

Supplement for “New Insights into Laplacian Similarity Search”

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Abstract

This is the supplement for our main paper “New Insights into Laplacian Similarity Search” [3]. Here, we show the proofs of all the theoretical arguments in the main paper.

Proof of Statements in Sec. 2.1: M is positive and symmetric, i.e., $\forall i, j, m_{ij} > 0$, and $m_{ij} = m_{ji}$. Regardless of Λ , m_{ii} is always the unique largest element in the i -th column and row of M .

Proof. (a) Since $L + \alpha\Lambda$ is symmetric, $M = (L + \alpha\Lambda)^{-1}$ is symmetric.

(b) Note that

$$\begin{aligned} M &= (L + \alpha\Lambda)^{-1} = (D + \alpha\Lambda - W)^{-1} \\ &= (I - (D + \alpha\Lambda)^{-1}W)^{-1}(D + \alpha\Lambda)^{-1} \\ &= \left(\sum_{k=0}^{\infty} [(D + \alpha\Lambda)^{-1}W]^k \right) (D + \alpha\Lambda)^{-1}, \end{aligned}$$

from which we can see that M is positive since the graph is connected.

(c) Now we show that m_{jj} is the unique largest in its column. Assume, to the contrary, there exists $i, j, i \neq j$, such that $m_{jj} \leq m_{ij}$. Denote $k = \arg \max_{i \neq j} m_{ij}$. Note that M is symmetric and $M > 0$. Let $B = (b_{ij}) := D + \alpha\Lambda - W$. Note that B is symmetric and strictly diagonally dominant, i.e., $\forall k, b_{kk} > \sum_{i \neq k} |b_{ki}|$. By $BM = I$, we have $0 = B(k, :)M(:, j) = \sum_i b_{ki}m_{ij} = b_{kk}m_{kj} + \sum_{i \neq k} b_{ki}m_{ij} \geq b_{kk}m_{kj} - (\sum_{i \neq k} |b_{ki}|)m_{kj} = (b_{kk} - \sum_{i \neq k} |b_{ki}|)m_{kj} > 0$, which contradicts the assumption. \square

Proof of Theorem 2.1:

$$\begin{aligned} M &= C + E, \text{ where } C = \frac{1}{\alpha \sum_i \lambda_i} \mathbf{1}\mathbf{1}^\top, \text{ and } E = \\ &\Lambda^{-\frac{1}{2}} \left(\sum_{i=2}^n \frac{1}{\gamma_i + \alpha} \mathbf{u}_i \mathbf{u}_i^\top \right) \Lambda^{-\frac{1}{2}}. \end{aligned}$$

Proof. By definition,

$$\begin{aligned} M &= (L + \alpha\Lambda)^{-1} \\ &= \Lambda^{-\frac{1}{2}} (\Lambda^{-\frac{1}{2}} L \Lambda^{-\frac{1}{2}} + \alpha I)^{-1} \Lambda^{-\frac{1}{2}} \\ &= \Lambda^{-\frac{1}{2}} \left(\sum_{i=1}^n (\gamma_i + \alpha) \mathbf{u}_i \mathbf{u}_i^\top \right)^{-1} \Lambda^{-\frac{1}{2}} \\ &= \Lambda^{-\frac{1}{2}} \left(\sum_{i=1}^n \frac{1}{\gamma_i + \alpha} \mathbf{u}_i \mathbf{u}_i^\top \right) \Lambda^{-\frac{1}{2}} \\ &= \frac{1}{\alpha \sum_i \lambda_i} \mathbf{1}\mathbf{1}^\top + \Lambda^{-\frac{1}{2}} \left(\sum_{i=2}^n \frac{1}{\gamma_i + \alpha} \mathbf{u}_i \mathbf{u}_i^\top \right) \Lambda^{-\frac{1}{2}}. \end{aligned}$$

\square

Proof of Corollary 2.2: $\lim_{\alpha \rightarrow 0} E = \Lambda^{-\frac{1}{2}} \bar{L}^\dagger \Lambda^{-\frac{1}{2}}$.

Proof. It follows from $\bar{L}^\dagger = \sum_{i=2}^n \frac{1}{\gamma_i} \mathbf{u}_i \mathbf{u}_i^\top$. \square

Proof of Statements in Sec. 2.1:

Ranking by $(h_{ij})_{i=1, \dots, n}$ is equivalent as ranking by the j -th column of $D^{-\frac{1}{2}} L_{sym}^\dagger D^{-\frac{1}{2}}$.

Proof. Let e_i denote the i -th unit vector in \mathbb{R}^n . The hitting time that a random walk from vertex i to hit vertex j can be computed by [1]:

$$\begin{aligned} H_{ij} &= d(\mathcal{V}) \left\langle \frac{1}{\sqrt{d_j}} e_j, L_{sym}^\dagger \left(\frac{1}{\sqrt{d_j}} e_j - \frac{1}{\sqrt{d_i}} e_i \right) \right\rangle \\ &= d(\mathcal{V}) \left(\frac{1}{d_j} e_j^\top L_{sym}^\dagger e_j - \frac{1}{\sqrt{d_i d_j}} e_i^\top L_{sym}^\dagger e_j \right). \end{aligned}$$

Thus given j , ranking by $(h_{ij})_{i=1, \dots, n}$ is determined by $-\frac{1}{\sqrt{d_i d_j}} e_i^\top L_{sym}^\dagger e_j$. Denote by $B = (b_{ij}) := D^{-\frac{1}{2}} L_{sym}^\dagger D^{-\frac{1}{2}}$. Then $b_{ij} = \frac{1}{\sqrt{d_i d_j}} e_i^\top L_{sym}^\dagger e_j$. This shows that ranking by $(h_{ij})_{i=1, \dots, n}$ in ascending order is the same as ranking by $(b_{ij})_{i=1, \dots, n}$ in descending order. Note that a smaller h_{ij} means vertices i and j are closer on the graph. \square

Proof of Lemma 3.1: (a) [2] $\mathcal{L}_f(\mathcal{S}_k) = \sum_{j \in \bar{\mathcal{S}}_k} a_{1j}$,
(b) $\lim_{\alpha \rightarrow 0} \mathcal{L}_f(\mathcal{S}_k) = \lambda(\bar{\mathcal{S}}_k)/\lambda(\mathcal{V})$, $1 \leq k \leq n$.

Proof. (a) Recall that \mathbf{f} is the first column of $M = (L + \alpha\Lambda)^{-1}$. We have $(L + \alpha\Lambda)\mathbf{f} = \mathbf{e}_1$, which can be written as:

$$\sum_{j \neq 1} w_{1j}(\mathbf{f}_1 - \mathbf{f}_j) = 1 - \alpha\lambda_1\mathbf{f}_1, \quad (1)$$

$$\sum_{j \neq i} w_{ij}(\mathbf{f}_i - \mathbf{f}_j) = -\alpha\lambda_i\mathbf{f}_i, \quad i \neq 1. \quad (2)$$

By Eq. (1) and Eq. (2), we have

$$\begin{aligned} \mathcal{L}_f(\mathcal{S}_k) &= \sum_{i=1}^k \sum_{j \neq i} w_{ij}(\mathbf{f}_i - \mathbf{f}_j) = 1 - \sum_{i=1}^k \alpha\lambda_i\mathbf{f}_i \\ &= 1 - \sum_{j \in \mathcal{S}_k} a_{1j} = \sum_{j \in \bar{\mathcal{S}}_k} a_{1j}. \end{aligned} \quad (3)$$

Note that in Eq. (3), since $A = (a_{ij}) = (L + \alpha\Lambda)^{-1}\alpha\Lambda$, $\alpha\lambda_i\mathbf{f}_i = a_{1i}$. We also use the fact that $\sum_j a_{1j} = 1$.

(b) By Theorem 2.1.,

$$\begin{aligned} A &= (L + \alpha\Lambda)^{-1}\alpha\Lambda \\ &= \frac{1}{\sum_i \lambda_i} \mathbf{1}\mathbf{1}^\top \Lambda + \alpha\Lambda^{-\frac{1}{2}} \left(\sum_{i=2}^n \frac{1}{\gamma_i + \alpha} \mathbf{u}_i \mathbf{u}_i^\top \right) \Lambda^{\frac{1}{2}}. \end{aligned}$$

Therefore, $\lim_{\alpha \rightarrow 0} A = \frac{1}{\sum_i \lambda_i} \mathbf{1}\mathbf{1}^\top \Lambda$. By Eq. (3), we have $\lim_{\alpha \rightarrow 0} \mathcal{L}_f(\mathcal{S}_k) = \lambda(\bar{\mathcal{S}}_k)/\lambda(\mathcal{V})$. \square

Proof of Theorem 3.4: $\mathcal{R}_f(\mathcal{S}_c) < 1/(c-1)$.

Proof. Since $\mathcal{L}_f(\mathcal{S}_k) = \sum_{j \in \bar{\mathcal{S}}_k} a_{1j}$ strictly decreases when k increases, $\forall k < c$, $\mathcal{L}_f(\mathcal{S}_c) < \mathcal{L}_f(\mathcal{S}_k)$. \square

Proof of Theorem 3.5:

(a) If $d_i = b$, $\forall i$, for some constant b , then $\lim_{\alpha \rightarrow 0} \mathcal{R}_i(\mathcal{S}_c) = \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$.

(b) Suppose for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{|\mathcal{S}_c \setminus \mathcal{S}_k|} > \frac{d(\bar{\mathcal{S}}_c)}{|\bar{\mathcal{S}}_c|}$. Then $\lim_{\alpha \rightarrow 0} \mathcal{R}_i(\mathcal{S}_c) > \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$.

(c) Suppose for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{|\mathcal{S}_c \setminus \mathcal{S}_k|} < \frac{d(\bar{\mathcal{S}}_c)}{|\bar{\mathcal{S}}_c|}$. Then $\lim_{\alpha \rightarrow 0} \mathcal{R}_i(\mathcal{S}_c) < \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$.

Proof. (a) It follows from $d(\bar{\mathcal{S}}_k) = b|\bar{\mathcal{S}}_k|$, for $1 \leq k \leq c$.

(b) Since for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\bar{\mathcal{S}}_c)} > \frac{|\mathcal{S}_c \setminus \mathcal{S}_k|}{|\bar{\mathcal{S}}_c|}$, we have $\frac{d(\bar{\mathcal{S}}_k)}{d(\bar{\mathcal{S}}_c)} = \frac{d(\mathcal{S}_c \setminus \mathcal{S}_k) + d(\bar{\mathcal{S}}_c)}{d(\bar{\mathcal{S}}_c)} > \frac{|\mathcal{S}_c \setminus \mathcal{S}_k| + |\bar{\mathcal{S}}_c|}{|\bar{\mathcal{S}}_c|} = \frac{|\mathcal{S}_k|}{|\bar{\mathcal{S}}_c|}$.

(c) The proof is similar to that of (b). \square

Proof of Lemma 3.6:

$$\lim_{d(\mathcal{S}_c)/d(\bar{\mathcal{S}}_c) \rightarrow 0} \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c) = \frac{1}{c-1}.$$

Proof. $\frac{1}{\lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)} = \frac{\sum_{k=1}^{c-1} d(\bar{\mathcal{S}}_k)}{d(\bar{\mathcal{S}}_c)} = \frac{\sum_{k=1}^{c-1} (d(\mathcal{S}_c \setminus \mathcal{S}_k) + d(\bar{\mathcal{S}}_c))}{d(\bar{\mathcal{S}}_c)} = c - 1 + \frac{\sum_{k=1}^{c-1} d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\bar{\mathcal{S}}_c)}$. As $\frac{d(\mathcal{S}_c)}{d(\bar{\mathcal{S}}_c)} \rightarrow 0$, we have $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\bar{\mathcal{S}}_c)} \rightarrow 0$ for $k < c$, which completes the proof. \square

Proof of Lemma 3.7: $\lim_{d(\mathcal{S}_c)/d(\bar{\mathcal{S}}_c) \rightarrow \infty} \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c) = 0$,

if $d_1 < td(\mathcal{S}_c)$ for a fixed scalar t , $0 < t < 1$.

Proof. $\frac{1}{\lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)} = \frac{\sum_{k=1}^{c-1} d(\bar{\mathcal{S}}_k)}{d(\bar{\mathcal{S}}_c)} \geq \frac{d(\bar{\mathcal{S}}_1)}{d(\bar{\mathcal{S}}_c)} \geq \frac{d(\bar{\mathcal{S}}_1) + d_1 - d_1}{d(\bar{\mathcal{S}}_c)} \geq \frac{d(\bar{\mathcal{S}}_c) - d_1}{d(\bar{\mathcal{S}}_c)} \geq \frac{(1-t)d(\bar{\mathcal{S}}_c)}{d(\bar{\mathcal{S}}_c)} \rightarrow \infty$, as $\frac{d(\bar{\mathcal{S}}_c)}{d(\bar{\mathcal{S}}_c)} \rightarrow \infty$. \square

Proof of Theorem 3.8: Suppose for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{|\mathcal{S}_c \setminus \mathcal{S}_k|} < \frac{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}}{|\mathcal{S}_c|}$. Then $\lim_{\alpha \rightarrow 0} \mathcal{R}_i(\mathcal{S}_c) < \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$.

Proof. Since for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} < \frac{|\mathcal{S}_c \setminus \mathcal{S}_k|}{|\mathcal{S}_c|}$, we have $\frac{d(\mathcal{S}' \setminus \mathcal{S}_k) + \tau \hat{d}}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} = \frac{d(\mathcal{S}_c \setminus \mathcal{S}_k) + d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} < \frac{|\mathcal{S}_c \setminus \mathcal{S}_k| + |\mathcal{S}_c|}{|\mathcal{S}_c|} = \frac{|\mathcal{S}_k|}{|\mathcal{S}_c|}$. Therefore, for $1 \leq k < c$, $\frac{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}}{d(\mathcal{S}' \setminus \mathcal{S}_k) + \tau \hat{d}} > \frac{|\mathcal{S}_c|}{|\mathcal{S}_k|}$. This proves $\lim_{\alpha \rightarrow 0} \mathcal{R}_i(\mathcal{S}_c) < \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$. \square

Proof of Lemma 3.9:

$$\lim_{\max_{i \in \mathcal{S}_c} d_i / \hat{d} \rightarrow 0} \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c) = \frac{1}{c-1}.$$

Proof. $\frac{1}{\lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)} = \frac{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} = \frac{\sum_{k=1}^{c-1} (d(\mathcal{S}_c \setminus \mathcal{S}_k) + d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d})}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} = c - 1 + \frac{\sum_{k=1}^{c-1} d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}}$. As $\frac{\max_{i \in \mathcal{S}_c} d_i}{\hat{d}} \rightarrow 0$, we have $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\mathcal{S}' \setminus \mathcal{S}_c) + \tau \hat{d}} \rightarrow 0$ for $k < c$, which completes the proof. \square

Proof of Theorem 3.10: Suppose for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{|\mathcal{S}_c \setminus \mathcal{S}_k|} > \frac{d(\bar{\mathcal{S}}_c)}{|\mathcal{S}^* \setminus \mathcal{S}_c| + d(\bar{\mathcal{S}}_*)}$. Then $\lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c) > \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$.

Proof. Since for $1 \leq k < c$, $\frac{d(\mathcal{S}_c \setminus \mathcal{S}_k)}{d(\bar{\mathcal{S}}_c)} > \frac{|\mathcal{S}_c \setminus \mathcal{S}_k| \hat{d}}{|\mathcal{S}^* \setminus \mathcal{S}_c| \hat{d} + d(\bar{\mathcal{S}}_*)}$, we have $\frac{d(\bar{\mathcal{S}}_k)}{d(\bar{\mathcal{S}}_c)} = \frac{d(\mathcal{S}_c \setminus \mathcal{S}_k) + d(\bar{\mathcal{S}}_c)}{d(\bar{\mathcal{S}}_c)} > \frac{|\mathcal{S}_c \setminus \mathcal{S}_k| \hat{d} + |\mathcal{S}^* \setminus \mathcal{S}_c| \hat{d} + d(\bar{\mathcal{S}}_*)}{|\mathcal{S}^* \setminus \mathcal{S}_c| \hat{d} + d(\bar{\mathcal{S}}_*)} = \frac{|\mathcal{S}^* \setminus \mathcal{S}_k| \hat{d} + d(\bar{\mathcal{S}}_*)}{|\mathcal{S}^* \setminus \mathcal{S}_c| \hat{d} + d(\bar{\mathcal{S}}_*)}$. Therefore, for $1 \leq k < c$, $\frac{d(\bar{\mathcal{S}}_c)}{d(\bar{\mathcal{S}}_k)} < \frac{|\mathcal{S}^* \setminus \mathcal{S}_c| \hat{d} + d(\bar{\mathcal{S}}_*)}{|\mathcal{S}^* \setminus \mathcal{S}_k| \hat{d} + d(\bar{\mathcal{S}}_*)}$. This proves $\lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c) < \lim_{\alpha \rightarrow 0} \mathcal{R}_\partial(\mathcal{S}_c)$. \square

References

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