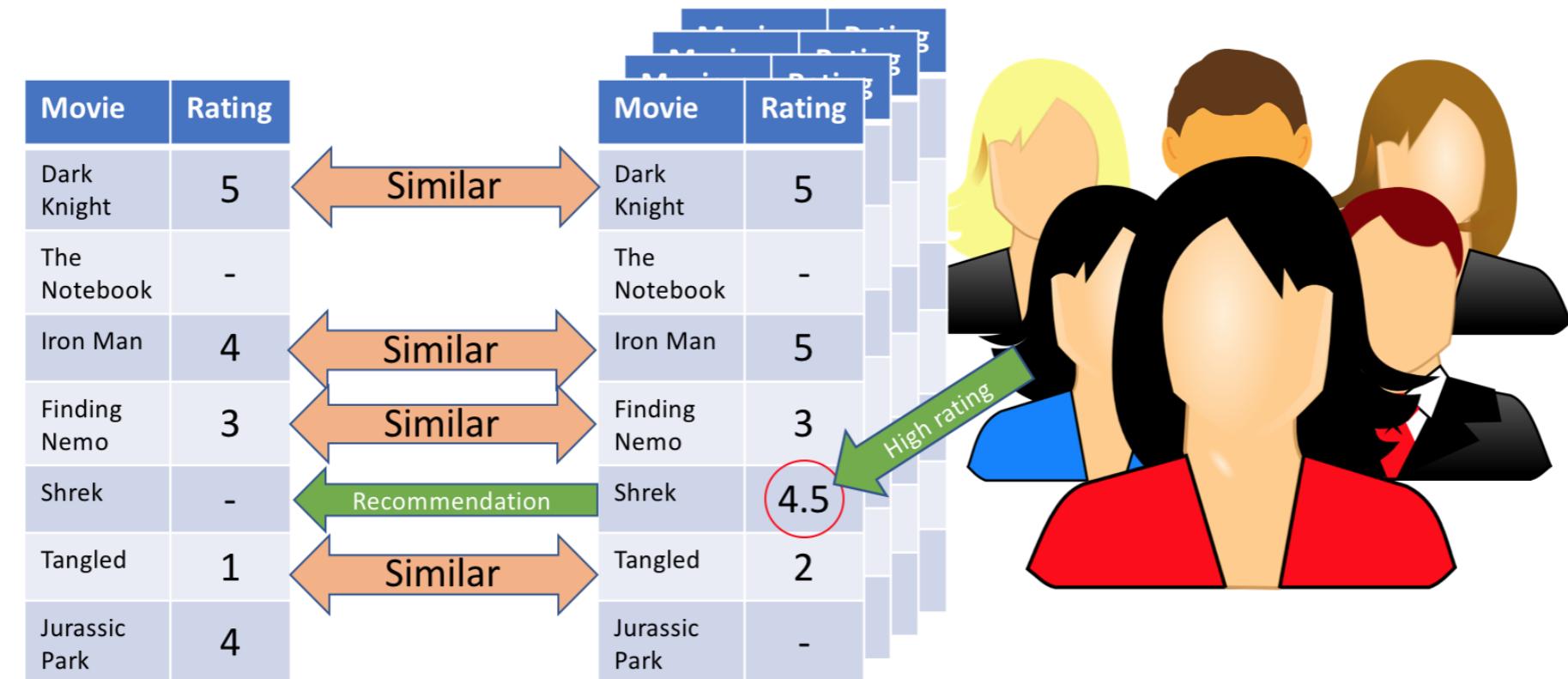
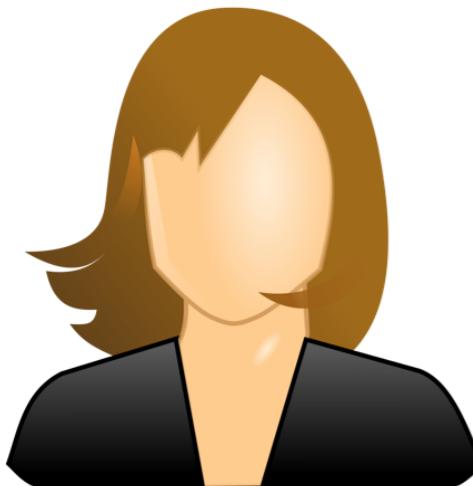
How recommendation engines work



How recommendation engines work



How recommendation engines work



The Power of Recommendation Engines

Powered by increasingly sophisticated models that analyze transaction data and digital footprints (for example, what topics are hot on social networks). Already, 35 percent of what consumers purchase on Amazon and 75 percent of what they watch on Netflix come from product recommendations based on such algorithms. Company-directed marketing is also competing for attention through social networks, user-generated content, and other channels. In fact, 24 percent of all consumer purchases are driven by recommendations for the average consumer.

Ian Mackenzie, Chris Meyer, and Steve Noble
McKinsey & Company, October 2013

Prerequisites





BUILDING RECOMMENDATION ENGINES WITH PYSPARK

Let's practice!



BUILDING RECOMMENDATION ENGINES WITH PYSPARK

**Collaborative filtering vs.
content-based filtering and
implicit vs explicit ratings**

Jamen Long
Data Scientist

Two types of recommendation engines:

CONTENT-BASED FILTERING

Based on features of items

- Genre: Comedy, Action, Drama
- Animation: Animated, Not animated
- Language: English, Spanish, Korean
- Decade Produced: 1950's, 1980's
- Actors: Meryl Streep, Tom Hanks

COLLABORATIVE FILTERING

Two types of recommendation engines

CONTENT-BASED FILTERING

Based on features of items



COLLABORATIVE FILTERING

Two types of recommendation engines

CONTENT-BASED FILTERING

Based on features of items

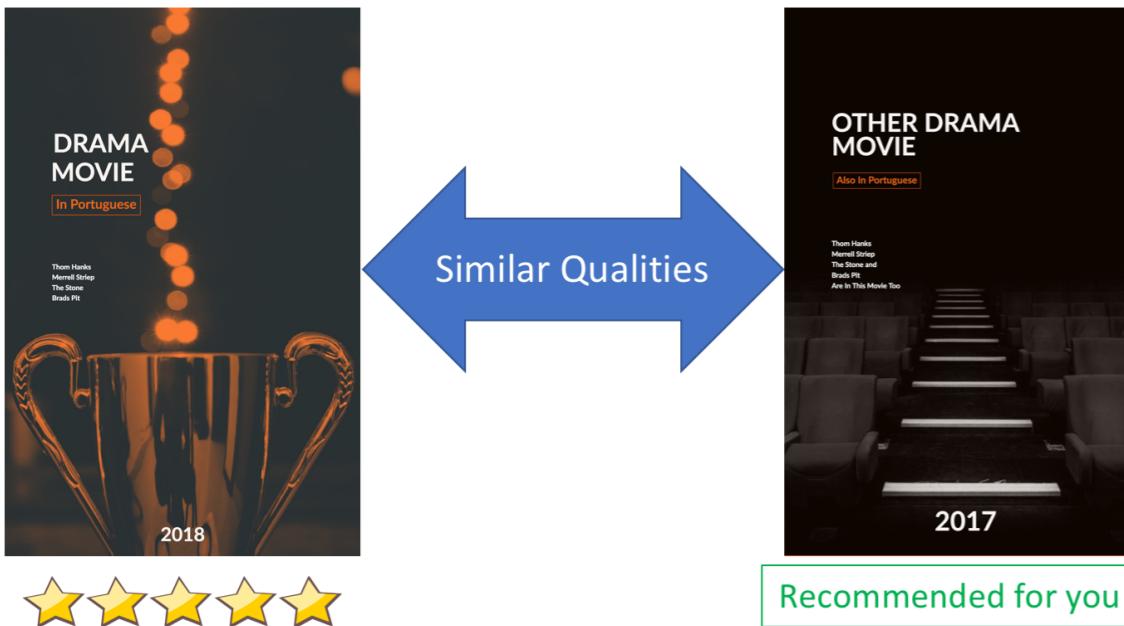


COLLABORATIVE FILTERING

Two types of recommendation engines

CONTENT-BASED FILTERING

Based on features of items

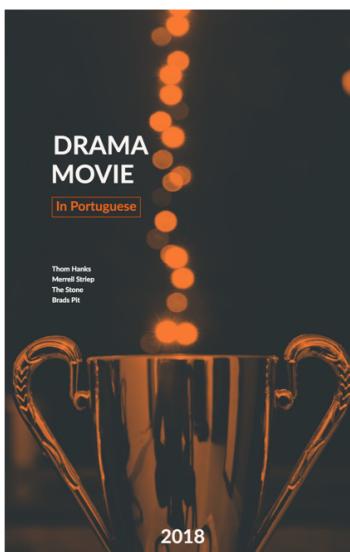


COLLABORATIVE FILTERING

Two types of recommendation engines

CONTENT-BASED FILTERING

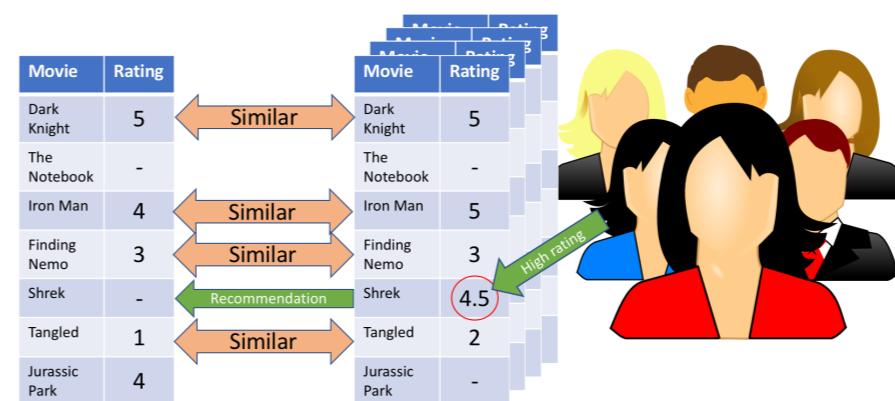
Based on features of items



Recommended for you!

COLLABORATIVE FILTERING

Based on similar user preferences



Two types of ratings

Two types of ratings

EXPLICIT RATINGS

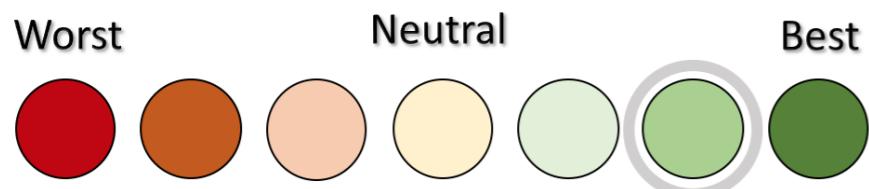
Two types of ratings

EXPLICIT RATINGS

IMPLICIT RATINGS

Two types of ratings

EXPLICIT RATINGS



IMPLICIT RATINGS

Two types of ratings

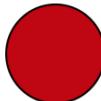
EXPLICIT RATINGS



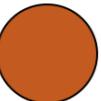
You rated this item:



Worst



Neutral



Best



IMPLICIT RATINGS



Two types of ratings

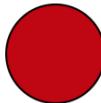
EXPLICIT RATINGS



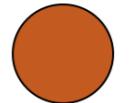
You rated this item:



Worst



Neutral



Best



IMPLICIT RATINGS



≈ Low Confidence Rating



Two types of ratings

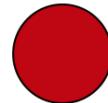
EXPLICIT RATINGS



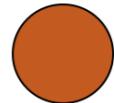
You rated this item:



Worst



Neutral



Best



IMPLICIT RATINGS



≈ Low Confidence Rating



≈ High Confidence Rating



BUILDING RECOMMENDATION ENGINES WITH PYSPARK

Let's practice!



BUILDING RECOMMENDATION ENGINES WITH PYSPARK

Alternate Uses for ALS

Jamen Long
Data Scientist

Original Ratings Matrix

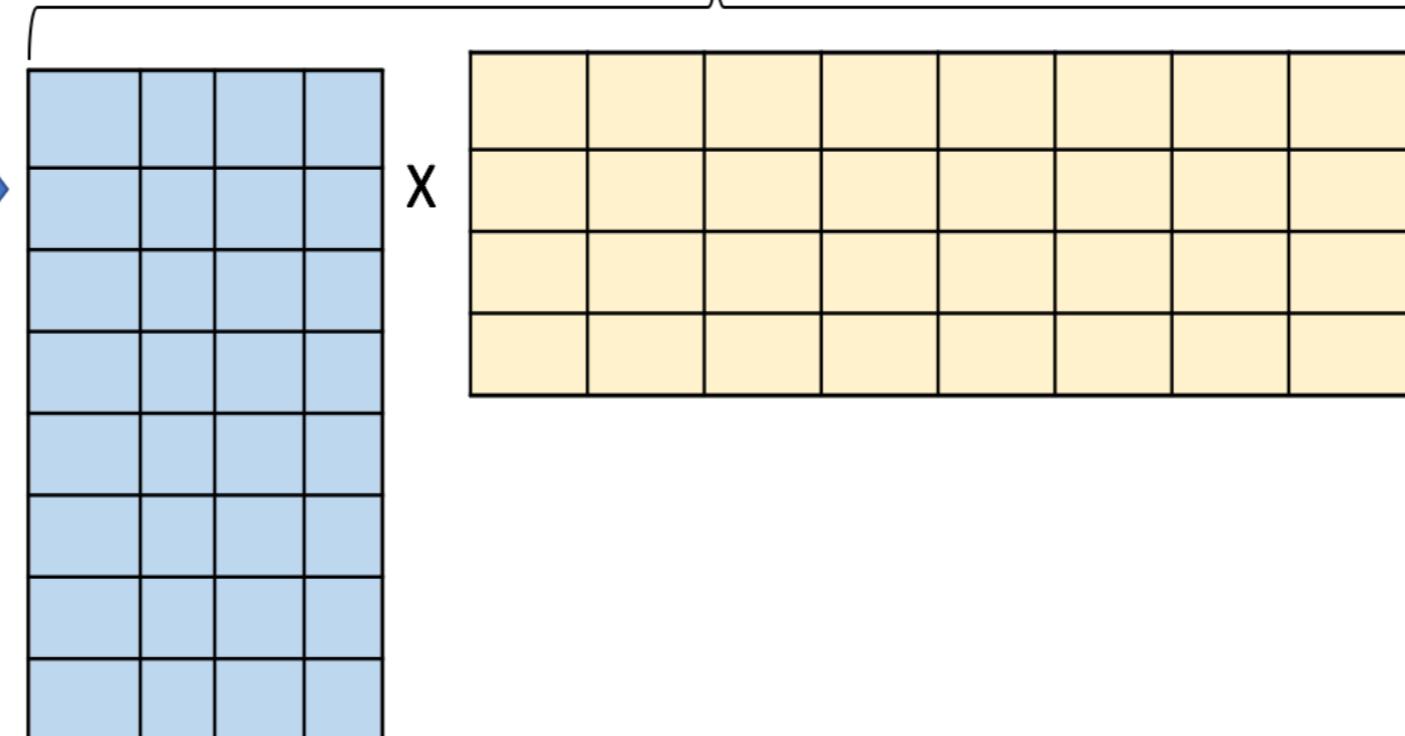
	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie <i>n</i>
User 1							
User 2							
User 3							
User 4							
User 5							
User...							
User <i>m</i>							

Original Ratings Matrix

	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie <i>n</i>
User 1							
User 2							
User 3							
User 4							
User 5							
User ...							
User <i>m</i>							

ALS

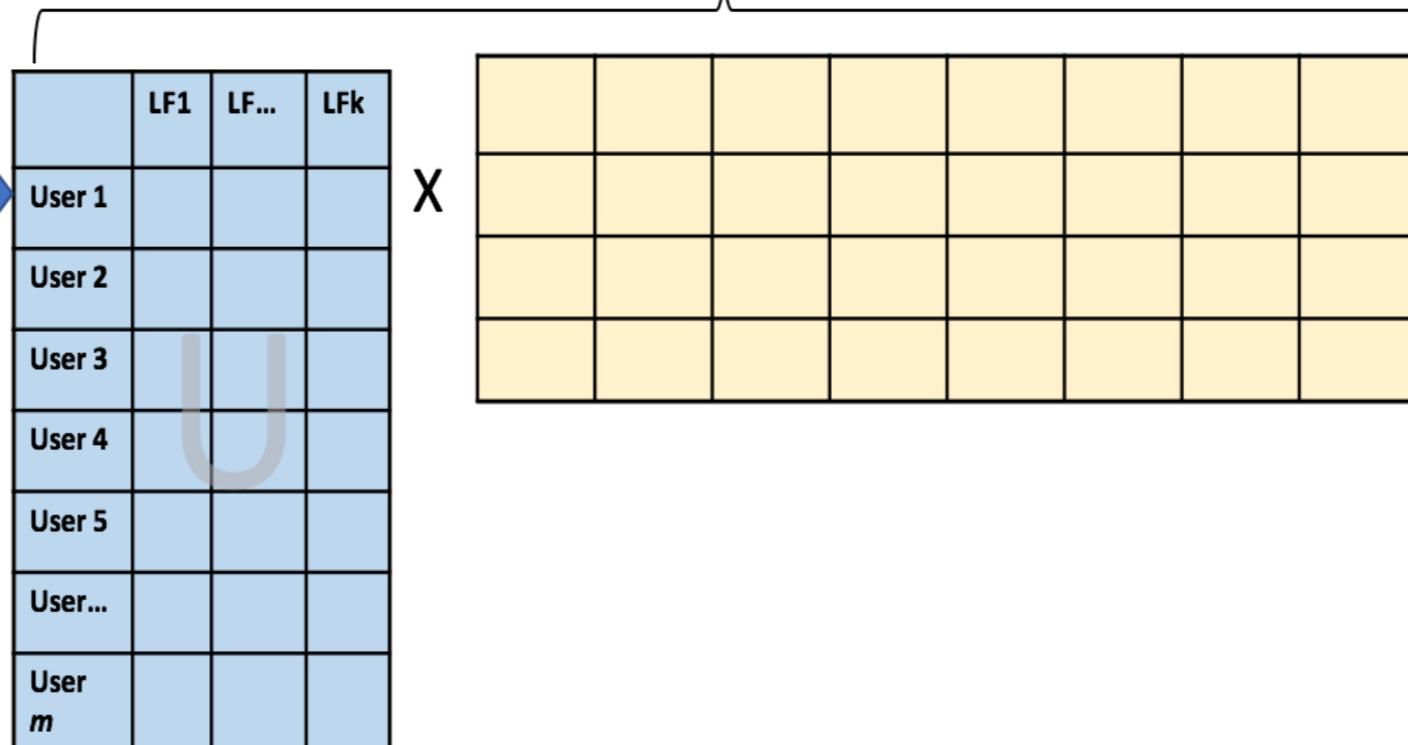
Factor Matrices



Original Ratings Matrix

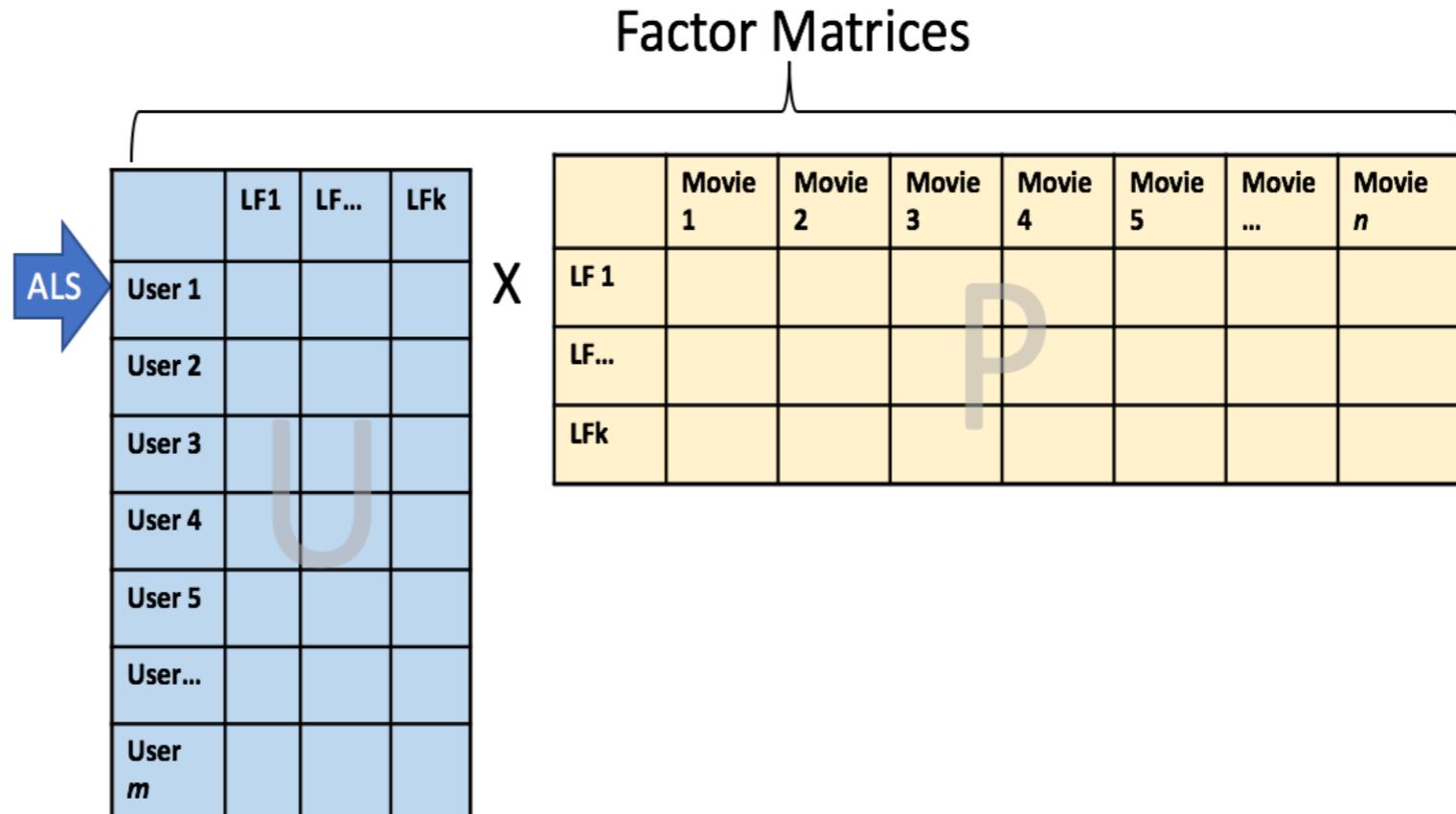
	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie <i>n</i>
User 1							
User 2							
User 3							
User 4							
User 5							
User ...							
User <i>m</i>							

ALS

Factor Matrices

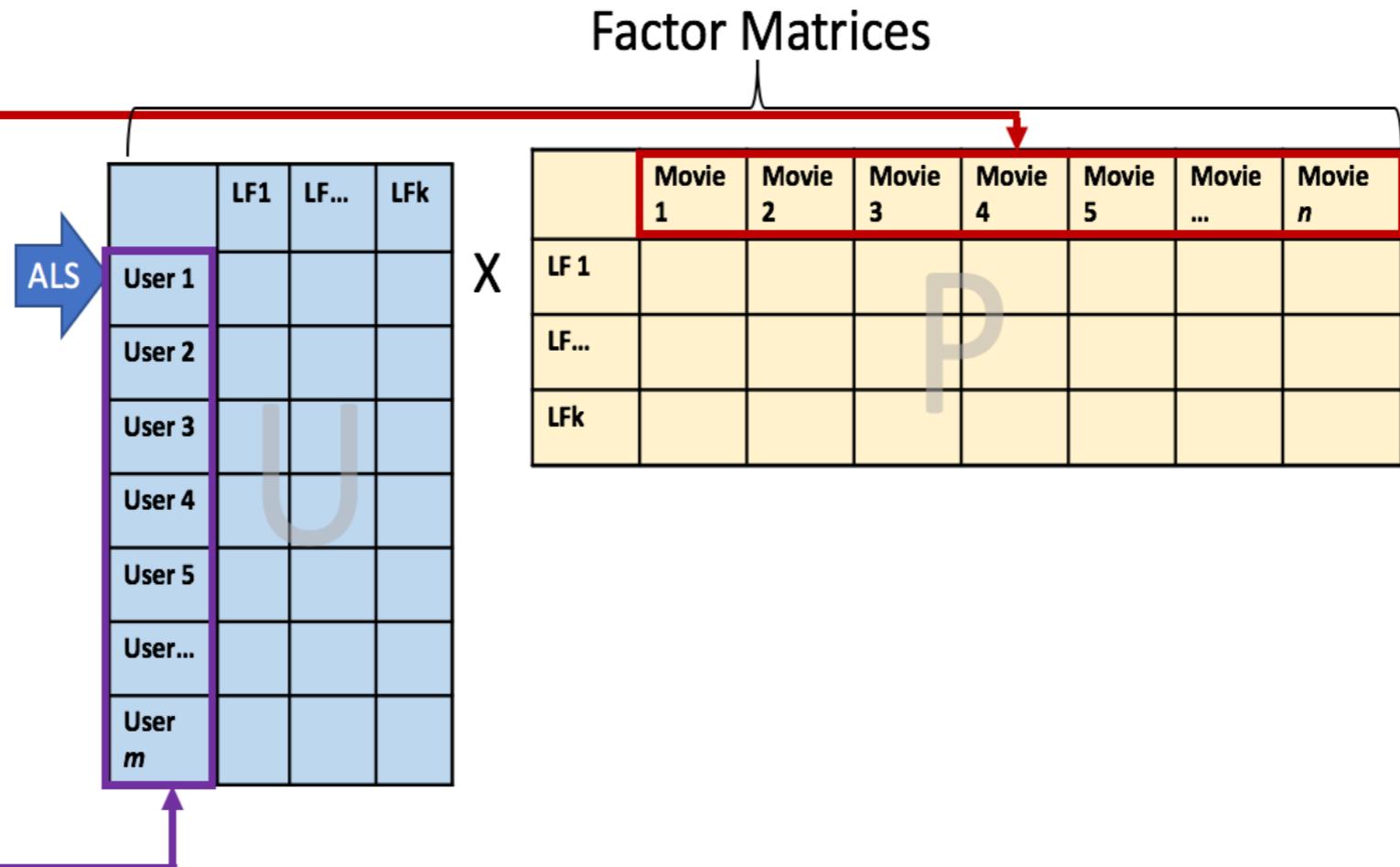
Original Ratings Matrix

	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie n
User 1							
User 2							
User 3							
User 4							
User 5							
User ...							
User m							



Original Ratings Matrix

	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie n
User 1							
User 2							
User 3							
User 4							
User 5							
User ...							
User m							

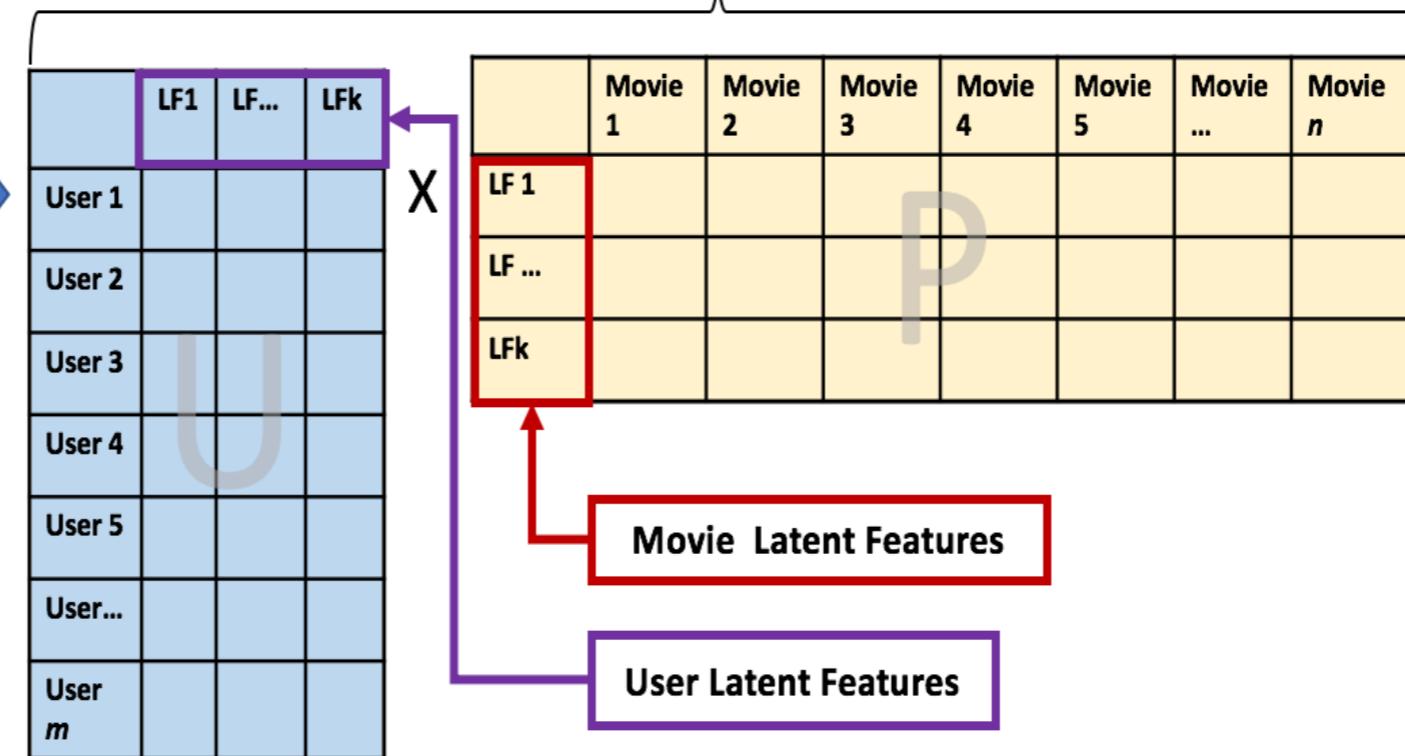


Original Ratings Matrix

	Movie 1	Movie 2	Movie 3	Movie 4	Movie 5	Movie ...	Movie n
User 1							
User 2							
User 3							
User 4							
User 5							
User ...							
User m							

ALS

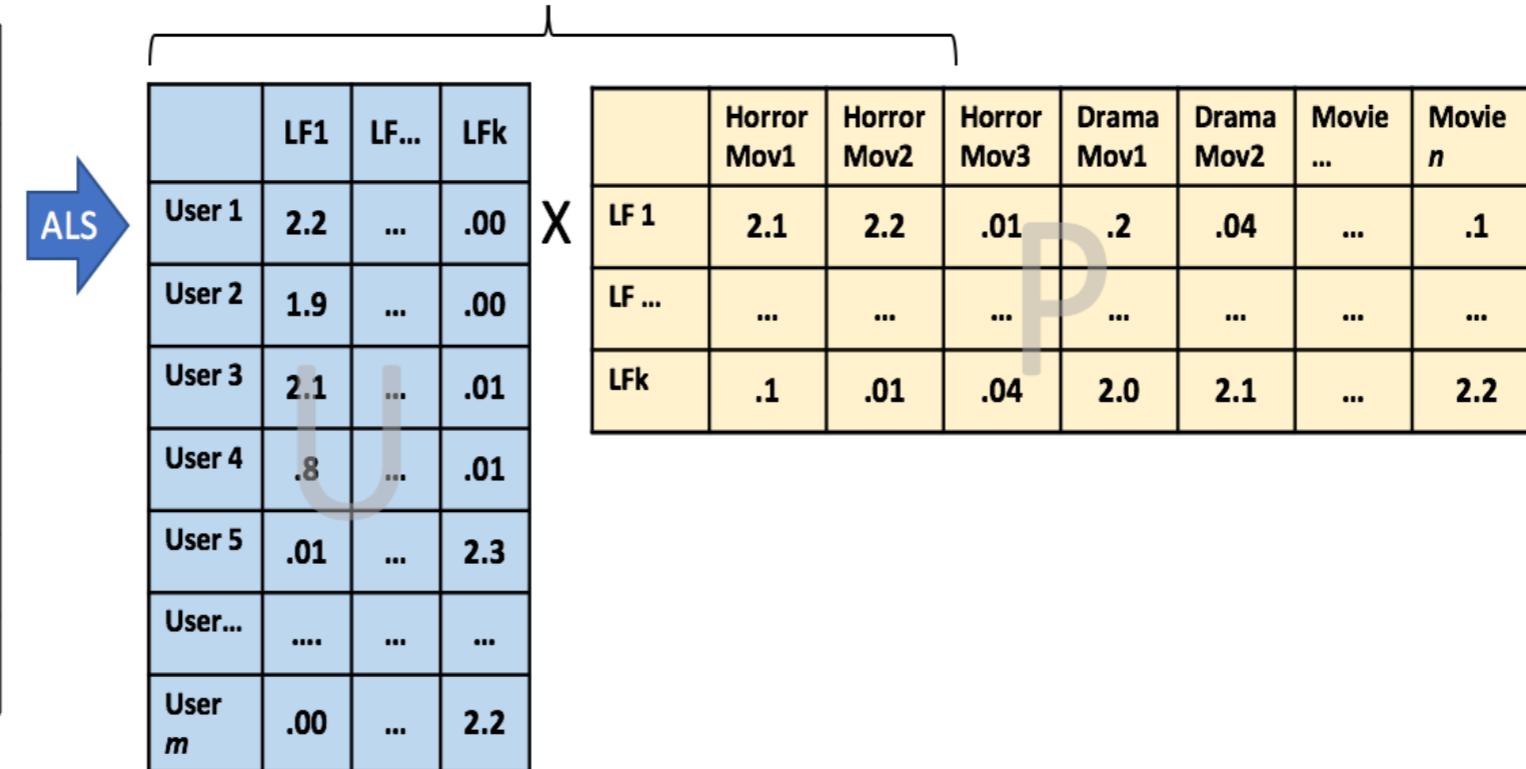
Factor Matrices



Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

Factor Matrices

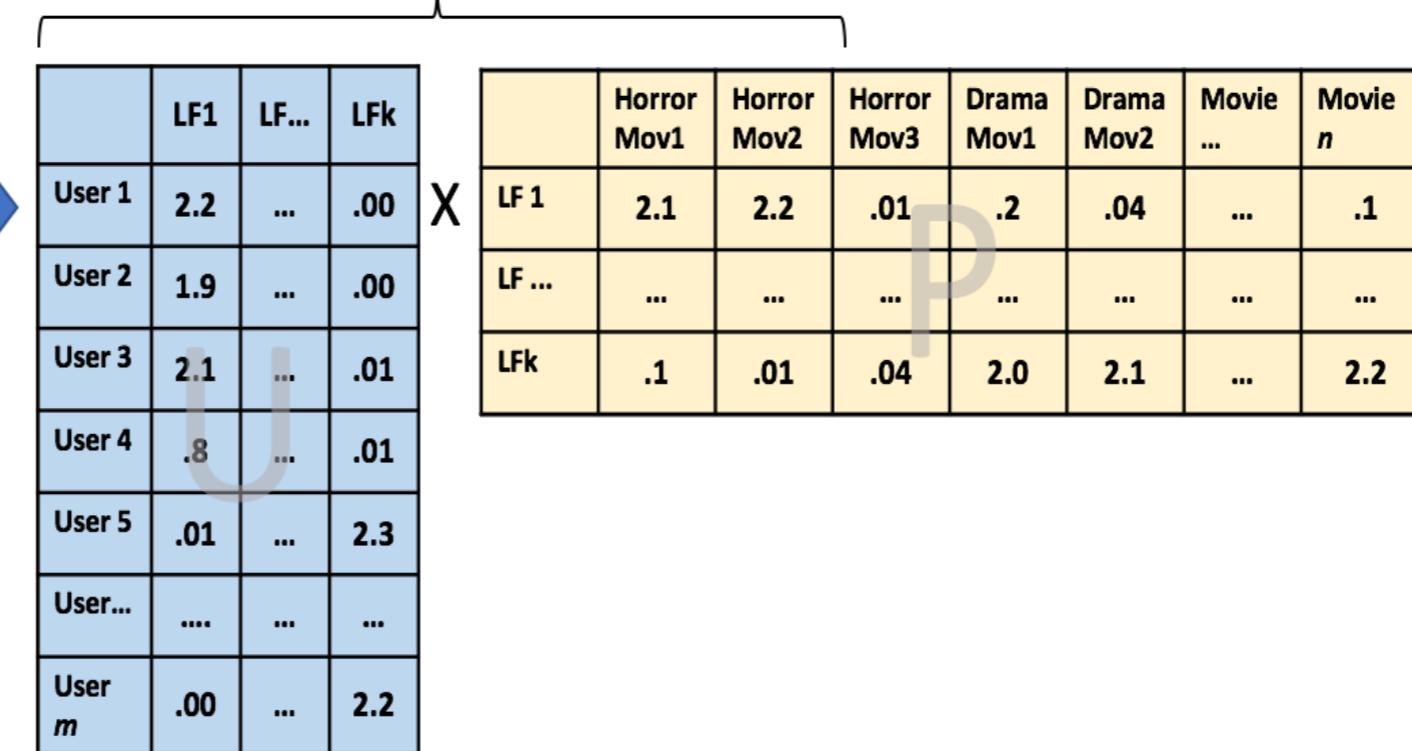


Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

ALS

Factor Matrices

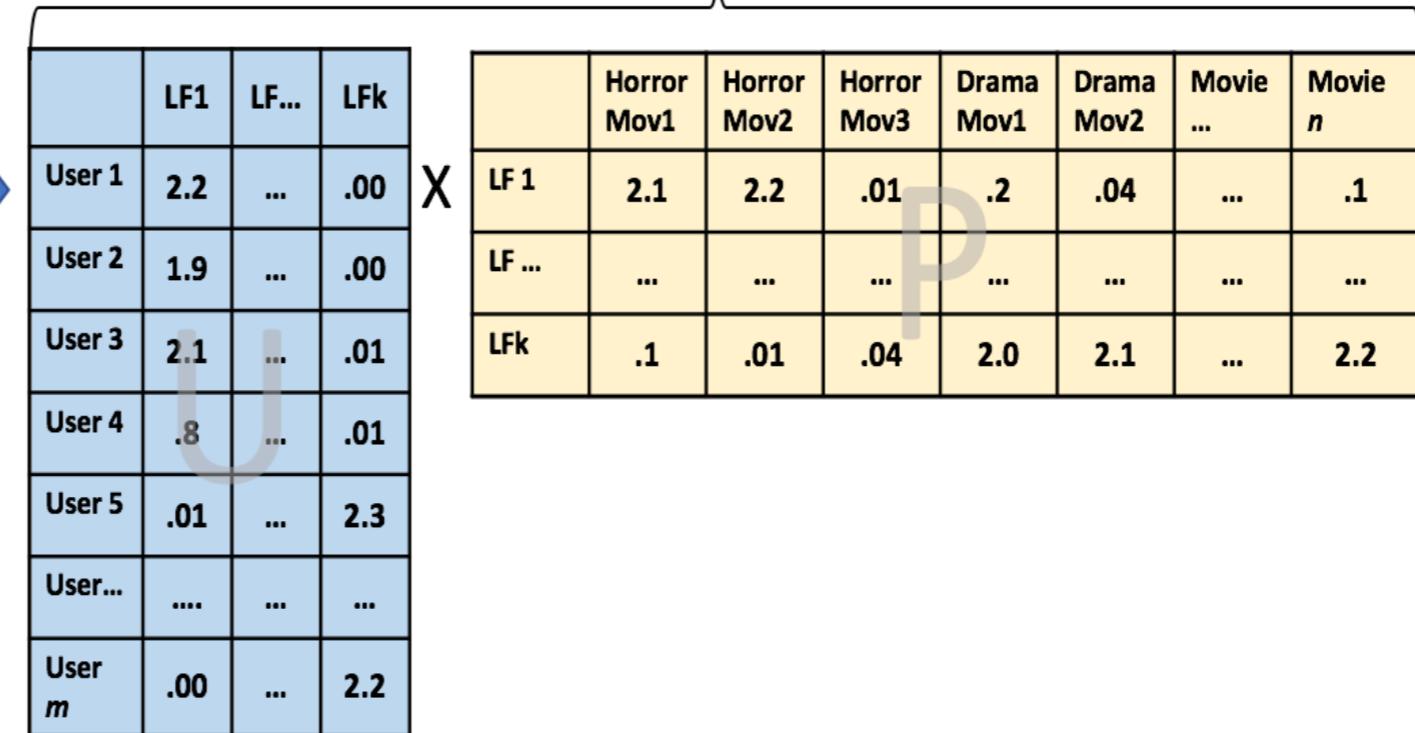


Original Ratings Matrix

	[No Title]	Horror	Horror	Horror	Drama	Drama	Movie	Drama
	Mov1	Mov2	Mov3	Mov1	Mov2	...	Mov1	MovN
User 1	5	5	1	1	1	...	1	
User 2	4	4	1	1	1	...	1	
User 3	4	5	1	1	1	...	1	
User 4	2	2	1	1	1	...	1	
User 5	1	1	1	5	5	...	5	
User...
User <i>m</i>	1	1	1	4	5	...	5	

ALS

Factor Matrices



Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

ALS

Factor Matrices

X

	LF1	LF...	LFk
User 1	2.200
User 2	1.900
User 3	2.101
User 4	.801
User 5	.01	...	2.3
User...
User <i>m</i>	.00	...	2.2

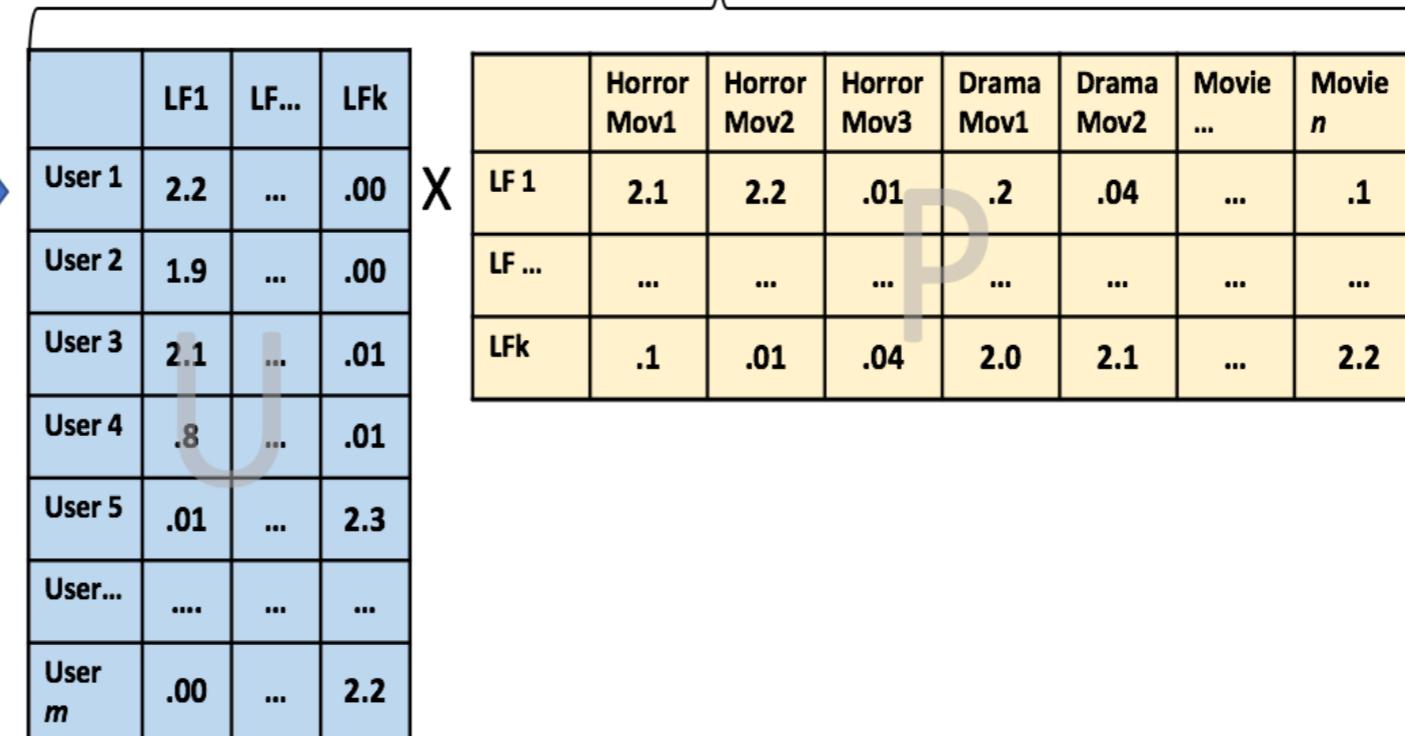
	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Movie <i>n</i>
LF 1	2.1	2.2	.01	.2	.041
LF
LFk	.1	.01	.04	2.0	2.1	...	2.2

Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

ALS

Factor Matrices

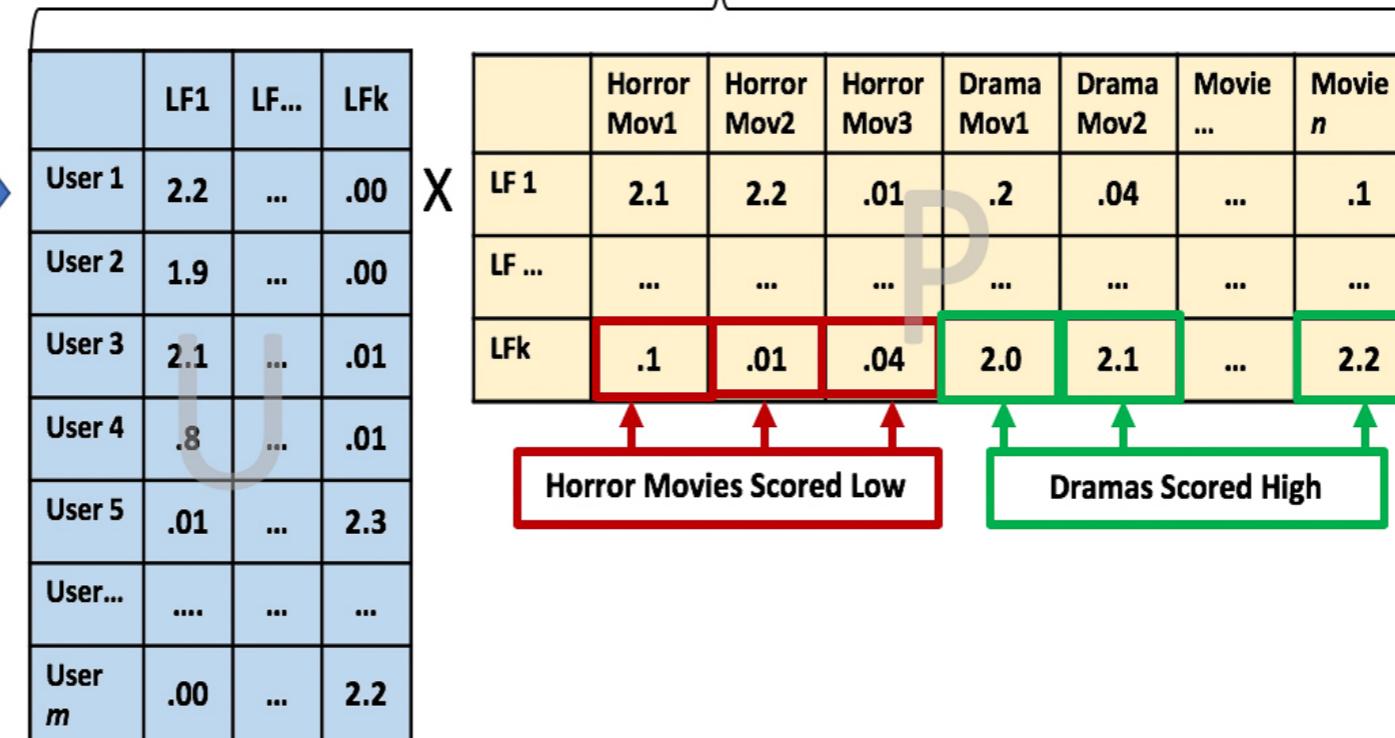


Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

ALS

Factor Matrices

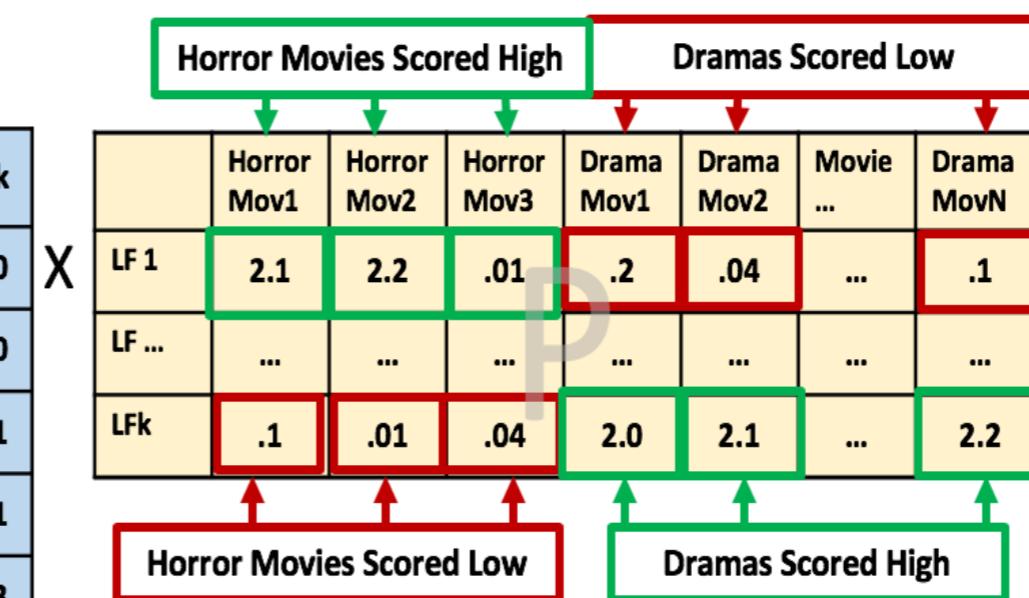


Original Ratings Matrix

	Horror Mov1	Horror Mov2	Horror Mov3	Drama Mov1	Drama Mov2	Movie ...	Drama MovN
User 1	5	5	1	1	1	...	1
User 2	4	4	1	1	1	...	1
User 3	4	5	1	1	1	...	1
User 4	2	2	1	1	1	...	1
User 5	1	1	1	5	5	...	5
User...
User <i>m</i>	1	1	1	4	5	...	5

ALS

	LF1	LF...	LFk
User 1	2.200
User 2	1.900
User 3	2.101
User 4	.801
User 5	.01	...	2.3
User...
User <i>m</i>	.00	...	2.2

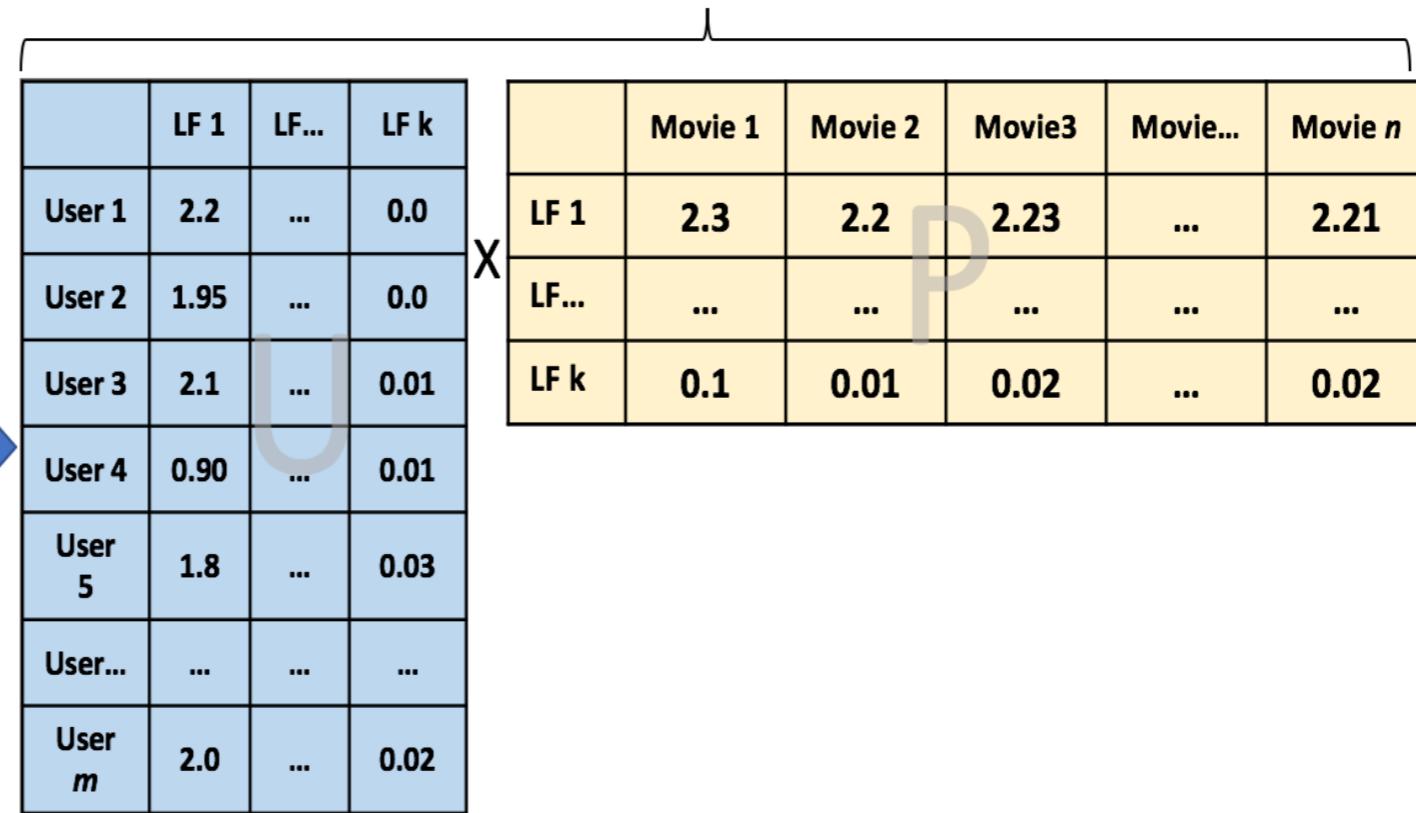


Original Matrix

	Movie 1	Movie 2	Movie3	Movie...	Movie <i>n</i>
User 1	5	4.8	4.9	...	4.9
User 2	4.5	4.3	4.4	...	4.3
User 3	4.8	4.6	4.7	...	4.6
User 4	2.1	1.98	2	...	2
User 5	4.1	4	4	...	4
User...
User <i>m</i>	4.6	4.4	4.5	...	4.4

ALS

Factor Matrices

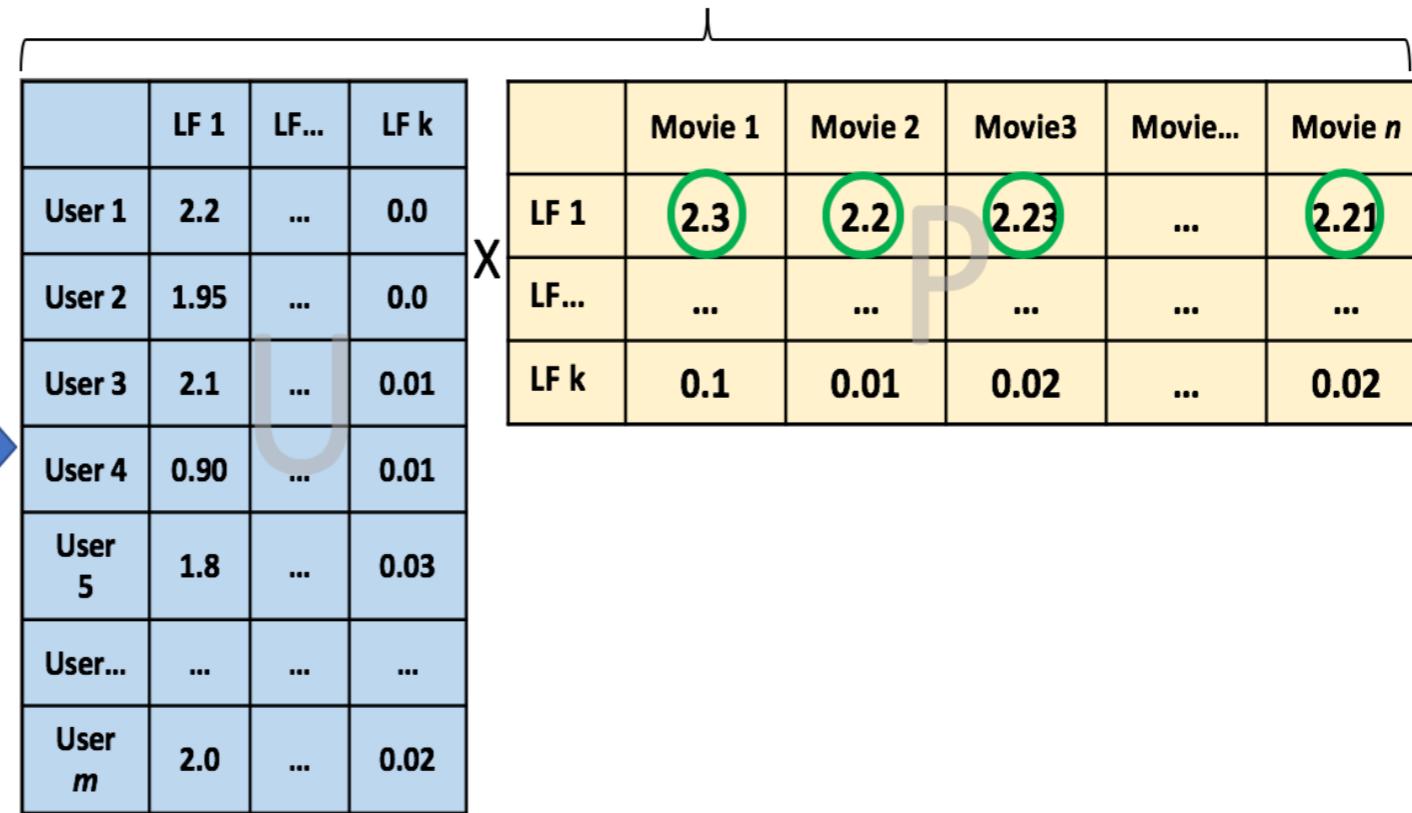


Original Matrix

	Movie 1	Movie 2	Movie3	Movie...	Movie <i>n</i>
User 1	5	4.8	4.9	...	4.9
User 2	4.5	4.3	4.4	...	4.3
User 3	4.8	4.6	4.7	...	4.6
User 4	2.1	1.98	2	...	2
User 5	4.1	4	4	...	4
User...
User <i>m</i>	4.6	4.4	4.5	...	4.4

ALS

Factor Matrices

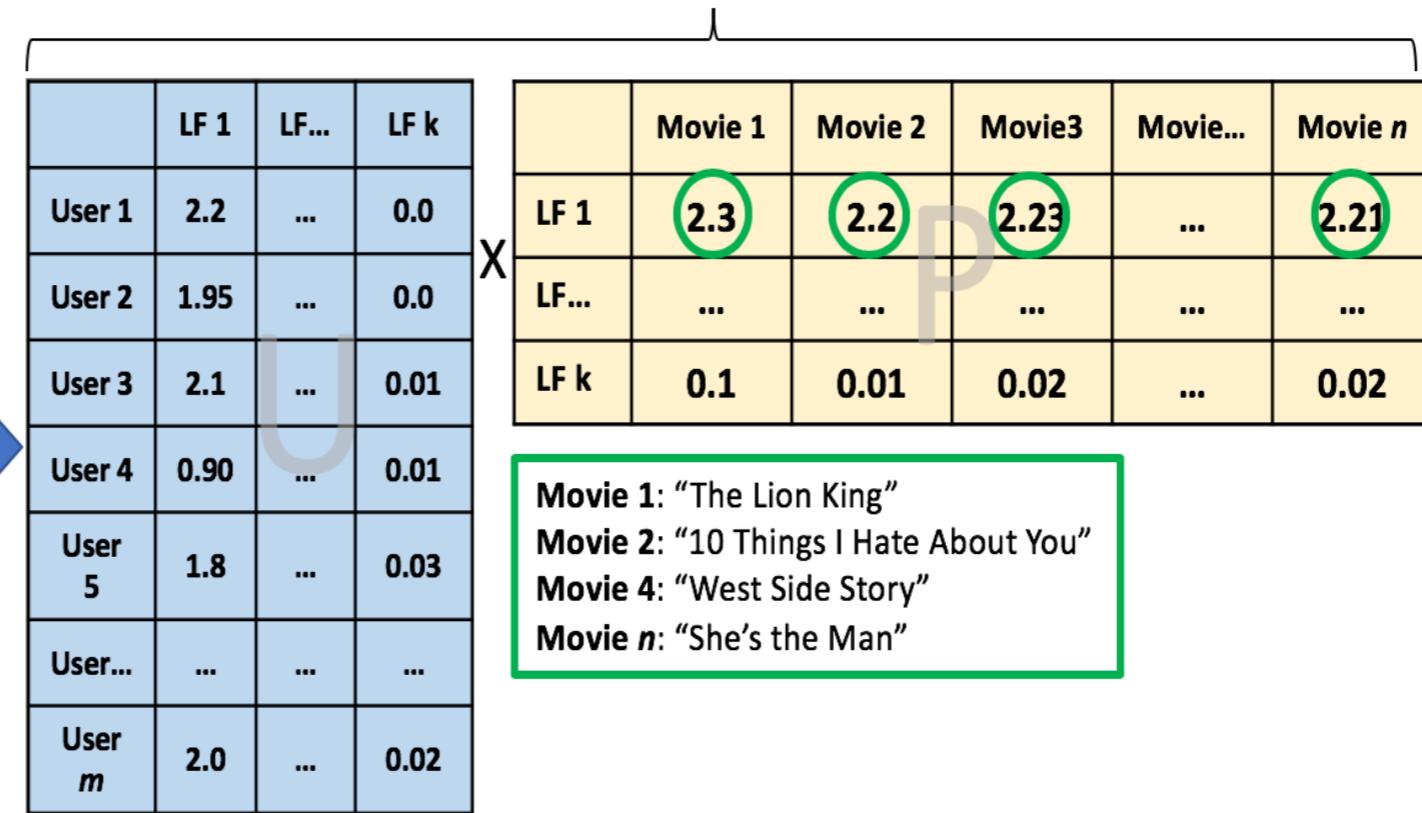


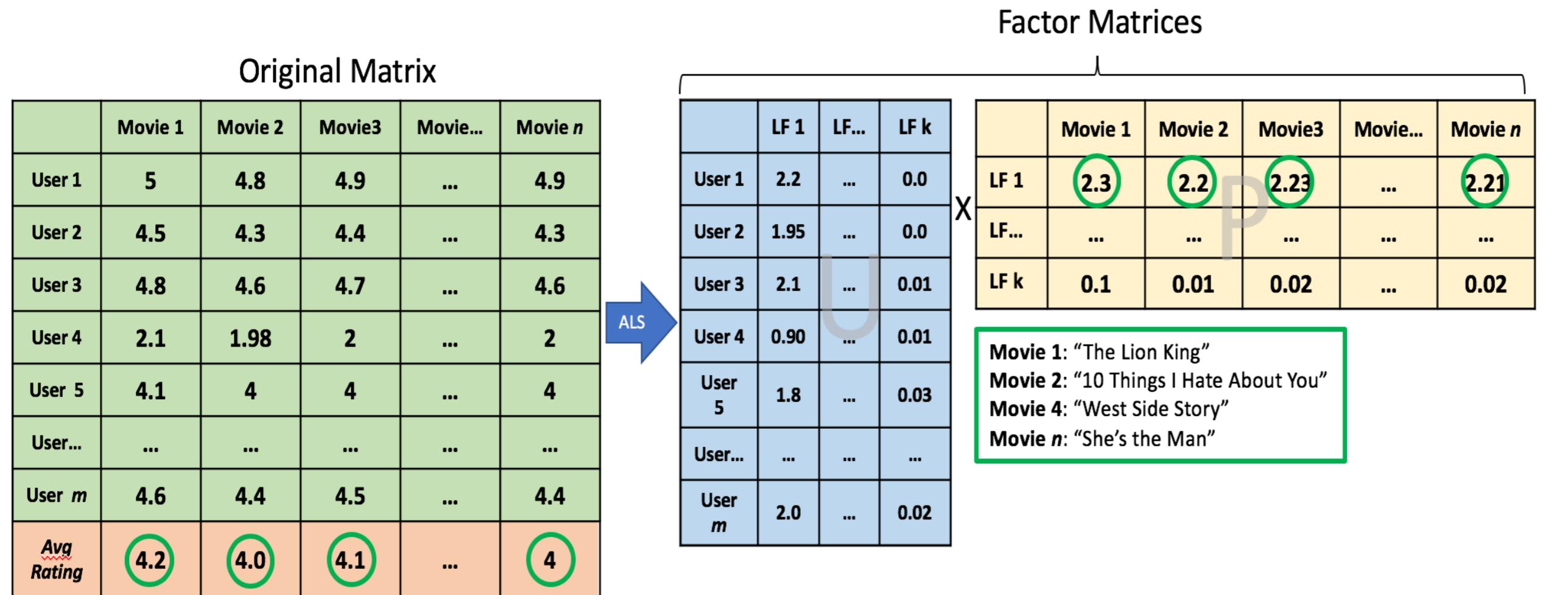
Original Matrix

	Movie 1	Movie 2	Movie3	Movie...	Movie <i>n</i>
User 1	5	4.8	4.9	...	4.9
User 2	4.5	4.3	4.4	...	4.3
User 3	4.8	4.6	4.7	...	4.6
User 4	2.1	1.98	2	...	2
User 5	4.1	4	4	...	4
User...
User <i>m</i>	4.6	4.4	4.5	...	4.4

ALS

Factor Matrices

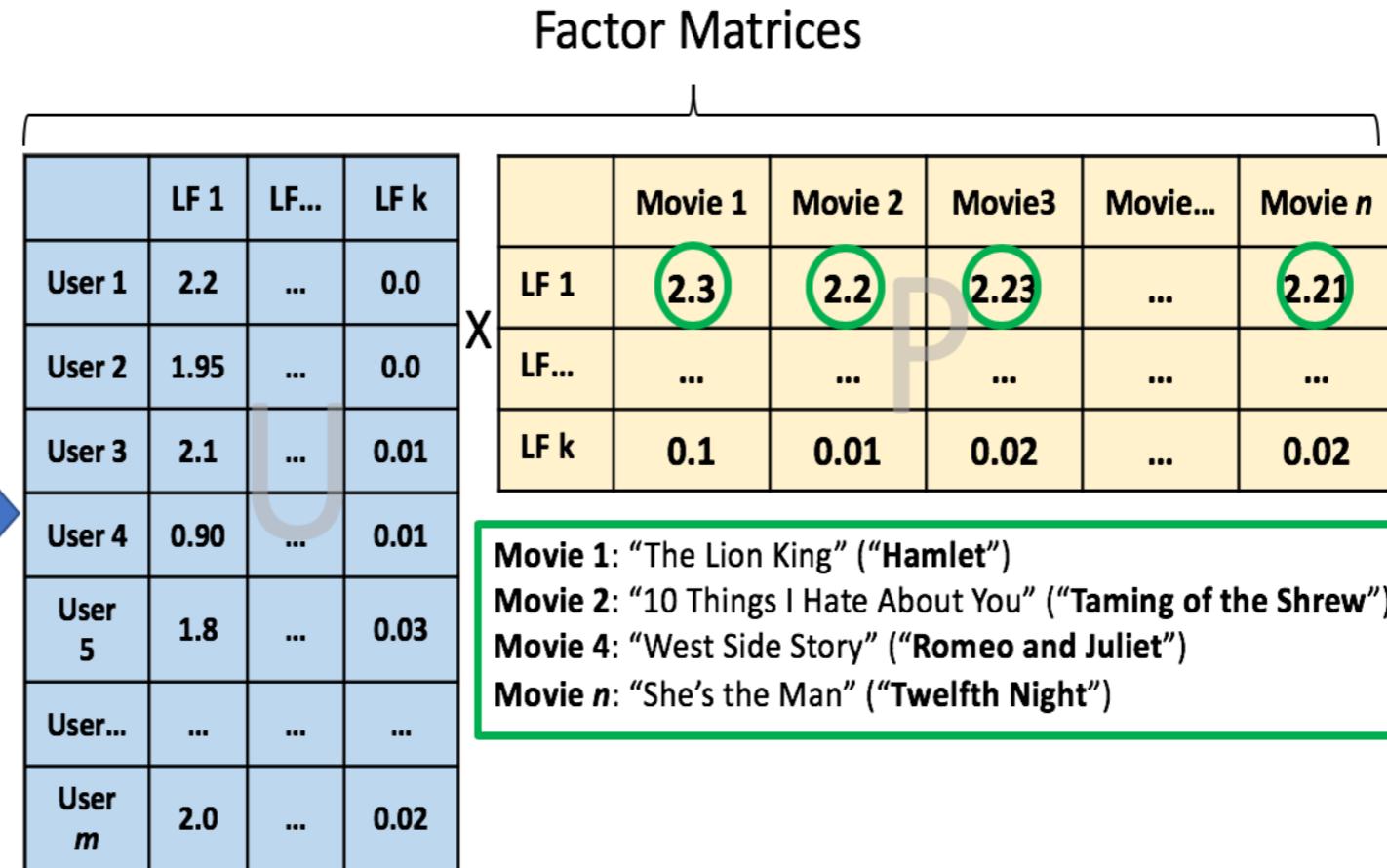




	Movie 1	Movie 2	Movie3	Movie...	Movie <i>n</i>
User 1	5	4.8	4.9	...	4.9
User 2	4.5	4.3	4.4	...	4.3
User 3	4.8	4.6	4.7	...	4.6
User 4	2.1	1.98	2	...	2
User 5	4.1	4	4	...	4
User...
User <i>m</i>	4.6	4.4	4.5	...	4.4
Avg Rating	4.2	4.0	4.1	...	4

ALS

Original Matrix





BUILDING RECOMMENDATION ENGINES WITH PYSPARK

Let's practice!