

Recommender system

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Recommender systems

Problem
Formulations

Content based
approach

Collaborative
Filtering

Types of Recommender systems

**Problem
Formulations**

Content based
approach

Collaborative
Filtering

Problem Formulation

Lets say you are in
Amazon or Netflix
companies

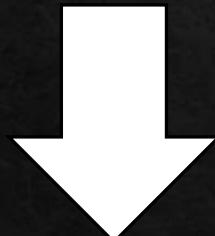
We want to predict
The series rating



Problem Formulation

We want to predict
The series rating

Series
name



House MD

The office

The modern

Family

Lucifer

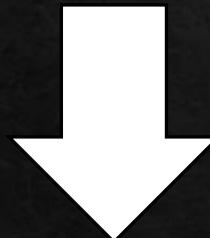


User rates
Series
using 1-5
stars

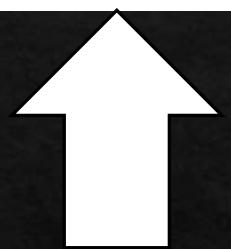
Problem Formulation

We want to predict
The series rating

Series
name



	Mohamed	Ahmed	Samar	Sara	
House MD	5	5	?	0	User rates
The office	?	4	3	0	Series
The modern Family	4	0	5	4	using 1-5 stars
Lucifer	0	0	0	?	Mohamed Elhaj-Abdou



Problem Formulation



Series name	Mohamed	Ahmed	Samar	Sara
House MD	5	5	?	0
The office	?	4	3	0
The modern Family	4	0	5	4
Lucifer	0	0	0	?

$N_u \rightarrow$ No. Users

4

Notations
 $N_m \rightarrow$ No. Movies

4

$r(i, j) \rightarrow$ 1 if user j has rates movie I
 0 other wise
 $y(i, j) \rightarrow$ rating (1-5) given by user j to movie I

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Problem Formulation

Notations

$N_u \rightarrow$ Number of users

$N_m \rightarrow$ Number of Movies

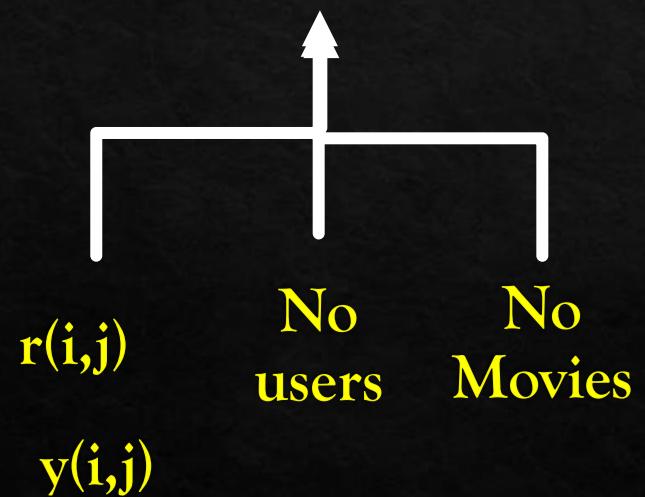
$r(i, j)$

$1 \rightarrow$ if user j has rates movie i
 $0 \rightarrow$ other wise

$y(i, j) \rightarrow$ rating (1-5) stars given by user j to movie i (defined only
of $r(i, j)=1$) \rightarrow this means that the user make a rates not ?

Problem Formulation

Given
Dataset



Looks all the
missing what
question marks
should be

We want to
predict the
rating (?) for
each user

As the question marks means
that the user did not watched
the movie

Problem Formulation

Depends on the predicted rating we can recommend the movie that have high rating to the user

Example

Series name	Mohamed	Ahmed	Samar	Sara
House MD	5	5	?	0
The office	?	4	3	0
The modern Family	4	0	5	4
Lucifer	0	0	0	?

Mohamed → likes mstry and family series types, because he likes (house and modern family

Mohamed → donot likes fantasy he gives to Lucifer zero rating

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Example

Series name	Mohamed	Ahmed	Samar	Sara
House MD	5	5	?	0
The office	4	4	3	0
The modern Family	4	0	5	4
Lucifer	0	0	0	?

Mohamed → likes mstry and family series types, because he likes (house and modern family

Mohamed → donot likes fantasy he gives to Lucifer zero rating

The office seems to be funny and family Mohamed Elhaj-Abdou

Example

Series name	Mohamed	Ahmed	Samar	Sara
House MD	5	2	?	0
The office	4	4	3	0
The modern Family	4	3	5	4
Lucifer	0	5	0	?

Ahmed → likes the office and, and likes fantasy so he likes Lucifer

Ahmed → seems little bit likes modern fam but not much

House Md seems to not be the best choice for him

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Recommender systems

Problem
Formulations

**Content based
approach**

Collaborative
Filtering

Content based R.S

Series name	Mohamed Ahmed	Samar	Sara
House MD	5	?	?
The office	?	4	3
The modern Family	4	3	5
Lucifer	0	5	?

How we can predict these ? Marks to recommend the right movie

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Content based R.S

Suppose that you have for each movie a set of features for them → lets say three

Funny

Family

Mystery

These features is a measurements for the degree of funny or family or mystery

Content based R.S

	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
House MD	5	?	?	0	0.3	0	0.9
The office	?	4	3	0	1	0.4	0
The modern Family	4	3	5	4	0.4	1	0
Lucifer	0	5	0	?	1	0	0.4

Content based R.S

	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
House MD	5	?	?	0	0.3	0	0.9
The office	?	4	3	0	1	0.4	0
The modern Family	4	3	5	4	0.4	1	0
Lucifer	0	5	0	?	1	0	0.4

Content based R.S

By this each movie can be represented with a feature vector

Feature vectors 1

House MD → X^1

Feature vectors 2

The office → X^2

$$X^1 = \begin{bmatrix} 0.3 \\ 0 \\ 0.9 \end{bmatrix}$$

$$X^2 = \begin{bmatrix} 1 \\ 0.4 \\ 0 \end{bmatrix}$$

Feature vectors 3

The modern Family → X^3

$$X^3 = \begin{bmatrix} 0.4 \\ 1 \\ 0 \end{bmatrix}$$

And So on

Content based R.S

For Each user j , learn a parameter $\theta^{(j)}$ predict user j as rating movie i with this following equation

$$\theta^{(j)T} \chi^i$$

Parameter to be optimized

Feature i

Separate linear regression problem for each user

Content based R.S

Lets take a concrete example

	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
Mohamed	$\theta^{(1)}$	$\theta^{(2)}$	$\theta^{(3)}$	$\theta^{(4)}$	0.3	0	0.9
The office	5	?	?	0	1	0.4	0
The modern Family	?	4	3	0	0.4	1	0
Lucifer	4	3	5	4	1	0	0.4
	0	5	0	?			

The parameters associated with each user

Now lets make a prediction for Mohamed about the office

	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
Mohamed	$\theta^{(1)}$	$\theta^{(2)}$	$\theta^{(3)}$	$\theta^{(4)}$	0.3	0	0.9
House MD	5	?	?	0			
The office	?	4	3	0	1	0.4	0
The modern Family	4	3	5	4	0.4	1	0
Lucifer	0	5	0	?	1	0	0.4

Now lets make a prediction for Mohamed about the office

$$X^2 = \begin{bmatrix} 1 \\ 0.4 \\ 0 \end{bmatrix} \quad \theta^{(1)} = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

The office parameters

Some how the parameters are arrived with this values. Later we will talking about it how we can get it

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	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
House MD	$\theta^{(1)}$ 5	$\theta^{(2)}$?	$\theta^{(3)}$?	$\theta^{(4)}$ 0	0.3	0	0.9
The office	2	4	3	0	1	0.4	0
The modern Family	4	3	5	4	0.4	1	0
Lucifer	0	5	0	?	1	0	0.4

Now lets make a prediction for Mohamed about the office

$$X^2 = \begin{bmatrix} 1 \\ 0.4 \\ 0 \end{bmatrix} \quad \theta^{(1)} = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix} \rightarrow$$

$$\theta^{(1)T} X^3 = 0.4 * 5 = 2$$

The office parameters

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Problem Formulation

$r(i, j) \rightarrow 1$ if user j has rates movie i ($0 \rightarrow$ otherwise)

$y(i, j) \rightarrow$ rating by user j to movie i (if defined)

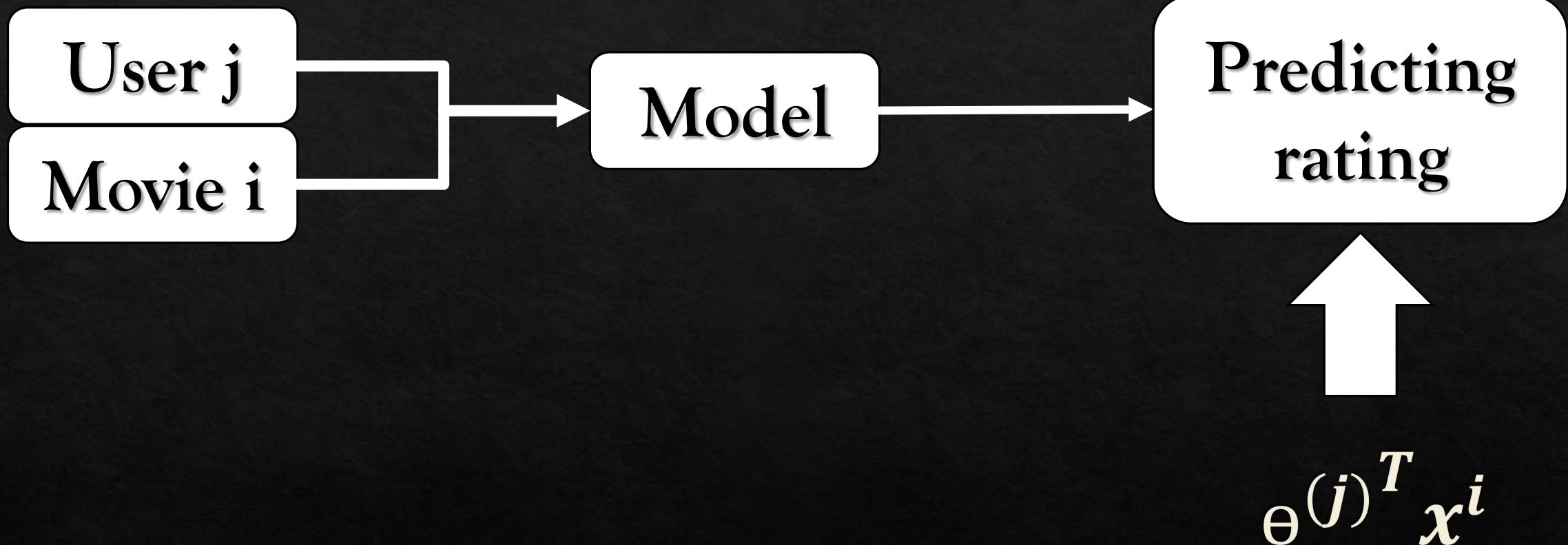
$\theta^{(j)}$ → parameter vector for user j

$x^{(i)}$ → feature vector for movie i

$N_u \rightarrow$ Number of users

$N_m \rightarrow$ Number of Movies

$m^{(j)} \rightarrow$ Number of Movies rated by user j



to learn $\theta^{(j)}$ we want to minimize the parameter theta we need to recap linear regression looks like

$$\text{Min } \theta^{(j)} \rightarrow \frac{1}{2m^{(j)}} \sum_{i:r(i,j)=1}^m \left(\theta^{(j)T} x^i - y_{true} \right)^2$$

Summing over all the movies that user j has rated

Prediction

True $y(i, j)$

The diagram illustrates the cost function for learning $\theta^{(j)}$. It shows the formula $\frac{1}{2m^{(j)}} \sum_{i:r(i,j)=1}^m \left(\theta^{(j)T} x^i - y_{true} \right)^2$. A red box surrounds the summation index i in the term $\sum_{i:r(i,j)=1}$. An arrow points from the text "Summing over all the movies that user j has rated" to this red box. Another arrow points from the text "True $y(i, j)$ " to the term y_{true} . The word "Prediction" is centered below the summation symbol.

In content based Recommender system we need to make some modifications

$$\cancel{\text{Min } \theta^{(j)} \rightarrow \frac{1}{2m^{(j)}} \sum_{i:r(i,j)=1}^m \left(\theta^{(j)T} x^i - y_{true} \right)^2}$$

We want to learn the parameters for N_u users

$$\text{Min } \theta^{(j)} \rightarrow \frac{1}{2N_u} \sum_{j=1}^{N_u} \sum_{i:r(i,j)=1}^m \left(\theta^{(j)T} x^i - y_{true} \right)^2$$

In content based Recommender system we need to make some modifications

$$\text{Min } \theta^{(j)} \rightarrow \frac{1}{2m^{(j)}} \sum_{j=1}^{N_u} \sum_{i:r(i,j)=1}^m \left(\theta^{(j)T} x^i - y_{true} \right)^2$$

Making prediction for all the users parameters ($\theta^{(j)}$) for all the movies

And then use the gradient decent

Content base approach need x_{features} that represents how much the movie is action or family or funny

The problem is that its hard and rare to find a movies with this kind of data

collaborative filtering based approach is found

Recommender systems

Problem
Formulations

Content based
approach

**Collaborative
Filtering**

Collaborative filtering

In Collaborative filtering we have no idea about the features

In Collaborative filtering use inferencing method to find the features

	Mohamed θ^1	Ahmed θ^2	Samar θ^3	Sara θ^4	X1 Funny	X2 Family	X3 Mystery
House MD	5	?	?	0	?	?	?
The office	?	5	3	0	?	?	?
The modern Family	4	3	5	4	?	?	?
Lucifer	0	0	0	?	?	?	?

$$\theta^1 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

$$\theta^2 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

$$\theta^3 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

$$\theta^4 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

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	Mohamed θ^1	Ahmed θ^2	Samar θ^3	Sara θ^4	X1 Funny	X2 Family	X3 Mystery
House MD	5	5	?	0	?	?	?
The office	?	5	3	0	?	?	?
The modern Family	4	?	5	4	?	?	?
Lucifer	0	0	0	?	?	?	?

$$\theta^1 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

Mohamed likes mstry

$$\theta^2 = \begin{bmatrix} 5 \\ 0 \\ 0 \end{bmatrix}$$

Ahmed likes Funny

$$\theta^3 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Samar likes family

$$\theta^4 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Sara likes family
Mohamed Elhaj-Abdou

	Mohamed θ^1	Ahmed θ^2	Samar θ^3	Sara θ^4	X1 Funny	X2 Family	X3 Mystery
House MD	5	5	?	0	?	?	?
The office	?	5	3	0	?	?	?
The modern Family	4	?	5	5	?	?	?
Lucifer	0	0	0	?	?	?	?

$$\theta^1 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

Mohamed likes mstry

$$\theta^2 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

Ahmed likes mstry

$$\theta^3 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Samar likes family

$$\theta^4 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Sara likes family
Mohamed Elhaj-Abdou

	Mohamed θ^1	Ahmed θ^2	Samar θ^3	Sara θ^4	X1 Funny	X2 Family	X3 Mystery
House MD	5	5	?	0	?	?	?
The office	?	5	3	0	?	?	?
The modern Family	4	?	5	5	?	?	?
Lucifer	0	0	0	?	?	?	?

$$\theta^1 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

Mohamed likes mstry

$$\theta^2 = \begin{bmatrix} 0 \\ 0 \\ 5 \end{bmatrix}$$

Ahmed likes mstry

$$\theta^3 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Samar likes family

$$\theta^4 = \begin{bmatrix} 0 \\ 5 \\ 0 \end{bmatrix}$$

Sara likes family
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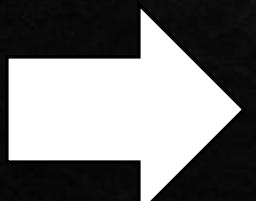
	Mohamed	Ahmed	Samar	Sara	X1 Funny	X2 Family	X3 Mystery
	θ^1	θ^2	θ^3	θ^4			
House MD	5	5	?	0	0	0	1
The office	?	5	3	0	?	?	?
The modern Family	4	?	5	5	?	?	?
Lucifer	0	0	0	?	?	?	?

Mohamed likes mstry

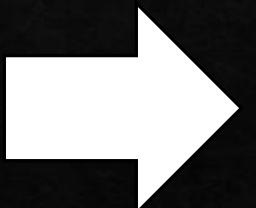
Ahmed likes mstry

Samar likes family

Sara likes family



As mohamed and Ahmed gives 5 starts to house MD



We can deduce that house MD is mstry series

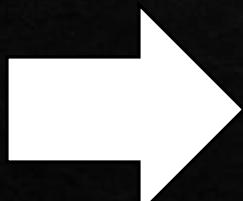
	Mohamed θ^1	Ahmed θ^2	Samar θ^3	Sara θ^4	X1 Funny	X2 Family	X3 Mystery
House MD	5	5	?	0	0	0	1
The office	?	5	3	0	?	?	?
The modern Family	4	?	5	5	0	1	0
Lucifer	0	0	0	?	?	?	?

Mohamed likes mstry

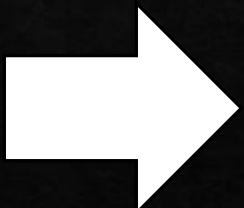
Ahmed likes mstry

Samar likes family

Sara likes family

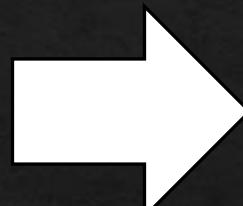


As Samar and sara rates 5 stars on modern family



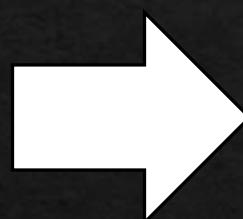
We can deduce that modern family is family series

Given the
features



Estimate the
parameters

Given the
parameters



Estimate the
Features

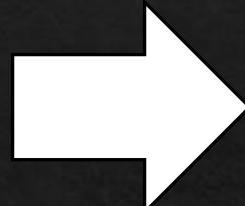
Given $x^{(1)}, \dots, x^{(n_m)}$, **estimate** $\theta^{(1)}, \dots, \theta^{(n_u)}$:

$$\min_{\theta^{(1)}, \dots, \theta^{(n_u)}} \frac{1}{2} \sum_{j=1}^{n_u} \sum_{i:r(i,j)=1} ((\theta^{(j)})^T x^{(i)} - y^{(i,j)})^2 + \frac{\lambda}{2} \sum_{j=1}^{n_u} \sum_{k=1}^n (\theta_k^{(j)})^2$$

Given $\theta^{(1)}, \dots, \theta^{(n_u)}$, **estimate** $x^{(1)}, \dots, x^{(n_m)}$:

$$\min_{x^{(1)}, \dots, x^{(n_m)}} \frac{1}{2} \sum_{i=1}^{n_m} \sum_{j:r(i,j)=1} ((\theta^{(j)})^T x^{(i)} - y^{(i,j)})^2 + \frac{\lambda}{2} \sum_{i=1}^{n_m} \sum_{k=1}^n (x_k^{(i)})^2$$

What if we
want the
both



We can go back and forth to solve the problem simultaneously

Given $x^{(1)}, \dots, x^{(n_m)}$, **estimate** $\theta^{(1)}, \dots, \theta^{(n_u)}$:

$$\min_{\theta^{(1)}, \dots, \theta^{(n_u)}} \frac{1}{2} \sum_{j=1}^{n_u} \sum_{i:r(i,j)=1} ((\theta^{(j)})^T x^{(i)} - y^{(i,j)})^2 + \frac{\lambda}{2} \sum_{j=1}^{n_u} \sum_{k=1}^n (\theta_k^{(j)})^2$$

Given $\theta^{(1)}, \dots, \theta^{(n_u)}$, **estimate** $x^{(1)}, \dots, x^{(n_m)}$:

$$\min_{x^{(1)}, \dots, x^{(n_m)}} \frac{1}{2} \sum_{i=1}^{n_m} \sum_{j:r(i,j)=1} ((\theta^{(j)})^T x^{(i)} - y^{(i,j)})^2 + \frac{\lambda}{2} \sum_{i=1}^{n_m} \sum_{k=1}^n (x_k^{(i)})^2$$

Content based

Collaborative Filtering