Mathematics, Pusan National University

Linear Algebra and Learning from Data IV.10 Distance Matrices

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Introduction



Suppose *n* points are at positions $\mathbf{x}_1, \ldots, \mathbf{x}_n$ in *d*-dimensional space. Distance matrix *D* is $n \times n$ matrix and $D_{ij} = ||\mathbf{x}_i - \mathbf{x}_j||^2$ It is easy to see that *D* is symmetric.

Example 1.1



Let $\mathbf{x}_1 = (1, 2, 3), \mathbf{x}_2 = (3, 2, 1), \mathbf{x}_3 = (1, 1, 2)$, then the distance matrix D is $D = \begin{bmatrix} 0 & 8 & 2 \\ 8 & 0 & 6 \\ 2 & 6 & 0 \end{bmatrix}$

Question

Can we recover the positions x_1, \ldots, x_n from the Euclidean distance matrix *D*?

Answer : No.

Example 1.2



Let $\mathbf{x}_1 = (1, 2, 3), \mathbf{x}_2 = (3, 2, 1), \mathbf{x}_3 = (1, 1, 2)$ and $\tilde{\mathbf{x}}_1 = (0, 1, 2), \tilde{\mathbf{x}}_2 = (2, 1, 0), \tilde{\mathbf{x}}_3 = (0, 0, 1)$, then the distance matrix *D* and \tilde{D} are same

$$\tilde{D} = D = \begin{bmatrix} 0 & 8 & 2 \\ 8 & 0 & 6 \\ 2 & 6 & 0 \end{bmatrix}$$



Another Question

Are there always positions x_1, \ldots, x_n consistent with the distance matrix *D*?

Answer : Yes!

Application of distance matrix

- 1. Wireless sensor networks
- 2. Shapes of molecules
- 3. Machine learning



Positions *X* from Distances *D*



et
$$X = \begin{bmatrix} | & | \\ \mathbf{x}_1 & \cdots & \mathbf{x}_n \\ | & | \end{bmatrix}$$
, then $X^T X = \begin{bmatrix} - & \mathbf{x}_1^T & - \\ \vdots \\ - & \mathbf{x}_n^T & - \end{bmatrix} \begin{bmatrix} | & | \\ \mathbf{x}_1 & \cdots & \mathbf{x}_n \\ | & | \end{bmatrix}$,

$$\|\mathbf{x}_i - \mathbf{x}_j\|^2 = (\mathbf{x}_i - \mathbf{x}_j)^T (\mathbf{x}_i - \mathbf{x}_j) = \mathbf{x}_i^T \mathbf{x}_i - \mathbf{x}_i^T \mathbf{x}_j - \mathbf{x}_j^T \mathbf{x}_i + \mathbf{x}_j^T \mathbf{x}_j$$
(1)

We can connect (1) and $X^T X$.

- Let $G = X^T X$, then the first term and last term $\mathbf{x}_k^T \mathbf{x}_k$ are in diag(G).
- And middle terms of (1) $-\mathbf{x}_i^T \mathbf{x}_j \mathbf{x}_j^T \mathbf{x}_i = -2\mathbf{x}_i^T \mathbf{x}_j$ is in -2G.

Then by using columns vector of *n* ones 1,

$$D = 1 \operatorname{diag}(G)^T - 2G + \operatorname{diag}(G)\mathbf{1}^T$$
(2)

Positions *X* from Distances *D*

Example 2.1

Let
$$\mathbf{x}_1 = (1, 2, 3), \mathbf{x}_2 = (3, 2, 1), \mathbf{x}_3 = (1, 1, 2), \text{ then } X = \begin{bmatrix} 1 & 3 & 1 \\ 2 & 2 & 1 \\ 3 & 1 & 2 \end{bmatrix}, G = X^T X = \begin{bmatrix} 14 & 10 & 9 \\ 10 & 14 & 7 \\ 9 & 7 & 6 \end{bmatrix}.$$

$$\mathbf{x}_1^T \mathbf{x}_1 = 14, \quad \mathbf{x}_2^T \mathbf{x}_2 = 14, \quad \mathbf{x}_3^T \mathbf{x}_3 = 6$$
$$\mathbf{x}_1^T \mathbf{x}_2 = 10, \quad \mathbf{x}_1^T \mathbf{x}_3 = 9, \quad \mathbf{x}_2^T \mathbf{x}_3 = 7$$
$$\mathbf{1} \operatorname{diag}(G)^T - 2G + \operatorname{diag}(G) \mathbf{1}^T = \begin{bmatrix} 0 & 8 & 2 \\ 8 & 0 & 6 \\ 2 & 6 & 0 \end{bmatrix} = D$$





Our problem is to recover ${\rm G}$ from ${\rm D}.$

If *G* is positive semidefinite, we can find *X* with $X^T X = G$.

Since $1 \operatorname{diag}(G)^T$ and $\operatorname{diag}(G) 1^T$ terms in (2) are rank 1 matrices, D has rank d + 1 or d + 2.

Now we solve equation (2) for G.

- 1. Place first point at the origin: $\mathbf{x}_1 = \mathbf{0}$.
- **2**. Then every $||\mathbf{x}_i \mathbf{x}_1|| = ||\mathbf{x}_i||$.
- 3. Then the first column \mathbf{d}_1 of D must be exactly the same as $\operatorname{diag}(X^T X) = \operatorname{diag}(G) = (\|\mathbf{x}_1\|^2, \|\mathbf{x}_2\|^2, \dots, \|\mathbf{x}_n\|^2).$

diag(G) =
$$\mathbf{d}_1$$
 and $G = -\frac{1}{2}(D - \mathbf{1}\mathbf{d}_1^T - \mathbf{d}_1\mathbf{1}^T).$ (3)

Positions X from Distances D

Theorem 2.2

A symmetric matrix D with zero diagonal is a Euclidian distance matrix if and only is $\mathbf{x}^T D \mathbf{x} \le 0$ for every vector with $\sum x_i = 0$ (so $\mathbf{1}^T \mathbf{x} = 0$).

We can solve $X^T X = G$ by several factorizations.

- Eigenvalue decomposition
 - If $G = Q\Lambda Q^T$, then $X = \sqrt{\Lambda}Q^T$.
- Cholesky factorization
 - If $G = U^T U$, then X = U.



Centering X or Rotating to Match Anchor Points

What is centering?

Centering : Often we might prefer to place the **centroid** of the x'x at the origin. The centroid of x_1, \ldots, x_n is just the average of those vectors:

Cdntroid
$$c = \frac{1}{n}(\mathbf{x}_1 + \dots + \mathbf{x}_n) = \frac{1}{n}X\mathbf{1}.$$

Just miltiply any position matrix X by the matrix $I - \frac{1}{n} \mathbf{1} \mathbf{1}^T$ to put the centroid at 0.



Centering X or Rotating to Match Anchor Points



Centering *X* or Rotating to Match Anchor Points

What is anchor point?

original original rotated rotated 2.5 anchor point anchor point 1.5 2 15 0.5 0.5 0 0 -0.5 -0.5 .1 -0.5 0.5 1.5 -1 -0.5 0.5 1.5 2.5

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The best orthogonal matrix ${\rm Q}$ solves the Procrustes problem in Section IV.9.

Orthogonal Procrustes problem

$$R = \arg\min_{Q} ||X - YQ||_F^2 \quad \text{subject to} \quad Q^TQ = I$$
(5)

(5) can be equivalent to find an orthogonal matrix R which most closely maps X to Y.





Problem 1

 $\|\mathbf{x}_1 - \mathbf{x}_2\|^2$ and $\|\mathbf{x}_2 - \mathbf{x}_3\|^2$ and $\|\mathbf{x}_1 - \mathbf{x}_3\|^2$ will violate the triangle inequality. Construct *G* and comfirm that it is not positive semidefinite : no solution *X* to $G = X^T X$.









Problem 2

 $\|\mathbf{x}_1 - \mathbf{x}_2\|^2 = 9$ and $\|\mathbf{x}_2 - \mathbf{x}_3\|^2 = 16$ and $\|\mathbf{x}_1 - \mathbf{x}_3\|^2 = 25$ do satisfy the triangle inequality 3 + 4 > 5. Construct *G* and find points $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3$ that match these distances.









Problem 3

If all $||\mathbf{x}_i - \mathbf{x}_j||^2 = 1$ for $\mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4$, find *G* and then *X*. The points lie in \mathbb{R}^d for which dimension *d*?





