

Manifold Learning and Sparse Representation:

Assignment 3

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Exercise 46.

The results are listed in Table 1. (The code in Matlab is attached.)

Euclidean distance	0.734042
l1 distance	0.668856
Angular distance	0.773333
Cosine distance	0.645510

Thus, *Cosine distance* offers the best dissimilarity metrics in terms of classification.

Exercise 48.

Prove $\frac{1}{2} \sum_{i,j} W_{ij} \|\mathbf{f}_i - \mathbf{f}_j\|^2 = \text{tr}(\mathbf{F}\mathbf{L}_W\mathbf{F}^T)$, where $\mathbf{F} = (\mathbf{f}_1, \dots, \mathbf{f}_n)$.

Proof.

$$\begin{aligned}
 \frac{1}{2} \sum_{i,j} W_{ij} \|\mathbf{f}_i - \mathbf{f}_j\|^2 &= \frac{1}{2} \sum_{i,j} W_{ij} (\mathbf{f}_i - \mathbf{f}_j)^T (\mathbf{f}_i - \mathbf{f}_j) \\
 &= \frac{1}{2} \sum_{i,j} W_{ij} (\mathbf{f}_i^T \mathbf{f}_i + \mathbf{f}_j^T \mathbf{f}_j - \mathbf{f}_i^T \mathbf{f}_j - \mathbf{f}_j^T \mathbf{f}_i) \\
 &= \sum_i \left(\sum_j W_{ij} \right) \mathbf{f}_i^T \mathbf{f}_i - \sum_{i,j} W_{ij} \mathbf{f}_j^T \mathbf{f}_i \\
 &= \sum_i d_i \mathbf{f}_i^T \mathbf{f}_i - \sum_{i,j} W_{ij} \mathbf{f}_j^T \mathbf{f}_i \\
 &= \text{tr} \left(\sum_i d_i \mathbf{f}_i^T \mathbf{f}_i \right) - \text{tr} \left(\sum_{i,j} W_{ij} \mathbf{f}_j^T \mathbf{f}_i \right) \\
 &= \text{tr} \left(\sum_i \mathbf{f}_i d_i \mathbf{f}_i^T \right) - \text{tr} \left(\sum_{i,j} \mathbf{f}_i W_{ij} \mathbf{f}_j^T \right) \\
 &= \text{tr}(\mathbf{F}\mathbf{D}_W\mathbf{F}^T) - \text{tr}(\mathbf{F}\mathbf{W}\mathbf{F}^T) \\
 &= \text{tr}(\mathbf{F}(\mathbf{D}_W - \mathbf{W})\mathbf{F}^T) \\
 &= \text{tr}(\mathbf{F}\mathbf{L}_W\mathbf{F}^T)
 \end{aligned}$$

□

Derive the Laplacian matrix of $\frac{1}{2} \sum_{i,j} W_{ij}(f_i - f_j)^2$ under the constrain $\sum_i W_{ii} f_i^2 = 1$.

Let $\mathbf{g} \triangleq \mathbf{D}_W^{\frac{1}{2}} \mathbf{f}$ so that the constrain $\sum_i W_{ii} f_i^2 = 1$ deduces to $\|\mathbf{g}\| = 1$. Then, we have

$$\begin{aligned}
 \frac{1}{2} \sum_{i,j} W_{ij}(f_i - f_j)^2 &= \sum_i W_{ii} f_i^2 - \sum_{i,j} W_{ij} f_i f_j \\
 &= 1 - \mathbf{f}^T \mathbf{W} \mathbf{f} = \mathbf{g}^T \mathbf{g} - \mathbf{g}^T \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}} \mathbf{g} \\
 &= \mathbf{g}^T (\mathbf{I} - \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}}) \mathbf{g}^T \\
 &\triangleq \mathbf{g}^T \tilde{\mathbf{L}}_W \mathbf{g}^T
 \end{aligned}$$

where the Laplaceian matrix $\tilde{\mathbf{L}}_W = \mathbf{I} - \mathbf{D}^{-\frac{1}{2}} \mathbf{W} \mathbf{D}^{-\frac{1}{2}}$.