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## Algorithm

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**Input:**

$P$  data matrices  $\mathbf{X}^1, \mathbf{X}^2, \dots, \mathbf{X}^P$ , parameters  $\beta, \omega$

**Output:**

$P$  basis matrices  $\mathbf{U}^1, \mathbf{U}^2, \dots, \mathbf{U}^P$ ,  $P$  relation matrices  $\mathbf{W}^1, \mathbf{W}^2, \dots, \mathbf{W}^P$ , factor matrices  $\mathbf{V}$ , weight vector  $\Pi(\pi^1, \dots, \pi^P)$

1: **Begin**

2: Initialize  $\mathbf{U}^1, \mathbf{U}^2, \dots, \mathbf{U}^P, \mathbf{W}^1, \mathbf{W}^2, \dots, \mathbf{W}^P, \mathbf{V}$

3: Initialize  $(\pi^1, \dots, \pi^P) = \left(\frac{1}{P}, \frac{1}{P}, \dots, \frac{1}{P}\right)$

4: **loop**

5:   **for**  $p=1$  to  $P$  **do**

6:     Fix  $\mathbf{V}$ , update  $\mathbf{U}^p, \mathbf{W}^p$

7:   **end for**

8:   Fix  $\mathbf{U}^1, \mathbf{U}^2, \dots, \mathbf{U}^P$ , update  $\mathbf{V}_l$

9:   Fix  $\mathbf{U}^1, \mathbf{U}^2, \dots, \mathbf{U}^P$ , update  $\mathbf{V}_{ul}$

10:   **for**  $p=1$  to  $P$  **do**

11:     Fix  $\mathbf{U}, \mathbf{W}, \mathbf{V}$ , compute  $c^p = \|\mathbf{X}^p - \mathbf{U}^p \mathbf{W}^p \mathbf{V}\|_F^2$

12:   **end for**

13:   **Update**  $\Pi$

14: **break** loop if convergence

15: **End**

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$$\min_{\{\mathbf{U}^{(p)}\}_{p=1}^P, \{\mathbf{W}^{(p)}\}_{p=1}^P, \mathbf{V}} \sum_{p=1}^P \pi^{(p)} \left\| \mathbf{X}^{(p)} - \mathbf{U}^{(p)} \mathbf{W}^{(p)} \mathbf{V} \right\|_F^2 + \beta \left\{ \text{tr} \left[ \mathbf{V}_l \mathbf{L}^a (\mathbf{V}_l)^T \right] - \text{tr} \left[ \mathbf{V}_l \mathbf{L}^p (\mathbf{V}_l)^T \right] \right\} + \omega \|\Pi\|^2$$

$$s.t. \quad \forall p, U_{ij}^{(p)} \geq 0, V_{ij} \geq 0, \pi^{(p)} \geq 0, \sum_{p=1}^P \pi^{(p)} = 1$$

**We define:**

$$O(\mathbf{U}, \mathbf{V}, \mathbf{W}, \Pi) = \sum_{p=1}^P \pi^{(p)} \left\| \mathbf{X}^{(p)} - \mathbf{U}^{(p)} \mathbf{W}^{(p)} \mathbf{V} \right\|_F^2 + \beta \left\{ \text{tr} \left[ \mathbf{V}_l \mathbf{L}^a (\mathbf{V}_l)^T \right] - \text{tr} \left[ \mathbf{V}_l \mathbf{L}^p (\mathbf{V}_l)^T \right] \right\} + \omega \|\Pi\|^2$$

1. Fix  $\mathbf{V}, \mathbf{W}$ , update  $\mathbf{U}$

Lagrange function:

$$L(\mathbf{U}^{(p)}) = O(\mathbf{U}^{(p)}) + \text{tr}(\Psi^T \mathbf{U}^{(p)})$$

$$\frac{\partial L(\mathbf{U}^{(p)})}{\partial \mathbf{U}^{(p)}} = -2\mathbf{X}^{(p)} \mathbf{V}^T (\mathbf{W}^{(p)})^T + 2\mathbf{U}^{(p)} \mathbf{W}^{(p)} \mathbf{V} \mathbf{V}^T (\mathbf{W}^{(p)})^T + \Psi$$

Using KKT condition, we get:

$$U^{(P)} \leftarrow U^{(P)} \circ \frac{X^{(P)} V^T (W^{(P)})^T}{U^{(P)} W^{(P)} V V^T (W^{(P)})^T}$$

2. Fix U,V, update W

Lagrange function:

$$L(W^{(P)}) = O(W^{(P)}) + tr(\Phi^T W^{(P)})$$

$$\frac{\partial L(W^{(P)})}{\partial W^{(P)}} = -2(U^{(P)})^T (X - U^{(P)} W^{(P)} V) V^T + \Phi$$

Using KKT condition, we get:

$$W^{(P)} \leftarrow W^{(P)} \circ \frac{(U^{(P)})^T X^{(P)} V^T}{(U^{(P)})^T U^{(P)} W^{(P)} V V^T}$$

3. Fix U,W, update  $V_l$

$$L(V) = O(V) + tr(\Gamma^T V)$$

$$\frac{\partial L}{\partial V_l} = \sum_{p=1}^P \pi^{(p)} \left( -2(W^{(p)})^T (U^{(p)})^T X_l^{(p)} + 2(W^{(p)})^T (U^{(p)})^T U^{(p)} W^{(p)} V_l \right) + \beta(2V_l L^a - 2V_l L^p)_l + \Gamma_l$$

$$L^a = D^a - W^a, L^p = D^p - W^p$$

Using KKT condition,

$$V_l \leftarrow V_l \circ \frac{\sum_{p=1}^P \pi^{(p)} \left( (W^{(p)})^T (U^{(p)})^T X_l^{(p)} \right) + \beta V_l (D^p + W^a)}{\sum_{p=1}^P \pi^{(p)} (W^{(p)})^T (U^{(p)})^T U^{(p)} W^{(p)} V_l + \beta V_l (D^a + W^p)}$$

update  $V_{ul}$

$$\frac{\partial L}{\partial V_{ul}} = \sum_{p=1}^P \pi^{(p)} \left( -2(W^{(p)})^T (U^{(p)})^T X_{ul}^{(p)} + 2(W^{(p)})^T (U^{(p)})^T U^{(p)} W^{(p)} V_{ul} \right) + \Gamma_{ul}$$

Using KKT condition,

$$V_{ul} \leftarrow V_{ul} \circ \frac{\sum_{p=1}^P \pi^{(p)} \left( (W^{(p)})^T (U^{(p)})^T X_{ul}^{(p)} \right)}{\sum_{p=1}^P \pi^{(p)} (W^{(p)})^T (U^{(p)})^T U^{(p)} W^{(p)} V_{ul}}$$

4. Update  $\Pi$ , using optimization tools

When U,V,W are fixed, minimization of  $O(\Pi)$  is a convex optimization

$$\begin{aligned}
& \min \sum_{p=1}^P \pi^{(p)} \left\| X^{(p)} - U^{(p)} W^{(p)} V \right\|_F^2 + \lambda \|\Pi\|^2 \\
& s.t. \quad \pi^{(p)} > 0, \sum_{p=1}^P \pi^{(p)} = 1
\end{aligned}
\quad \rightarrow \quad
\begin{aligned}
& \min \sum_{p=1}^P \pi^{(p)} c^{(p)} + \omega \|\Pi\|^2 \\
& s.t. \quad \pi^{(p)} > 0, \sum_{p=1}^P \pi^{(p)} = 1
\end{aligned}$$

Graph Laplacian :

$$\frac{1}{2} \sum_i \sum_j W_{ij} \|v_i - v_j\|_2^2 = \text{tr}[VLV^T]$$

Prove:

$$\begin{aligned}
left &= \frac{1}{2} \sum_i \sum_j W_{ij} \sum_k \left[ v_{i(k)} - v_{j(k)} \right]^2 \\
&= \frac{1}{2} \sum_i \sum_j W_{ij} \sum_k \left[ v_{i(k)}^2 + v_{j(k)}^2 - 2v_{i(k)}v_{j(k)} \right] \\
&= \frac{1}{2} \sum_i \sum_j W_{ij} \left( v_i^T v_i - 2v_i^T v_j + v_j^T v_j \right) \\
&= \frac{1}{2} \times 2 \left( \sum_i v_i^T v_i D_{ii} - \sum_{i,j} v_i^T v_j W_{ij} \right) \\
&= \sum_i v_i^T v_i D_{ii} - \sum_{i,j} v_i^T v_j W_{ij} \\
&= \text{tr}(VDV^T) - \text{tr}(VWV^T) \\
&= \text{tr}(VLV^T) = right
\end{aligned}$$