

Appendix of "Adaptive Consensus Clustering for Multiple K-means via Base Results Refining"

Appendix A: Derivation of Eq.(10)

We have

$$\begin{aligned} \min_{\mathbf{A}} \mathcal{J}_1 &= v \operatorname{tr} (\mathbf{A}^k \mathbf{X} \mathbf{X}^T \mathbf{A}^{kT}) - 2 \operatorname{tr} \left(\sum_{m=1}^v \mathbf{G}^{(m)} \mathbf{F}^{(m)} \mathbf{X}^T \mathbf{A}^k \right) \\ &\quad + \operatorname{tr}(\mathbf{\Lambda}_1^T \mathbf{A}) + \frac{\mu}{2} \|\mathbf{A} - (\mathbf{I} + \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})/2\|_F^2. \end{aligned} \quad (1)$$

Let $\mathcal{L}_1 = \operatorname{tr} (\mathbf{A}^k \mathbf{X} \mathbf{X}^T \mathbf{A}^{kT})$, $\mathcal{L}_2 = \operatorname{tr} (\sum_{m=1}^v \mathbf{G}^{(m)} \mathbf{F}^{(m)} \mathbf{X}^T \mathbf{A}^k)$, $\mathcal{L}_3 = \operatorname{tr}(\mathbf{\Lambda}_1^T \mathbf{A})$, and $\mathcal{L}_4 = \|\mathbf{A} - (\mathbf{I} + \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})/2\|_F^2$. Then we compute the partial derivative of \mathcal{L}_1 , \mathcal{L}_2 , \mathcal{L}_3 , and \mathcal{L}_4 w.r.t. \mathbf{A} respectively.

Defining an auxiliary variable $\mathbf{U} = \mathbf{A}^k$, according to the chain rule, we have

$$\begin{aligned} \frac{\partial \mathcal{L}_1}{\partial A_{ij}} &= \operatorname{tr} \left(\left(\frac{\partial \operatorname{tr}(\mathbf{U} \mathbf{X} \mathbf{X}^T \mathbf{U}^T)}{\partial \mathbf{U}} \right)^T \frac{\partial \mathbf{U}}{\partial A_{ij}} \right) \\ &= 2 \operatorname{tr} \left(\mathbf{X} \mathbf{X}^T \mathbf{U}^T \frac{\partial \mathbf{A}^k}{\partial A_{ij}} \right) \\ &= 2 \operatorname{tr} \left(\mathbf{X} \mathbf{X}^T \mathbf{U}^T \sum_{r=0}^{k-1} \mathbf{A}^r \mathbf{J}^{ij} \mathbf{A}^{k-1-r} \right) \\ &= 2 \sum_{r=0}^{k-1} \operatorname{tr} (\mathbf{A}^{k-1-r} \mathbf{X} \mathbf{X}^T (\mathbf{A}^k)^T \mathbf{A}^r \mathbf{J}^{ij}) \end{aligned} \quad (2)$$

where $\mathbf{J}^{ij} \in \mathbb{R}^{n \times n}$ is a single-entry matrix, whose the (i, j) -th element is 1 and other elements are all 0s. Then, it is easy to verify that

$$\frac{\partial \mathcal{L}_1}{\partial \mathbf{A}} = 2 \sum_{r=0}^{k-1} (\mathbf{A}^{k-1-r} \mathbf{X} \mathbf{X}^T (\mathbf{A}^k)^T \mathbf{A}^r)^T. \quad (3)$$

Denoting $\mathbf{B} = \sum_{m=1}^v \mathbf{X} \mathbf{F}^{(m)T} \mathbf{G}^{(m)T}$, we have

$$\frac{\partial \mathcal{L}_2}{\partial \mathbf{A}} = \frac{\partial \operatorname{tr}(\mathbf{B} \mathbf{A}^k)}{\partial \mathbf{A}} = \sum_{r=0}^{k-1} (\mathbf{A}^r \mathbf{B} \mathbf{A}^{k-r-1})^T. \quad (4)$$

Moreover, we have

$$\frac{\partial \mathcal{L}_3}{\partial \mathbf{A}} = \boldsymbol{\Lambda}_1. \quad (5)$$

and by denoting $\mathbf{C} = (\mathbf{I} + \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})/2$, we have

$$\frac{\partial \mathcal{L}_4}{\partial \mathbf{A}} = 2(\mathbf{A} - \mathbf{C}). \quad (6)$$

To sum up Eqs.(3), (4), (5), (6), we obtain

$$\begin{aligned} \frac{\partial \mathcal{J}_1}{\partial \mathbf{A}} &= 2v \sum_{r=0}^{k-1} (\mathbf{A}^{k-r-1} \mathbf{X} \mathbf{X}^T \mathbf{A}^{kT} \mathbf{A}^r)^T - 2 \sum_{r=0}^{k-1} (\mathbf{A}^r \mathbf{B} \mathbf{A}^{k-r-1})^T + \boldsymbol{\Lambda}_1 + \mu(\mathbf{A} - \mathbf{C}). \end{aligned} \quad (7)$$

Appendix B: Derivation of Eq.(12)

We have

$$\begin{aligned} \min_{\mathbf{W}} \mathcal{J}_2 &= \lambda \|\mathbf{W} - \mathbf{F}\|_F^2 - \frac{1}{2} \text{tr}(\boldsymbol{\Lambda}_1^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}}) \\ &\quad + \text{tr}(\boldsymbol{\Lambda}_2^T \mathbf{W}) + \frac{\mu}{8} \text{tr}(\mathbf{D}^{-\frac{1}{2}} \mathbf{W}^T \mathbf{D}^{-1} \mathbf{W} \mathbf{D}^{-\frac{1}{2}}) \\ &\quad - \frac{\mu}{2} \text{tr}(\mathbf{E}^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}}) + \frac{\mu}{2} \|\mathbf{W} - \mathbf{V}\|_F^2, \end{aligned} \quad (8)$$

Denote $\mathcal{I}_1 = \|\mathbf{W} - \mathbf{F}\|_F^2$, $\mathcal{I}_2 = \text{tr}(\boldsymbol{\Lambda}_1^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})$, $\mathcal{I}_3 = \text{tr}(\boldsymbol{\Lambda}_2^T \mathbf{W})$, $\mathcal{I}_4 = \text{tr}(\mathbf{D}^{-\frac{1}{2}} \mathbf{W}^T \mathbf{D}^{-1} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})$, $\mathcal{I}_5 = \text{tr}(\mathbf{E}^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}})$, and $\mathcal{I}_6 = \|\mathbf{W} - \mathbf{V}\|_F^2$.

We have

$$\frac{\partial \mathcal{I}_1}{\partial \mathbf{W}} = 2(\mathbf{W} - \mathbf{F}), \quad (9)$$

$$\frac{\partial \mathcal{I}_3}{\partial \mathbf{W}} = \boldsymbol{\Lambda}_2, \quad (10)$$

and

$$\frac{\partial \mathcal{I}_6}{\partial \mathbf{W}} = 2(\mathbf{W} - \mathbf{V}). \quad (11)$$

Now, we consider the derivative of \mathcal{I}_2 . Taking $D_{ii} = \sum_{r=1}^n W_{ir}$ into \mathcal{I}_2 , we have

$$\begin{aligned}\frac{\partial \mathcal{I}_2}{\partial W_{pq}} &= \frac{\partial \sum_{i,j=1}^n \Lambda_{1ij} W_{ij} \frac{1}{\sqrt{\sum_{r=1}^n W_{ir}}} \frac{1}{\sqrt{\sum_{r=1}^n W_{jr}}}}{\partial W_{pq}} \\ &= \frac{\Lambda_{1pq}}{\sqrt{\sum_{r=1}^n W_{pr}} \sqrt{\sum_{r=1}^n W_{qr}}} - \frac{1}{2} \sum_{j=1}^n \frac{\Lambda_{1pj} W_{pj}}{\left(\sqrt{\sum_{r=1}^n W_{pr}}\right)^3 \sqrt{\sum_{r=1}^n W_{jr}}} \\ &\quad - \frac{1}{2} \sum_{i=1}^n \frac{\Lambda_{1ip} W_{ip}}{\left(\sqrt{\sum_{r=1}^n W_{pr}}\right)^3 \sqrt{\sum_{r=1}^n W_{ir}}} \\ &= \frac{\Lambda_{1pq}}{\sqrt{\sum_{r=1}^n W_{pr}} \sqrt{\sum_{r=1}^n W_{qr}}} - \frac{1}{2} \frac{1}{\left(\sqrt{\sum_{r=1}^n W_{pr}}\right)^3} \sum_{i=1}^n \frac{\Lambda_{1pi} W_{pi} + \Lambda_{1ip} W_{ip}}{\sqrt{\sum_{r=1}^n W_{ir}}}\end{aligned}\tag{12}$$

Reformulating it to the matrix form, we have

$$\frac{\partial \mathcal{I}_2}{\partial \mathbf{W}} = \mathbf{D}^{-\frac{1}{2}} \mathbf{\Lambda}_1 \mathbf{D}^{-\frac{1}{2}} - \frac{1}{2} \text{diag} \left(\mathbf{D}^{-\frac{3}{2}} \left(\mathbf{\Lambda}_1^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} + \mathbf{\Lambda}_1 \mathbf{D}^{-\frac{1}{2}} \mathbf{W}^T \right) \right) \mathbf{1}^T.\tag{13}$$

Similarly, the derivative of \mathcal{I}_5 is

$$\frac{\partial \mathcal{I}_5}{\partial \mathbf{W}} = \mathbf{D}^{-\frac{1}{2}} \mathbf{E} \mathbf{D}^{-\frac{1}{2}} - \frac{1}{2} \text{diag} \left(\mathbf{D}^{-\frac{3}{2}} \left(\mathbf{E}^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} + \mathbf{E} \mathbf{D}^{-\frac{1}{2}} \mathbf{W}^T \right) \right) \mathbf{1}^T.\tag{14}$$

Now, we consider \mathcal{I}_4 . We have

$$\begin{aligned}\frac{\partial \mathcal{I}_4}{\partial W_{pq}} &= \frac{\partial \left(\sum_{i,j=1}^n \frac{1}{\sum_{r=1}^n W_{ir}} \frac{1}{\sum_{r=1}^n W_{jr}} W_{ji}^2 \right)}{\partial W_{pq}} \\ &= - \sum_{j=1}^n \frac{W_{jp}^2}{\sum_{r=1}^n W_{jr}} \frac{1}{(\sum_{r=1}^n W_{pr})^2} - \sum_{i=1}^n \frac{W_{pi}^2}{\sum_{r=1}^n W_{ir}} \frac{1}{(\sum_{r=1}^n W_{pr})^2} + \frac{2W_{pq}}{\sum_{r=1}^n W_{pr} \sum_{r=1}^n W_{qr}} \\ &= \frac{2W_{pq}}{\sum_{r=1}^n W_{pr} \sum_{r=1}^n W_{qr}} - \frac{1}{(\sum_{r=1}^n W_{pr})^2} \sum_{i=1}^n \frac{W_{ip}^2 + W_{pi}^2}{\sum_{r=1}^n W_{ir}}.\end{aligned}\tag{15}$$

Reformulating it to the matrix form, we have

$$\frac{\partial \mathcal{I}_4}{\partial \mathbf{W}} = 2\mathbf{D}^{-1} \mathbf{W} \mathbf{D}^{-1} - \text{diag} \left(\mathbf{D}^{-2} \left(\mathbf{W}^T \mathbf{D}^{-1} \mathbf{W} + \mathbf{W} \mathbf{D}^{-1} \mathbf{W}^T \right) \right) \mathbf{1}^T.\tag{16}$$

To sum up Eqs.(9), (13), (10), (16), (14) and (11), we obtain

$$\begin{aligned}\frac{\partial \mathcal{J}_2}{\partial \mathbf{W}} &= 2\lambda(\mathbf{W} - \mathbf{F}) - \frac{1}{2} \mathbf{D}^{-\frac{1}{2}} (\mathbf{\Lambda}_1 + \mu \mathbf{E}) \mathbf{D}^{-\frac{1}{2}} \\ &\quad + \frac{1}{4} \text{diag} \left(\mathbf{D}^{-\frac{3}{2}} \left((\mathbf{\Lambda}_1 + \mu \mathbf{E})^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} + (\mathbf{\Lambda}_1 + \mu \mathbf{E}) \mathbf{D}^{-\frac{1}{2}} \mathbf{W}^T \right) \right) \mathbf{1}^T \\ &\quad + \mathbf{\Lambda}_2 + \frac{\mu}{4} \mathbf{D}^{-1} \mathbf{W} \mathbf{D}^{-1} + \mu(\mathbf{W} - \mathbf{V}) \\ &\quad - \frac{\mu}{8} \text{diag} \left(\mathbf{D}^{-2} \left(\mathbf{W}^T \mathbf{D}^{-1} \mathbf{W} + \mathbf{W} \mathbf{D}^{-1} \mathbf{W}^T \right) \right) \mathbf{1}^T,\end{aligned}\tag{17}$$