OM Review - XVIII

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https://forms.gle/2tsHQa5xVvupV4Pq5

MDS objective function is:

$$\min_{Y} \|D(X) - D(Y)\|^2$$

where D(X) and D(Y) are $n \times n$ distance matrices in the original (say d) and compressed space respectively (say p).

If the distance function chosen is Euclidean, MDS reduces to PCA.

- (A) Always
- (B) Only if data lies on a linear subspace
- (C) Never

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Context

Ali's lecture discusses obtaining LLE as the following algorithm:

- (i) Construct K nearest neighbor graph
- (ii) Something with $\varepsilon(w) = \sum_{i=1}^n \|x_i \sum_{j=1}^K w_{ij} x_{v_i(j)}\|^2$
- (iii) Something with $\Phi(y) = \sum_{i=1}^t \|y_i \sum_{j=1}^t w_{ij}y_j\|^2$

In step 2 which variable are we optimizing over (what do we want to find)?

- (A) X
- (B) W
- (C) Y
- (D) All of the others

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In step 3 which variable are we optimizing over (what do we want to find)?

- (A) X
- (B) W
- (C) Y
- (D) All of the others

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In step 2, for any example x_i , consider w_i to be a vector of all weights in R^K space. How do we enforce that?

- (A) $w_i^T w_i = 1$
- (B) $e^T w_i = 1$ where $e \in R^K$ has all ones.
- (C) Both of the others
- (D) None of the others

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What is K in step 2 of the algorithm

- (A) The dimensionality of original data
- (B) The dimensionality of reduced data
- (C) The number of examples
- (D) The choice for the number of nearest neighbors.

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What happens when we choose to do LLE with very large number of neighbors?

- (A) LLE approaches to PCA
- (B) LLE keeps getting better and better

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