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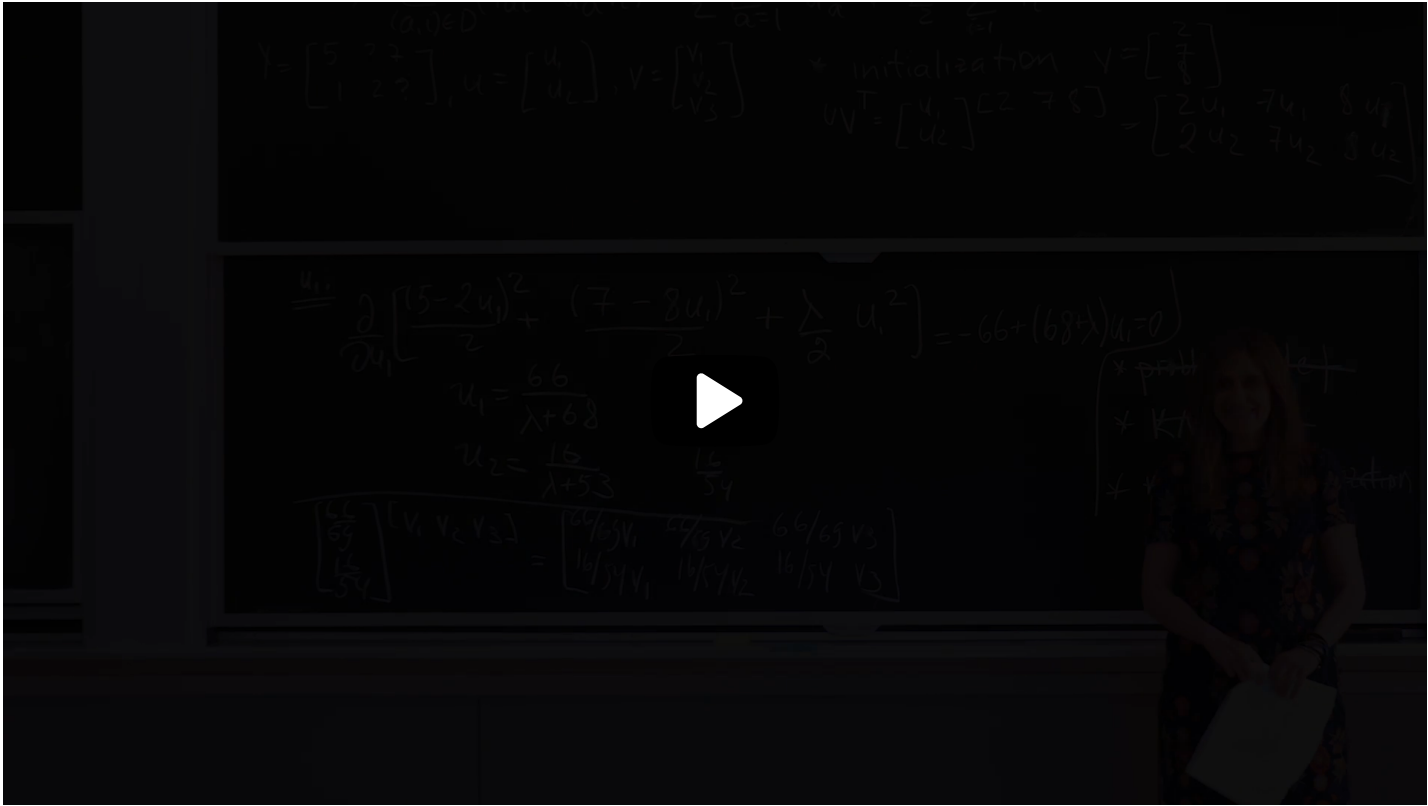


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# 6. Alternating Minimization

## Alternating Minimization



▶ 14:17 / 14:17

▶ Speed 1.50x

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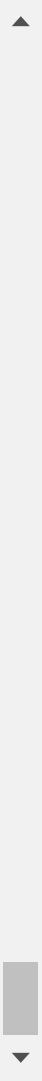
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But exactly the same idea will translate to rank 2, 3, and k.  
And you will try it in your exercises.  
But what I would like to tell you, that this very simple algorithm actually enables you, in a very interesting way, to find connection between different users and products.  
And given this very relatively simple machinery, you can actually solve a very non-trivial problem of product recommendation.  
**So with that, we completed the lecture.**



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## Alternating Minimization Concept Question

1/1 point (graded)

As in the video above, we now want to find  $U$  and  $V$  that minimize our new objective

$$J = \sum_{(a,i) \in D} \frac{(y_{ai} - [UV^T]_{ai})^2}{2} + \frac{\lambda}{2} \left( \sum_{a,k} U_{a,k}^2 + \sum_{i,k} V_{i,k}^2 \right).$$

To simplify the problem, we fix  $U$  and solve for  $V$ , then fix  $V$  to be the result from the previous step and solve for  $U$ , and repeat this alternate process until we find the solution.

When  $V$  is fixed, minimizing  $J$  becomes equivalent to minimizing ...

☐  $\frac{(Y_{ai} - u^{(a)}v^{(i)})^2}{2} + \frac{\lambda}{2} \|u^{(a)}\|^2$

☒  $\sum_{(a,i) \in D} \frac{(Y_{ai} - u^{(a)}v^{(i)})^2}{2} + \frac{\lambda}{2} \|u^{(a)}\|^2$  ✓

☐  $\sum_{(a,i) \in D} \frac{(Y_{ai} - u^{(a)}v^{(i)})^2}{2}$

☐  $\sum_{(a,i) \in D} \frac{(Y_{ai} - u^{(a)}v^{(i)})^2}{2} + \frac{\lambda}{2} \|v^{(i)}\|^2$

**Solution:**

Regarding terms related to only  $V$  as constants,  $J$  becomes equivalent to  $\sum_{(a,i) \in D} \frac{(Y_{ai} - u^{(a)}v^{(i)})^2}{2} + \frac{\lambda}{2} \|u^{(a)}\|^2$ .

You have used 1 of 3 attempts

**i** Answers are displayed within the problem

## Fixing $V$ and Finding $U$

2/2 points (graded)

Now, assume we have 2 users, 3 movies, and a 2 by 3 matrix  $Y$  given by

$$Y = \begin{bmatrix} 1 & 8 & ? \\ 2 & ? & 5 \end{bmatrix}$$

Our goal is to find  $U$  and  $V$  such that  $X = UV^T$  closely approximates the observed ratings in  $Y$ .

Assume we start by fixing  $V$  to initial values of  $[4, 2, 1]^T$ . Find the optimal  $2 \times 1$  vector  $U$  in this case. (Express your answer in terms of  $\lambda$ ).

First element of  $U$  is:

✓ Answer: 20/(20+lambda)

The second element of  $U$  is:

✓ Answer: 13/(17+lambda)

STANDARD NOTATION

### Solution:

To compute the first element ( $u_1$ ), compute the objective (ignore missing elements from  $Y$ ), derive and compare to zero to find the minimum:

$$\frac{\partial}{\partial u_1} \left[ \frac{(1 - 4u_1)^2}{2} + \frac{(8 - 2u_1)^2}{2} + \frac{\lambda}{2} u_1^2 \right] = (\lambda + 20) u_1 - 20 = 0.$$

You have used 2 of 3 attempts

**i** Answers are displayed within the problem

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