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LLE Algorithm Pseudocode

(Notes, e.g. [a] appear below)

Input X: D by N matrix consisting of N data items in D dimensions.

Output Y: d by N matrix consisting of $d < D$ dimensional embedding coordinates for the input points.

1. Find neighbours in X space [b,c].

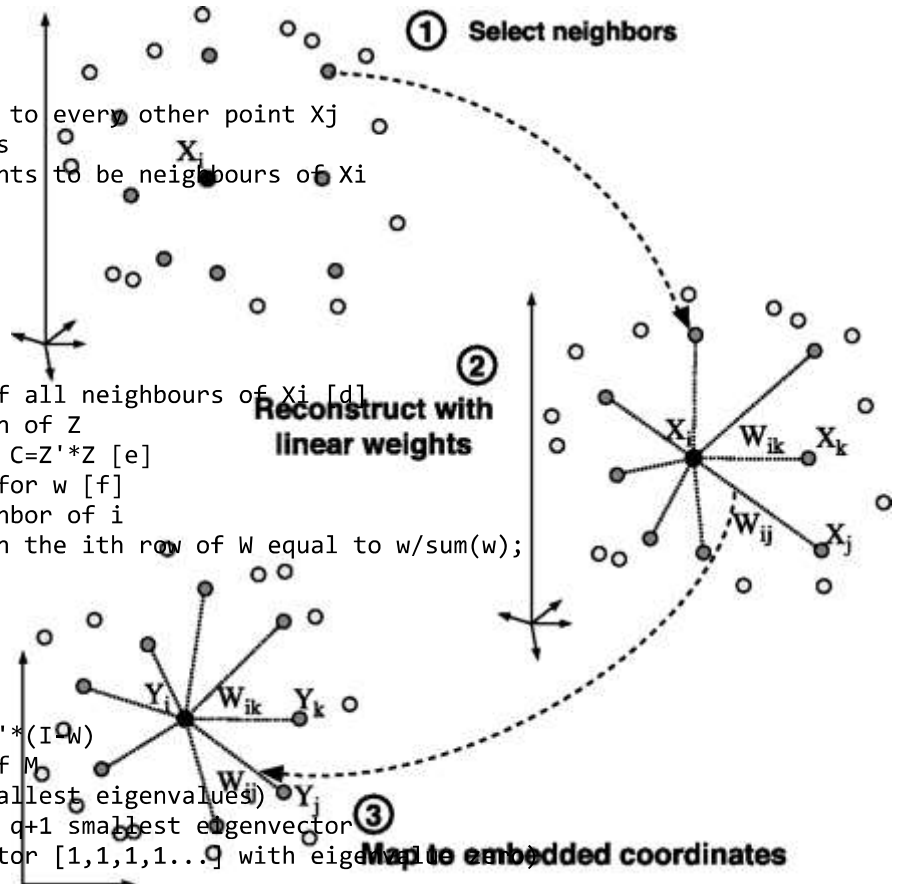
```
for i=1:N
  compute the distance from  $X_i$  to every other point  $X_j$ 
  find the K smallest distances
  assign the corresponding points to be neighbours of  $X_i$ 
end
```

2. Solve for reconstruction weights W.

```
for i=1:N
  create matrix Z consisting of all neighbours of  $X_i$  [d]
  subtract  $X_i$  from every column of Z
  compute the local covariance  $C=Z'*Z$  [e]
  solve linear system  $C*w = 1$  for w [f]
  set  $W_{ij}=0$  if j is not a neighbor of i
  set the remaining elements in the ith row of W equal to  $w/\text{sum}(w)$ ;
end
```

3. Compute embedding coordinates Y using weights W.

```
create sparse matrix  $M = (I-W)'*(I-W)$ 
find bottom d+1 eigenvectors of M
  (corresponding to the d+1 smallest eigenvalues)
set the qth ROW of Y to be the d+1 smallest eigenvector
  (discard the bottom eigenvector [1,1,1,1...] with eigenvalue 0)
```



Notes

[a] Notation

X_i and Y_i denote the i th column of X and Y
 (in other words the data and embedding coordinates of the i th point)
 M' denotes the transpose of matrix M
 $*$ denotes matrix multiplication
 (e.g. $M'*M$ is the matrix product of M left multiplied by its transpose)
 I is the identity matrix
 1 is a column vector of all ones

[b] This can be done in a variety of ways, for example above we compute the K nearest neighbours using Euclidean distance.
 Other methods such as epsilon-ball include all points within a certain radius or more sophisticated domain specific and/or adaptive local distance metrics.

[c] Even for simple neighbourhood rules like K-NN or epsilon-ball using Euclidean distance, there are highly efficient techniques for computing the neighbours of every point, such as KD trees.

[d] Z consists of all columns of X corresponding to the neighbours of X_i but not X_i itself

[e] If $K > D$, the local covariance will not be full rank, and it should be

regularized by setting $C = C + \epsilon I$ where I is the identity matrix and ϵ is a small constant of order $10^{-3} \times \text{trace}(C)$.
This ensures that the system to be solved in step 2 has a unique solution.

$\mathbf{1}$ denotes a column vector of all ones

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