

## Next

Methods Based for Representing of Finding Patterns in Data Based on Distances

- Multidimensional Scaling and Correspondence Analysis
- Cluster Analysis


## Distances

A distance (or dissimilarity) $d_{i j}$ between two points iand jmust satisfy:

1) Symmetry: $d_{i j}=d_{j i}$
2) Non-negativity: $d_{i j} \geq 0$
3) Identification: $d_{i j}=0$
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| How? |  |
| :---: | :---: |
| Let $X$ be the $n$ by $q$ matrix of coordinates we are looking for. |  |
| Consider $\mathrm{B}=\mathrm{XX}^{\top} \ldots$ |  |
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## Properties of Classical MDS

- If we are using Euclidean distance then the classical MDS result is the $\qquad$ same as the centered PCA solution.
- It gives the smallest values of $\qquad$

$$
\sum\left(d_{i j}^{2}-\hat{d}_{i j}^{2}\right)
$$

among all the possible projections into lower dimensional Euclidean
$\qquad$ space.

