

## 1. Unified function

$$\min_{\mathbf{U}^{(i)}, \mathbf{V}^{(i)}, \mathbf{S}^{(i)}, \mathbf{S}^*} \sum_{i=1}^m \left\| \mathbf{X}^{(i)} - \mathbf{U}^{(i)} \mathbf{V}^{(i)} \right\|_F^2 + \lambda_1 \left\| \mathbf{V}^{(i)} - \mathbf{V}^{(i)} \mathbf{S}^{(i)} \right\|_F^2 - \lambda_2 \text{Tr}(\mathbf{F} \mathbf{F}^\top \sum_{i=1}^m \mathbf{W}^{(i)\top} \mathbf{S}^{(i)} \mathbf{W}^{(i)}) \quad (1)$$

$$\text{s.t. } \mathbf{V}^{(i)} \geq 0, 0 \leq S_{j,k}^{(i)} \leq 1, \mathbf{S}^{(i)\top} \mathbf{1} = \mathbf{1}, \mathbf{F} \mathbf{F}^\top = \mathbf{I}$$

## 2. Optimization

### 2.1. Update $\mathbf{U}^{(i)}$

With  $\mathbf{V}^{(i)}$ ,  $\mathbf{Z}^{(i)}$ ,  $\alpha_i$  and  $\mathbf{Z}^*$  fixed, for each  $\mathbf{U}^{(i)}$ , we need to solve the following problem in Eq.(2)

$$\min_{\mathbf{U}^{(i)}} \left\| \mathbf{X}^{(i)} - \mathbf{U}^{(i)} \mathbf{V}^{(i)} \right\|_F^2 \quad (2)$$

Each of  $\mathbf{U}^{(i)}$  can be solved separately since views are independent from each other. Therefore the objective function that we minimize can be rewritten as

$$\mathcal{L}(\mathbf{U}^{(i)}) = \min_{\mathbf{U}^{(i)}} \text{Tr}(\mathbf{X}^{(i)\top} \mathbf{X}^{(i)} - 2\mathbf{V}^{(i)\top} \mathbf{U}^{(i)\top} \mathbf{X}^{(i)} + \mathbf{V}^{(i)\top} \mathbf{U}^{(i)\top} \mathbf{U}^{(i)} \mathbf{V}^{(i)}). \quad (3)$$

The closed-form solution of Eq.(3) can be easily obtained by setting  $\partial \mathcal{L}(\mathbf{U}^{(i)}) / \partial \mathbf{U}^{(i)} = -2\mathbf{X}^{(i)} \mathbf{V}^{(i)\top} + 2\mathbf{U}^{(i)} \mathbf{V}^{(i)} \mathbf{V}^{(i)\top}$  to zero. Then we can get the optimal solution

$$\mathbf{U}^{(i)} = \mathbf{X}^{(i)} \mathbf{V}^{(i)\top} (\mathbf{V}^{(i)} \mathbf{V}^{(i)\top})^{-1} \quad (4)$$

### 2.2. Update $\mathbf{V}^{(i)}$

Fixing  $\mathbf{U}^{(i)}$ ,  $\mathbf{Z}^{(i)}$ ,  $\alpha_i$  and  $\mathbf{Z}^*$ , the minimum problem for optimizing  $\mathbf{V}^{(i)}$  is reduced into the following problem,

$$\mathcal{L}(\mathbf{V}^{(i)}) = \min_{\mathbf{V}^{(i)}} \sum_{i=1}^m \left\| \mathbf{X}^{(i)} - \mathbf{U}^{(i)} \mathbf{V}^{(i)} \right\|_F^2 + \lambda_1 \left\| \mathbf{V}^{(i)} - \mathbf{V}^{(i)} \mathbf{S}^{(i)} \right\|_F^2 \quad \text{s.t. } \mathbf{V}^{(i)} \geq 0. \quad (5)$$

The partial derivation of  $\mathcal{L}(\mathbf{V}^{(i)})$  with respect to  $\mathbf{V}^{(i)}$  is

$$\frac{\partial \mathcal{L}(\mathbf{V}^{(i)})}{\partial \mathbf{V}^{(i)}} = \mathbf{V}^{(i)} \mathbf{P}^{(i)} + \mathbf{U}^{(i)\top} \mathbf{U}^{(i)} \mathbf{V}^{(i)} - \mathbf{U}^{(i)\top} \mathbf{X}^{(i)} + \Gamma, \quad (6)$$

where  $\Gamma$  is the Lagrange Multiplier and  $\mathbf{P}^{(i)} = \lambda_1(\mathbf{I} - \mathbf{S}^{(i)} - \mathbf{S}^{(i)\top} + \mathbf{S}^{(i)}\mathbf{S}^{(i)\top})$ . According to the optimization of semi-NMF [?] and the KKT condition, we can get

$$(\mathbf{V}^{(i)}\mathbf{P}^{(i)} + \mathbf{U}^{(i)\top}\mathbf{U}^{(i)}\mathbf{V}^{(i)} - \mathbf{U}^{(i)\top}\mathbf{X}^{(i)})_{jk}\mathbf{V}_{jk}^{(i)} = 0. \quad (7)$$

Denoting the positive element  $M_{jk}^+$  in matrix  $\mathbf{M}$  as  $M_{jk}^+ = (|M_{jk}| + M_{jk})/2$  and the absolute value of negative element  $M_{jk}^-$  as  $M_{jk}^- = (|M_{jk}| - M_{jk})/2$ , we can get  $M_{jk} = M_{jk}^+ - M_{jk}^-$ . Similarly,  $(\mathbf{U}^{(i)\top}\mathbf{U}^{(i)})_{jk} = (\mathbf{U}^{(i)\top}\mathbf{U}^{(i)})_{jk}^+ - (\mathbf{U}^{(i)\top}\mathbf{U}^{(i)})_{jk}^-$ . Based on this, we can write the updating rule for  $\mathbf{V}^{(i)}$ :

$$V_{jk}^{(i)} \leftarrow V_{jk}^{(i)} \sqrt{\frac{[\mathbf{V}^{(i)}(\mathbf{P}^{(i)})^-]_{jk} + [(\mathbf{U}^{(i)\top}\mathbf{U}^{(i)})^- \mathbf{V}^{(i)}]_{jk} + (\mathbf{U}^{(i)\top}\mathbf{X}^{(i)})_{jk}^+}{[\mathbf{V}^{(i)}(\mathbf{P}^{(i)})^+]_{jk} + [(\mathbf{U}^{(i)\top}\mathbf{U}^{(i)})^+ \mathbf{V}^{(i)}]_{jk} + (\mathbf{U}^{(i)\top}\mathbf{X}^{(i)})_{jk}^-}} \quad (8)$$

### 2.3. Update $\mathbf{S}^{(i)}$

When  $\mathbf{U}^{(i)}$ ,  $\mathbf{V}^{(i)}$ ,  $\alpha_i$  and  $\mathbf{S}^*$  are fixed, the optimization for  $\mathbf{S}^{(i)}$  can be simplified as

$$\begin{aligned} \min_{\mathbf{S}^{(i)}} \sum_{i=1}^m \lambda_1 \left\| \mathbf{V}^{(i)} - \mathbf{V}^{(i)}\mathbf{S}^{(i)} \right\|_F^2 - \lambda_2 \text{Tr}(\mathbf{W}^{(i)\top}\mathbf{S}^{(i)\top}\mathbf{W}^{(i)}\mathbf{F}\mathbf{F}^\top) \\ \text{s.t. } 0 \leq S_{j,k}^{(i)} \leq 1, \mathbf{S}^{(i)\top}\mathbf{1} = \mathbf{1}. \end{aligned} \quad (9)$$

The closed-form solution of optimizing Eq.(9) can be attained by setting its derivation w.r.t.  $S_{:,j}^{(i)}$  to zero

$$\mathbf{S}^{(i)}_{j,:} = (2\lambda_1 \mathbf{V}^{(i)\top}\mathbf{V}^{(i)})^{-1} (2\lambda_1 \mathbf{V}^{(i)\top}\mathbf{V}_{:,j}^{(i)} + \lambda_2 \mathbf{W}^{(i)}\mathbf{F}\mathbf{F}^\top\mathbf{W}_{j,:}^{(i)\top}) \quad (10)$$

### 2.4. Update $\mathbf{F}$

With  $\mathbf{U}^{(i)}$ ,  $\mathbf{V}^{(i)}$  and  $\mathbf{S}^{(i)}$  fixed, we need to minimize the following objective for  $\mathbf{F}$ .

$$\begin{aligned} \min_{\mathbf{F}} -\text{Tr}(\mathbf{F}\mathbf{F}^\top \sum_{i=1}^m \mathbf{W}^{(i)\top}\mathbf{S}^{(i)}\mathbf{W}^{(i)}) \\ \text{s.t. } \mathbf{F}^\top\mathbf{F} = \mathbf{I} \end{aligned} \quad (11)$$

Using  $\mathbf{M}$  to represent  $\sum_{i=1}^m \mathbf{W}^{(i)\top}\mathbf{S}^{(i)}\mathbf{W}^{(i)}$  in Eq.(11), we can transform the objective into the following form

$$\begin{aligned} \max_{\mathbf{F}} \text{Tr}(\mathbf{F}^\top\mathbf{M}\mathbf{F}) \\ \text{s.t. } \mathbf{F}^\top\mathbf{F} = \mathbf{I}. \end{aligned} \quad (12)$$

The optimal solution for  $\mathbf{F}$  can be attached by taking the eigenvectors corresponding to the largest  $k$  eigenvalues of  $\mathbf{M}$ .

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