

# Multiview Clustering

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## Abstract

## Index Terms

## I. MODEL

$$\sum_{v=1}^V \left\| X^{(v)} - V^{(v)} U^T \right\|_{2,1} + \alpha \sum_{v=1}^V (\beta_v)^r \text{tr}(U^T L^{(v)} U) \quad (1)$$

$$s.t. U^T U = I, U \geq 0$$

## II. OPTIMIZATION

The objective function in above is **not convex** in four variables **but is convex if we update the  $2V + 1$  variables alternatively**. Thus, we use Alternating Direction Method of Multiplier(**ADMM**) to optimize the objective function. by introducing four auxiliary variables  $E^{(v)} = X^{(v)} - V^{(v)} U^T$  and  $Z_1 = U$ . The objective function can be rewritten into the following equivalent problem:

$$\arg \min_{U, V^{(v)}, E^{(v)}, Z_1} \sum_{v=1}^V \left\| E^{(v)} \right\|_{2,1} + \alpha \sum_{v=1}^V (\beta_v)^r \text{tr}(Z_1^T L^{(v)} U) \quad (2)$$

$$s.t. E^{(v)} = X^{(v)} - V^{(v)} U^T, Z_1 = U, U^T U = I, Z_1 \geq 0$$

which can be solved by the following **ADMM problem**

$$\begin{aligned}
& \min_{U, V^{(v)}, E^{(v)}, Z_1, \lambda^{(v)}, \mu} \sum_{v=1}^V \|E^{(v)}\|_{2,1} + \alpha \sum_{v=1}^V (\beta_v)^r \text{tr}(Z_1^T L^{(v)} U) \\
& + \sum_{v=1}^V \langle \lambda^{(v)}, X^{(v)} - V^{(v)} U^T - E^{(v)} \rangle + \langle \lambda_1, Z_1 - U \rangle \\
& + \frac{\mu}{2} (\|Z_1 - U\|_F^2 + \sum_{v=1}^V \|X^{(v)} - V^{(v)} U^T - E^{(v)}\|_F^2) \\
& \text{s.t. } U^T U = I, Z_1 \geq 0
\end{aligned} \tag{3}$$

**Update  $E^{(v)}$ :** To update  $E^{(v)}$ , we fixed other variables except  $E^{(v)}$  and remove terms that are irrelevant to  $E^{(v)}$ . The the objective function becomes

$$\min_{E^{(v)}} \frac{1}{2} \left\| E^{(v)} - (X^{(v)} - V^{(v)} U^T + \frac{1}{\mu} \lambda^{(v)}) \right\|_F^2 + \frac{1}{\mu} \|E^{(v)}\|_{2,1} \tag{4}$$

Let  $B = X^{(v)} - V^{(v)} U^T + \frac{1}{\mu} \lambda^{(v)}$ , then  $E^{(v)}$  can be update as

$$e_{vi} = \begin{cases} (1 - \frac{1}{\mu \|b_i\|}) b_i, & \text{if } \|b_i\| \geq \frac{1}{\mu} \\ 0, & \text{otherwise} \end{cases} \tag{5}$$

**Update  $V^{(v)}$ :**

To update  $V^{(v)}$ , we fix other variables except  $V^{(v)}$ , we obtain the following objective function:

$$\min_{V^{(v)}} \frac{\mu}{2} \left\| X^{(v)} - V^{(v)} U^T - E^{(v)} + \frac{1}{\mu} \lambda^{(v)} \right\|_F^2 \tag{6}$$

Considering that  $U^T U = I$ , we can rewrite the above objective function:

$$\min_{V^{(v)}} \frac{1}{2} \left\| V^{(v)} - (X^{(v)} - E^{(v)} + \frac{1}{\mu} \lambda^{(v)}) U \right\|_F^2 \tag{7}$$

then  $V^{(v)} = (X^{(v)} - E^{(v)} + \frac{1}{\mu} \lambda^{(v)}) U$ .

**Update  $Z_1$ :**

$$\min_{Z_1 \geq 0} \frac{\mu}{2} \|Z_1 - U\|_F^2 + \langle \lambda_1, Z_1 - U \rangle + \alpha \sum_{v=1}^V (\beta_v)^r \text{tr}(Z_1^T L^{(v)} U) \tag{8}$$

we obtain

$$\min_{Z_1 \geq 0} \|Z_1 - K\|_F^2 \tag{9}$$

where  $K = (U - \frac{1}{\mu}\lambda_1 - \frac{\alpha}{\mu} \sum_{v=1}^V (\beta_v)^r L^{(v)} U)$ . The above object function can be further decomposed to element-wise optimization problem as

$$\min_{Z_{1ij} \geq 0} \|Z_{1ij} - K_{ij}\|_F^2 \quad (10)$$

Therefor, the optimal solution of above problems is

$$Z_{1ij} = \max(K_{ij}, 0) \quad (11)$$

**Update  $U$ :**

$$\begin{aligned} \min_{U^T U = I} & \langle \lambda_1, Z_1 - U \rangle + \sum_{v=1}^V \langle \lambda^{(v)}, X^{(v)} - V^{(v)} U^T - E^{(v)} \rangle \\ & + \frac{\mu}{2} (\|Z_1 - U\|_F^2 + \sum_{(v=1)}^V \|X^{(v)} - V^{(v)} U^T - E^{(v)}\|_F^2) + \alpha \sum_{v=1}^V (\beta_v)^r \text{tr}(Z_1^T L^{(v)} U) \end{aligned} \quad (12)$$

removing the irrelevant terms, we arrive at

$$\min_{U^T U = I} \frac{\mu}{2} \|U\|_F^2 - \mu \langle H, U \rangle \quad (13)$$

where

$$H = \frac{1}{\mu} \lambda_1 + Z_1 - \frac{\alpha}{\mu} \sum_{v=1}^V L Z_1 + \sum_{v=1}^V (X^{(v)} - E^{(v)} + \frac{1}{\mu} * \lambda^{(v)})^T V^{(v)} \quad (14)$$

Thus, we further obtain

$$\min_{U^T U = I} \|U - H\|_F^2 \quad (15)$$

Denote

$$L(U, \Lambda) = \|U - H\|_F^2 + \Lambda(UU^T - I) \quad (16)$$

we then obtain

$$U = N_u Q_u^T \quad (17)$$

where  $N_u$  and  $Q_u$  are the left and right singular vectors of the economic singular value decomposition of  $H$ . **PS: How to obtain the  $N_u$  and  $Q_u$ . The details refer to the pages 38 and 39 of my doctoral dissertation**

**Update  $\beta_v$ :** Denote  $p^{(v)} = \text{tr}(Z_1^T L^{(v)} U)$

$$\beta_v = (rp^{(v)})^{\frac{1}{1-r}} / \sum_{v=1} V(rp^{(v)})^{\frac{1}{1-r}} \quad (18)$$

**Update  $\lambda^{(v)}$ ,  $\lambda_1$ , and  $\mu$ :**

$$\lambda^{(v)} = \lambda^{(v)} + \mu(X^{(v)} - V^{(v)}U^T - E^{(v)}) \quad (19)$$

$$\lambda_1 = \lambda_1 + \mu(Z_1 - U) \quad (20)$$

$$\mu = \rho\mu \quad (21)$$