

CS 395T: Homework 3

Brady Zhou
brady.zhou@utexas.edu

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1 Low Rank Matrix Recovery (Alternating Minimization)

We are interested in minimizing the following

$$\operatorname{argmin}_{B,C} \sum_{i=1}^n \sum_{j=1}^n g_{ij} (a_{ij} - B_i^\top C_j)^2 + \frac{\mu}{2} (\|B\|_F^2 + \|C\|_F^2)$$

We can do this by taking partials with respect to row vectors B_k^\top and column vectors C_k .

$$\begin{aligned} \frac{\partial}{\partial B_k^\top} \left(\sum_{i=1}^n \sum_{j=1}^n g_{ij} (a_{ij} - B_i^\top C_j)^2 + \frac{\mu}{2} (\|B\|_F^2 + \|C\|_F^2) \right) &= -2 \sum_{j=1}^n g_{kj} (a_{kj} - B_k^\top C_j) C_j^\top + \mu B_k^\top \\ \frac{\partial}{\partial C_k} \left(\sum_{i=1}^n \sum_{j=1}^n g_{ij} (a_{ij} - B_i^\top C_j)^2 + \frac{\mu}{2} (\|B\|_F^2 + \|C\|_F^2) \right) &= -2 \sum_{i=1}^n g_{ik} (a_{ik} - B_i^\top C_k) B_i + \mu C_k \end{aligned}$$

Setting $\frac{\partial}{\partial B_k^\top} = 0$, we have the following

$$\begin{aligned} 0 &= -2 \sum_{j=1}^n g_{kj} (a_{kj} - B_k^\top C_j) C_j + \mu B_k^\top \\ 2 \sum_{j=1}^n g_{kj} a_{kj} C_j^\top &= B_k^\top \left(2 \sum_{j=1}^n g_{kj} C_j C_j^\top + \mu I \right) \\ 2 \sum_{j=1}^n g_{kj} a_{kj} C_j &= \left(2 \sum_{j=1}^n g_{kj} C_j C_j^\top + \mu I \right) B_k^{\top\top} \quad \text{Transpose for sanity.} \end{aligned}$$

So we can see the optimal B_k^\top is the solution to $Ax = b$, where

$$\begin{aligned} A &= 2 \sum_{j=1}^n g_{kj} C_j C_j^\top + \mu I \\ b &= 2 \sum_{j=1}^n g_{kj} a_{kj} C_j \end{aligned}$$

Similarly, for column vectors C_k ,

$$\begin{aligned} 0 &= -2 \sum_{i=1}^n g_{ik} (a_{ik} - B_i^\top C_k) B_i + \mu C_k \\ 2 \sum_{i=1}^n g_{ik} a_{ik} B_i &= \left(2 \sum_{i=1}^n g_{ik} B_i B_i^\top + \mu I \right) C_k \end{aligned}$$

And we have our result.

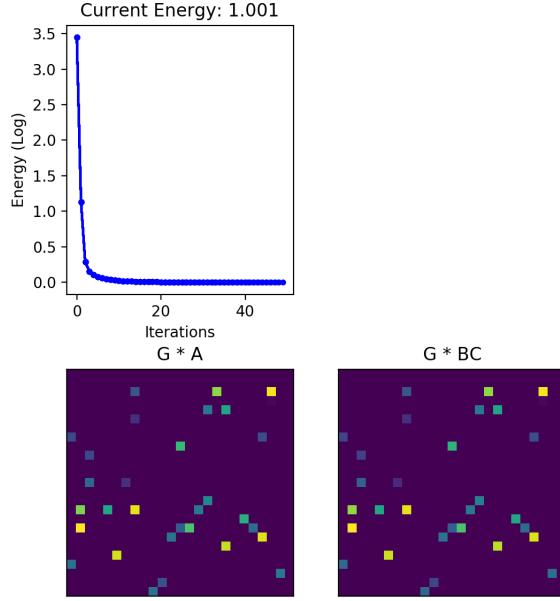


Figure 1: Results of alternating minimization.

2 Low Rank Matrix Recovery (Newton Trust Region)

To perform the Newton Trust Region method, we will have to calculate the Hessian of the objective function, where we are now minimizing over

$$x = \begin{bmatrix} B_1^{\top} \\ B_2^{\top} \\ \vdots \\ B_n^{\top} \\ \hline C_1 \\ C_2 \\ \vdots \\ C_n \end{bmatrix}$$

Restating the result from Problem 1,

$$\begin{aligned} \frac{\partial f}{\partial B_i^{\top}} &= -2 \sum_{k=1}^n g_{ik} (a_{ik} - B_i^{\top} C_k) C_k + \mu B_i^{\top} \\ \frac{\partial f}{\partial C_i} &= -2 \sum_{k=1}^n g_{ki} (a_{ki} - B_k^{\top} C_i) B_k^{\top} + \mu C_i \end{aligned}$$

By inspection, we can see $\frac{\partial f}{\partial B_i^{\top} \partial B_j^{\top}} = \frac{\partial f}{\partial C_i \partial C_j} = 0$ when $i \neq j$. For $i = j$, we have the following

$$\begin{aligned}\frac{\partial f}{\partial^2 B_i^{\top}} &= 2 \sum_{k=1}^n g_{ik} C_k C_k^{\top} + \mu I \\ \frac{\partial f}{\partial^2 C_i} &= 2 \sum_{k=1}^n g_{ki} B_k^{\top} B_k^{\top} + \mu I\end{aligned}$$

And now for the cross terms,

$$\begin{aligned}\frac{\partial f}{\partial B_i^{\top} \partial C_j} &= \frac{\partial}{\partial C_j} \left(-2 \sum_{k=1}^n g_{ik} (a_{ik} - B_i^{\top} C_k) C_k + \mu B_i^{\top} \right) \\ &= 2g_{ij} \left(-a_{ij} I + C_j^{\top} B_i^{\top} I + B_i^{\top} C_j^{\top} \right)\end{aligned}$$

And by symmetry of the Hessian, we have

$$\frac{\delta f}{\delta C_i \delta B_j^{\top}} = \left(\frac{\delta f}{\delta B_i^{\top} \delta C_j} \right)^{\top}$$

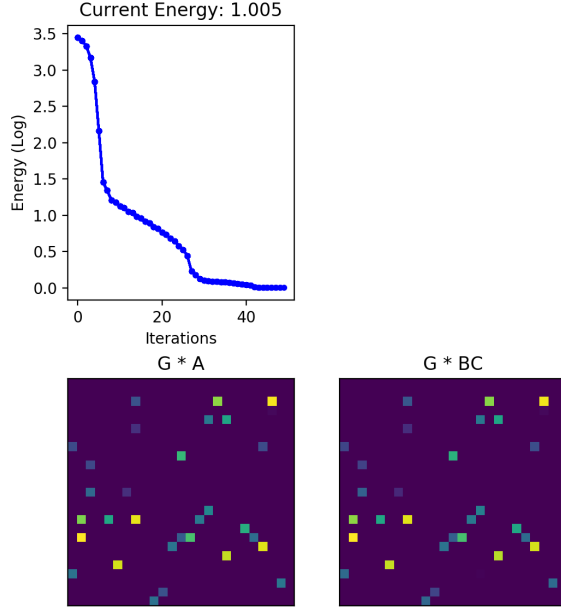


Figure 2: Results of trust region.

3 Trust Region Conditions with Arbitrary Norm

Let B be symmetric, A be positive semidefinite.

For p^* to be an optimal solution to the following problem

$$\begin{aligned}\underset{p}{\operatorname{argmin}} \quad & g^{\top} p + \frac{1}{2} p^{\top} B p \\ \text{subject to} \quad & \frac{1}{2} p^{\top} A p \leq \Delta^2\end{aligned}$$

we have three necessary and sufficient conditions.

1. $(B + \lambda A)p = -g$
2. $\exists \lambda \geq 0$ such that $\lambda(\frac{1}{2}p^\top Ap - \Delta^2) = 0$
3. $B + \lambda A$ is symmetric positive semidefinite

Lemma 3.1. *Let $\hat{m}(p)$ be the following*

$$\operatorname{argmin}_p p^\top g + \frac{1}{2}p^\top Bp$$

where B is positive semidefinite.

Then \hat{m} attains a minimum if and only if B is positive semi-definite and $g \in \operatorname{Range}(B)$.

Stated without proof (formal proof in lecture notes).

Proof. \Leftarrow

We want to show that if \bar{p} satisfies conditions 1, 2, 3, then $\bar{p} = p^*$.

Consider a similar problem.

$$\begin{aligned}\hat{m}(p) &= p^\top g + \frac{1}{2}p^\top (B + \lambda A)p \\ &= m(p) + \frac{1}{2}\lambda p^\top Ap\end{aligned}$$

By Lemma 3.1, we know that \bar{p} is the minimizer to \hat{m} , that is, $\hat{m}(\bar{p}) \leq \hat{m}(p) \forall p$.

$$\begin{aligned}\hat{m}(\hat{p}) &\leq \hat{m}(p) \\ m(\hat{p}) + \frac{\lambda}{2}\hat{p}^\top A\hat{p} &\leq m(p) + \frac{\lambda}{2}p^\top Ap \\ m(\hat{p}) &\leq m(p) + \frac{\lambda}{2}p^\top Ap - \frac{\lambda}{2}\hat{p}^\top A\hat{p} \\ &= m(p) + \frac{\lambda}{2}p^\top Ap - \frac{\lambda}{2}\hat{p}^\top A\hat{p} + \lambda\left(\frac{1}{2}\hat{p}^\top A\hat{p} - \Delta^2\right) \quad \text{By 2.} \\ &= m(p) + \lambda\left(p^\top Ap - \Delta^2\right) \\ &\leq m(p) \quad \text{Since } \frac{1}{2}p^\top Ap \leq \Delta^2 \text{ and } \lambda \geq 0.\end{aligned}$$

We can see that \bar{p} is the minimizer to m .

□

Proof. \Rightarrow

Consider the unconstrained formulation of m (Lagrangian) as

$$\operatorname{argmin}_{p, \lambda \geq 0} p^\top g + \frac{1}{2}p^\top Bp + \lambda\left(\frac{1}{2}p^\top Ap - \Delta^2\right)$$

p^* will lie at a stationary point of the Lagrangian. Taking the gradient with respect to p and setting it to 0, we have $(B + \lambda A)p = -g$. So we can see (1) is true.

Similarly, at a stationary point of the Lagrangian, either $\lambda = 0$, or $\frac{1}{2}p^\top Ap - \Delta^2 = 0$. Combining these two facts, we have (2).

Now we have to consider two cases.

Case 1: $\frac{1}{2}p^\top Ap < \Delta^2$, and $\lambda = 0$.

For this, we are on the interior and this turns into unconstrained optimization. We need to show $B + \lambda A$ is positive semi-definite. If p^* is a minimizer, then B is positive semi-definite, by Lemma 3.1, and $B + \lambda A = B$, since in this case $\lambda = 0$.

Case 2: $\frac{1}{2}p^\top Ap = \Delta^2$, and $\lambda > 0$.

$$\begin{aligned}
m(p^*) &\leq m(p) \\
g^\top p^* + \frac{1}{2}p^{*\top} B p^* &\leq g^\top p + \frac{1}{2}p^\top B p \\
g^\top (p^* - p) &\leq \frac{1}{2}p^\top B p - \frac{1}{2}p^{*\top} B p^* \\
-p^{*\top} (B + \lambda A)^\top (p^* - p) &\leq \frac{1}{2}p^\top B p - \frac{1}{2}p^{*\top} B p^* \\
(p - p^*)^\top (B + \lambda A) p^* &\leq \frac{1}{2}p^\top B p - \frac{1}{2}p^{*\top} B p^* \\
(p - p^*)^\top (B + \lambda A) p^* &\leq \frac{1}{2}p^\top B p - \frac{1}{2}p^{*\top} B p^* + \frac{\lambda}{2}p^\top A p - \frac{\lambda}{2}p^{*\top} A p^* \quad \text{Since } \frac{1}{2}p^\top A p = \frac{1}{2}p^{*\top} A p = \Delta^2 \\
(p - p^*)^\top (B + \lambda A) p^* &\leq \frac{1}{2}p^\top (B + \lambda A) p - \frac{1}{2}p^{*\top} (B + \lambda A) p^* \\
&\leq \frac{1}{2}p^\top (B + \lambda A) p - p^\top (B + \lambda A) p^* + \frac{1}{2}p^{*\top} (B + \lambda A) p^* \\
&\leq \frac{1}{2}(p - p^*)^\top (B + \lambda A) (p - p^*)
\end{aligned}$$

Finally, we must show that $p - p^*$ is dense, that is, $\forall u \in \mathbb{R}^n$, we can find $p - p^* = \alpha u$ such that $\frac{1}{2}p^\top A p = \Delta^2$. This is equivalent to showing there exists an α such that $\frac{1}{2}(\alpha u + p^*)^\top A (\alpha u + p^*) = \Delta^2$.

$$\begin{aligned}
\frac{1}{2}(\alpha u + p^*)^\top A (\alpha u + p^*) &= \Delta^2 \\
\frac{1}{2}\alpha^2 u^\top A u + \alpha u^\top A p^* + \frac{1}{2}p^{*\top} A p^* &= \Delta^2 \\
\frac{1}{2}\alpha^2 u^\top A u + \alpha u^\top A p^* &= 0 \\
\frac{1}{2}\alpha u^\top A u + u^\top A p^* &= 0
\end{aligned}$$

We need to show this term is 0 by picking α , and now we have three subcases, depending on properties of u .

Subcase 1: $u \in \text{Null}(A)$.

Clearly the term is 0, since $Au = 0$ by definition. So all values of α satisfy this.

Subcase 2: $u^\top A p^* \neq 0$.

We can simply solve $\alpha = \frac{-2u^\top A p^*}{u^\top A u}$.

Subcase 3: $u^\top A p^* = 0$.

We need to show that if this is the case, that $u^\top (B + \lambda A)u \geq 0$. Since the space is continuous, for any u , we can define a function $\hat{u}(\epsilon) = u + \epsilon p^*$.

From the two subcases above, we know $\lim_{\epsilon \rightarrow 0^-} \hat{u}(\epsilon)^\top (B + \lambda A) \hat{u}(\epsilon) = \lim_{\epsilon \rightarrow 0^+} \hat{u}(\epsilon)^\top (B + \lambda A) \hat{u}(\epsilon) \geq 0$. By continuity, we have our result that $u^\top (B + \lambda A)u \geq 0$.

□