

# LOCALITY PRESERVING PROJECTION

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# Introduction

- *linear dimensionality reduction algorithm.*
- *Its an alternative to PCA – Principle component analysis.*
- *These are linear projective maps that arise by solving a variation problem that optimally preserves the neighborhood structure of the data set.*
- *LPP finds an embedding that preserves local information, and obtains a face space that best detects the essential manifold structure.*

# Dimensionality Reduction

## □ LINEAR METHODS:

- PCA – Principle Component Analysis.
- MDS – Multi Dimensional Scaling.
- LDA- Linear discriminant Analysis.

## □ NON LINEAR METHODS:

- ISOMAP-Isometric mapping methods.
- LLE-Locally Linear Embedding.
- Laplacian Eigen maps.

# Significance of LPP

- LPP is linear which is suitable for practical applications.
- The locality preserving quality is used in information retrieval applications.
- LPP is an approximation of non-linear methods that takes locality into account.
- It has higher accuracy when compared to PCA and LDA.



# ALGORITHM

## STEP 1 : Constructing the adjacency graph

Consider a graph  $G$  having  $k$  nodes and contains an edge between nodes  $i$  and  $j$  such that  $X_i$  and  $Y_j$  are close . There are 2 variations:

- a) There is an edge between nodes  $i$  and  $j$  where  $|| X_i - Y_j ||^2 < \epsilon$  (neighbor hoods  $\in \mathbb{R}^l$ ).
- b) There is an edge between node  $i$  if  $i$  is among  $n$  nearest neighbors of  $j$  Or  $j$  is among  $n$  nearest neighbors of  $i$ .

## STEP 2 : Determination of Weights

There are 2 variations :

- a) Heat Kernel: : If there is an edge between 2 vertices i and j then ,

$$W_{ij} = e^{-\frac{\|\mathbf{X}_i - \mathbf{X}_j\|^2}{t}}$$

- b) Simple Minded : If there is an edge connecting 2 vertices suppose i and j ,then  $W_{ij} = 1$  .

## STEP 3 : Eigen Maps

Determination Eigen Vectors and Eigen values for the problem:

$$XLX^T w = \lambda XDX^T w$$

Where L is the Laplacian matrix, i.e.  $D - S$ , where S corresponds to the similarity values defined, and D is a column matrix which reflects how important a certain projection is. The more data points that surround a given point



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