Robust PCA via Weighted Low Rank

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1 Original Model

$$\min_{L,S} \|L\|_* + \lambda \|S\|_1, \quad s.t. \ W \circ F = W \circ L + S \tag{1}$$

2 Lagrangian function \mathcal{L} for the Original Model

$$\mathcal{L}(L, S, Y, \mu) = \|L\|_* + \lambda \|S\|_1 + \langle Y, W \circ (F - L) - S \rangle + \frac{\mu}{2} \|W \circ (F - L) - S\|_F^2$$
 (2)

3 Derivative Trace Hadamard Production

1. The trace is equivalent to the inner product:

$$X \cdot Y = tr(X^T Y) \tag{3}$$

where \cdot is inner product operator, $X, Y \in \mathbb{R}^{m \times n}$.

2. The Hadamard and inner products commute:

$$X \circ Y \cdot Z = X \cdot Y \circ Z \tag{4}$$

where \circ is Hadamard product operator, $X, Y \in \mathbb{R}^{m \times n}$.

Proof. It is easy to verify that Hadamard product operator is supper to inner product operator.

$$\begin{bmatrix} x_{11} & x_{12} & \cdots & x_{1n} \\ x_{21} & x_{22} & \cdots & x_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ x_{m-1,1} & x_{m-1,2} & \ddots & \vdots \\ x_{m1} & x_{m2} & \cdots & x_{mn} \end{bmatrix} \circ \begin{bmatrix} y_{11} & y_{12} & \cdots & y_{1n} \\ y_{21} & y_{22} & \cdots & y_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ y_{m-1,1} & y_{m-1,2} & \ddots & \vdots \\ y_{m-1,1} & y_{m-1,2} & \cdots & \vdots \\ y_{m1} & y_{m2} & \cdots & y_{mn} \end{bmatrix} \cdot \begin{bmatrix} z_{11} & z_{12} & \cdots & z_{1n} \\ z_{21} & z_{22} & \cdots & z_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ z_{m-1,1} & z_{m-1,2} & \ddots & \vdots \\ z_{m1} & z_{m2} & \cdots & z_{mn} \end{bmatrix}$$

$$= \begin{bmatrix} x_{11}y_{11} & x_{12}y_{12} & \cdots & x_{1n}y_{1n} \\ x_{21}y_{21} & x_{22}y_{22} & \cdots & x_{2n}y_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ x_{m-1,1}y_{m-1,1} & x_{m-1,2}y_{m-1,2} & \ddots & \vdots \\ x_{m1}y_{m1} & x_{m2}y_{m2} & \cdots & x_{mn}y_{mn} \end{bmatrix} \cdot \begin{bmatrix} z_{11} & z_{12} & \cdots & z_{1n} \\ z_{21} & z_{22} & \cdots & z_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ z_{m-1,1} & z_{m-1,2} & \ddots & \vdots \\ z_{m1} & z_{m2} & \cdots & z_{mn} \end{bmatrix}$$

$$= \sum_{j=1}^{n} \sum_{i=1}^{m} x_{ij} y_{ij} z_{ij}$$

$$\begin{bmatrix} z_{11} & z_{12} & \cdots & z_{1n} \\ z_{21} & z_{22} & \cdots & z_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ z_{m-1,1} & z_{m-1,2} & \cdots & \vdots \\ z_{m1} & z_{m2} & \cdots & z_{mn} \end{bmatrix} \cdot \begin{bmatrix} y_{11} & y_{12} & \cdots & y_{1n} \\ y_{21} & y_{22} & \cdots & y_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ y_{m-1,1} & y_{m-1,2} & \ddots & \vdots \\ y_{m1} & y_{m2} & \cdots & y_{mn} \end{bmatrix} \circ \begin{bmatrix} z_{11} & z_{12} & \cdots & z_{1n} \\ z_{21} & z_{22} & \cdots & z_{2n} \\ \vdots & \ddots & \vdots & \vdots \\ z_{m-1,1} & z_{m-1,2} & \ddots & \vdots \\ z_{m-1,1} & z_{m-1,2} & \cdots & \vdots \\ z_{m1} & z_{m2} & \cdots & z_{mn} \end{bmatrix}$$

$$= \sum_{n=1}^{\infty} \sum_{j=1}^{m} x_{ij} y_{ij} z_{ij}$$

$$: r = A \cdot C \circ B = A \circ C \cdot B \tag{5}$$

where r is a function which dependents on A, B, C and $r \in \mathbb{R}$.

$$\therefore dr = A \circ C \cdot dB \Rightarrow \frac{\partial r}{\partial B} = A \circ C \tag{6}$$

4 Derivative $\frac{\partial (A \cdot A \circ B)}{\partial A}$

$$A \cdot A \circ B = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ a_{m-1,1} & a_{m-1,2} & \ddots & \ddots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \cdot \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ a_{m-1,1} & a_{m-1,2} & \ddots & \ddots \\ a_{m1} & a_{m2} & \cdots & a_{mn} \end{bmatrix} \circ \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ b_{21} & b_{22} & \cdots & b_{2n} \\ \vdots & \ddots & \ddots & \vdots \\ b_{m-1,1} & b_{m-1,2} & \ddots & \ddots \\ b_{m-1,1} & b_{m-1,2} & \cdots & \ddots \\ b_{m1} & b_{m2} & \cdots & b_{mn} \end{bmatrix}$$
$$= \sum_{j=1}^{n} \sum_{i=1}^{m} a_{ij}^{2} b_{ij}$$

so, we can get that

$$\frac{\partial A \cdot A \circ B}{\partial A} = 2A \circ B \tag{7}$$

5 Computing for the Original Model

1. update
$$L$$

$$L^* = \arg\min_{r} \|L\|_* - \langle Y, W \circ L \rangle + \frac{\mu}{2} \|W \circ (F - L) - S\|_F^2$$
 (8)

We denote that

$$q = -\langle Y, W \circ L \rangle + \frac{\mu}{2} \| W \circ (F - L) - S \|_F^2$$

$$= -Y \cdot W \circ L + \frac{\mu}{2} \left[\langle W \circ (F - L) - S, W \circ (F - L) - S \rangle \right]$$

$$= -Y \cdot W \circ L + \frac{\mu}{2} \langle W \circ F, W \circ F - W \circ L - S \rangle - \frac{\mu}{2} \langle W \circ L, W \circ F - W \circ L - S \rangle$$

$$- \frac{\mu}{2} \langle S, W \circ F - W \circ L - S \rangle$$

$$= -Y \cdot W \circ L + \frac{\mu}{2} (W \circ F \cdot W \circ F) - \frac{\mu}{2} (W \circ F \cdot W \circ L) - \frac{\mu}{2} (W \circ F \cdot S) - \frac{\mu}{2} (W \circ L \cdot W \circ F)$$

$$+ \frac{\mu}{2} (W \circ L \cdot W \circ L) + \frac{\mu}{2} (W \circ L \cdot S) - \frac{\mu}{2} (S \cdot W \circ F) + \frac{\mu}{2} (S \cdot W \circ L) + \frac{\mu}{2} (S \cdot S)$$

$$(9)$$

then we can get that

$$\frac{\partial q}{\partial L} = \frac{\partial \left(-Y \cdot W \circ L - \frac{\mu}{2} \left(W \circ F \cdot W \circ L\right) - \frac{\mu}{2} \left(W \circ L \cdot W \circ F\right) + \frac{\mu}{2} \left(W \circ L \cdot W \circ L\right) + \mu \left(W \circ L \cdot S\right)\right)}{\partial L}$$

And

$$\begin{split} &-Y\cdot W\circ L-\frac{\mu}{2}\left(W\circ F\cdot W\circ L\right)-\frac{\mu}{2}\left(W\circ L\cdot W\circ F\right)+\frac{\mu}{2}\left(W\circ L\cdot W\circ L\right)+\mu\left(W\circ L\cdot S\right)\\ &=-Y\circ W\cdot L-\frac{\mu}{2}W\circ F\circ W\cdot L-\frac{\mu}{2}W\circ W\circ F\cdot L+\frac{\mu}{2}L\cdot L\circ W\circ W+\mu W\circ S\cdot L \end{split}$$

so, we can get that

$$\frac{\partial q}{\partial L} = -Y \circ W - \frac{\mu}{2} \left(W \circ F \circ W \right) - \frac{\mu}{2} \left(W \circ W \circ F \right) + \mu \left(L \circ W \circ W \right) + \mu W \circ S \tag{10}$$

According to LADMM, minimizing Equation 8 can be replaced by solving the following:

$$L^{k+1} = \underset{L}{\operatorname{arg\,min}} \|L\|_{*} + \langle \nabla_{L}q(L_{k}), L - L_{k} \rangle + \frac{\eta}{2} \|L - L_{k}\|_{F}^{2}$$

$$= \underset{L}{\operatorname{arg\,min}} \|L\|_{*} + \frac{\eta}{2} \|L - L_{k} + \frac{\nabla_{L}q(L_{k})}{\eta} \|$$

$$= \underset{L}{\operatorname{arg\,min}} \frac{1}{\eta} \|L\|_{*} + \frac{1}{2} \|L - \left(L_{k} - \frac{\nabla_{L}q(L_{k})}{\eta}\right) \|$$
(11)

2. update S

$$S^{k+1} = \underset{S}{\operatorname{arg\,min}} \lambda \|S\|_{1} + \langle Y, W \circ (F - L) - S \rangle + \frac{\mu}{2} \|W \circ (F - L) - S\|_{F}^{2}$$

$$= \underset{S}{\operatorname{arg\,min}} \lambda \|S\|_{1} + \frac{\mu}{2} \|W \circ (F - L) - S + \frac{Y}{\mu}\|_{F}^{2}$$

$$= \underset{S}{\operatorname{arg\,min}} \frac{\lambda}{\mu} \|S\|_{1} + \frac{1}{2} \|\left(W \circ (F - L) + \frac{Y}{\mu}\right) - S\|_{F}^{2}$$
(12)

3. update Y

$$Y^{k+1} = Y^k + \mu (W \circ (F - L) - S)$$
(13)

6 Our Model

$$\min_{L,S} \|L\|_{S_{2/3}}^{2/3} + \lambda \|S\|_{l_{2/3}}^{2/3}, \quad s.t. \quad W \circ F = W \circ L + S$$
(14)

7 Lagrangian function \mathcal{L} for Our Model

$$\mathcal{L}(L, S, Y, \mu) = \|L\|_{S_{2/3}}^{2/3} + \lambda \|S\|_{l_{2/3}}^{2/3} + \langle Y, W \circ F - W \circ L - S \rangle + \frac{\mu}{2} \|W \circ F - W \circ L - S\|_F^2 \quad (15)$$

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