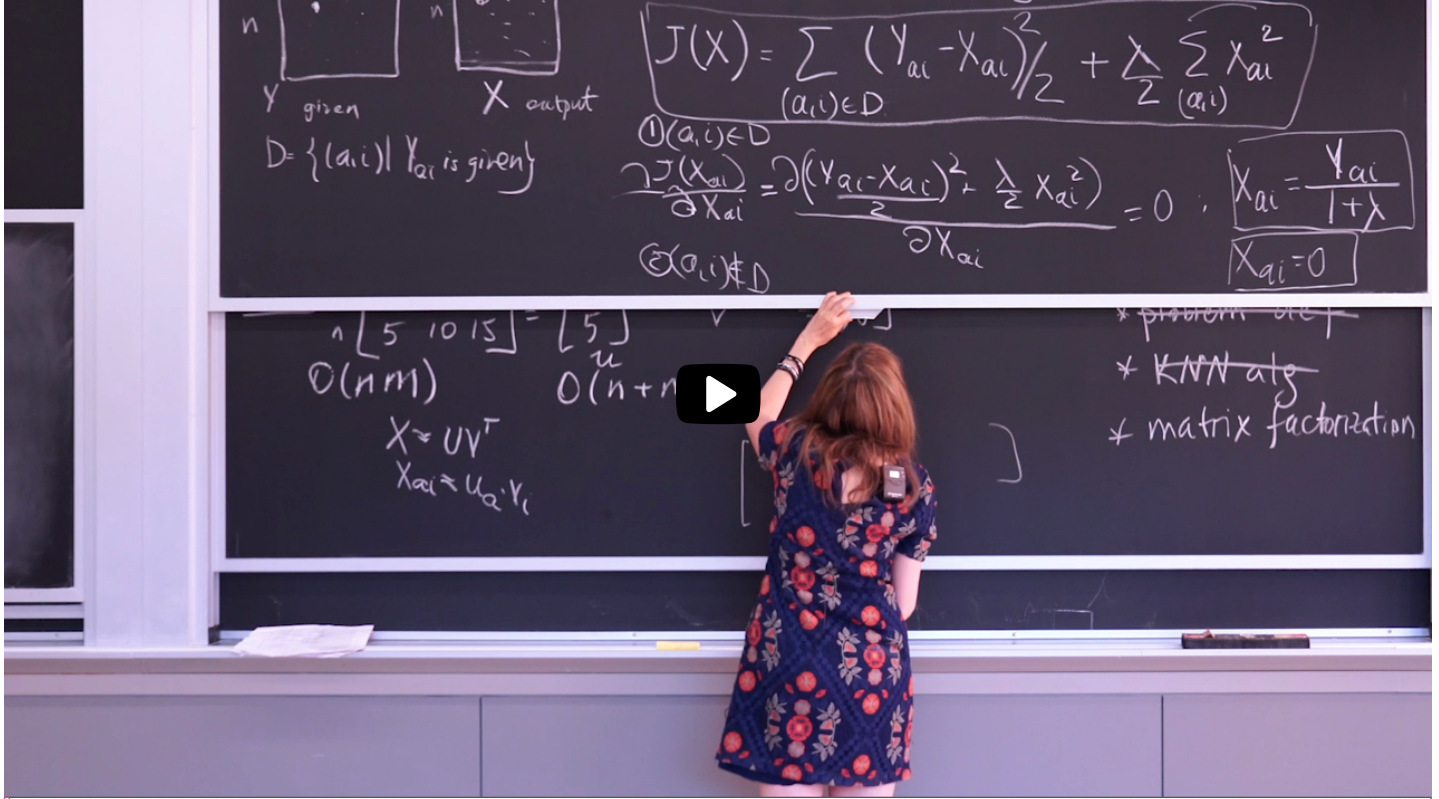


## 6. Alternating Minimization

### Alternating Minimization



▶ 0:00 / 0:00

▶ 1.50x 🔊 ⌂ CC 🔊

#### Video

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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

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

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.

Consider the case  $k = 1$ . The matrices  $U$  and  $V$  reduce to vectors  $u$  and  $v$ , such that  $u_a = U_{a1}$  and  $v_i = V_{i1}$ .

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

☐  $\frac{(Y_{ai} - u_a v_i)^2}{2} + \frac{\lambda}{2} \sum_a (u_a)^2$

☒  $\sum_{(a,i) \in D} \frac{(Y_{ai} - u_a v_i)^2}{2} + \frac{\lambda}{2} \sum_a (u_a)^2$

$$\sum_{(a,i) \in D} \frac{(Y_{ai} - u_a v_i)^2}{2}$$

$$\sum_{(a,i) \in D} \frac{(Y_{ai} - u_a v_i)^2}{2} + \frac{\lambda}{2} \sum_i (v_i)^2$$



### Solution:

Regarding terms containing only  $V$  as constants, minimizing  $J$  is equivalent to minimizing

$$\sum_{(a,i) \in D} \frac{(Y_{ai} - u_a v_i)^2}{2} + \frac{\lambda}{2} \sum_a (u_a)^2.$$

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You have used 1 of 3 attempts

**i** Answers are displayed within the problem

## Fixing V and Finding U

2.0/2 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:

20/(lambda+20)

✓ Answer: 20/(20+lambda)

The second element of  $U$  is:

13/(lambda+17)

✓ 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:

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

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You have used 2 of 3 attempts

**i** Answers are displayed within the problem

## Discussion

Hide Discussion

**Topic:** Unit 2 Nonlinear Classification, Linear regression, Collaborative Filtering (2 weeks); Lecture 7. Recommender Systems / 6. Alternating Minimization

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## Sage & Coffee - Alternating Minimization

question posted 8 months ago by [GuilhermeKinzel](#)

Tried some ways to see this algorithm working, so, for this one I used Sage (its came from Python) because of symbolic equations/derivatives. Using Sage Cell Server:

<https://sagecell.sagemath.org/>

I used the same dataset of the video, with:

$$Y = \begin{pmatrix} 5 & ? & 7 \\ 1 & 2 & ? \end{pmatrix}$$

$$u_1 = \begin{pmatrix} \frac{66}{69} & \frac{16}{54} \end{pmatrix}$$

$X$  will be updated every iteration, but at even iteration is calculated  $u \cdot v = X$  without random values. For instance:

- First iteration: update  $v$
- Second iteration: update  $u$
- Is calculated  $u \cdot v = X$ , the line "`X_Done = uAntigo.transpose()*vAntigo`"

```
SHOW EVERY STEP = 0
var('U_1, U_2, V_1, V_2, V_3')

Y = matrix(SR,[[5,0,7],[1,2,0]])

#v = matrix(SR,[[V_1, V_2, V_3]])
v = matrix(SR,[[2, 7, 8]]) ##Random vector

for i in range(100):

    u = matrix(SR,[[U_1, U_2]])
    X = u.transpose()*v
    #show(X) Importante

    H = 0 ##Initialize
    for i in range(X.nrows()):
        H = 0
        for j in range(X.ncols()):
            if(Y[i,j] != 0):
                H = H + (Y[i,j] - X[i,j])^2/2
        Eq = H + u[0,i]^2/2
        #show(Eq)
        Eqq = diff(Eq, u[0,i])
        SolveEq = Eqq.solve(u[0,i])
        u[0,i] = SolveEq[0].rhs().n()
    #u.n()
    uAntigo = u
    #show(u)

    ## Agora zera o v
    v = matrix(SR,[[V_1, V_2, V_3]]) ##Random vector
    X = u.transpose()*v
    #show(X) ##Importante

    H = 0 ##Initialize
    for j in range(X.ncols()):
        H = 0
        for i in range(X.nrows()):
            if(Y[i,j] != 0):
                H = H + (Y[i,j] - X[i,j])^2/2
        Eq = H + v[0,j]^2/2
        #show(Eq)
        Eqq = diff(Eq, v[0,j])
        #show(Eqq)
        SolveEq = Eqq.solve(v[0,j])
        v[0,j] = SolveEq[0].rhs().n()
    #v = v.n()
    vAntigo = v
    #show(v)

    X_Done = uAntigo.transpose()*vAntigo
    if SHOW EVERY STEP:
        show(X_Done)

print("Finished")
X_Done = uAntigo.transpose()*vAntigo
if SHOW EVERY STEP == 0:
    show(X_Done)
#X_Done
```



Well, after some iterations, only a far away decimals change between them (occurring convergence I guess). I took  $X_{100}$  just for instance.

$$X_{100} = \begin{pmatrix} 4.36412901058397 & 2.71245640601817 & 6.19908252148648 \\ 1.25122626729700 & 0.777680195951360 & 1.77732025456042 \end{pmatrix}.$$

Comparing with:

$$Y = \begin{pmatrix} 5 & ? & 7 \\ 1 & 2 & ? \end{pmatrix}$$

I guess the user 1, movie 2 ( $Y_{12}$ ) will be rated as "bad movie" (3 stars for him) if he watch. I guess; and the user 2 watch the movie 3 ( $Y_{23}$ ), he will probably rate with 2 stars.

If someone want to see all the steps, remove the comments of show(X) or show(u) or show(v). The teachers matrix will appears as the first one.

Remark: with no closed formula or solutions, I don't know if the algo is 100% right (I even don't know why  $X_{22}$  become smaller than  $X_{21}$ ). Disclaimer needed. ☺

This post is visible to everyone.

**dkonomis** (Staff)

8 months ago - marked as answer 8 months ago by **dkonomis** (Staff)

It looks 100% correct to me, thanks for sharing your solution. I get the same values for  $X_{100}$ .

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Preview

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