

Name \_\_\_\_\_ Student ID \_\_\_\_\_ Total score \_\_\_\_\_  
Last First

**(8 pts) Problem 1**

(2 pts) 1.1

Which one of the clustering algorithms below may use ring-shaped cluster representatives?

- A. K-means  
B. GMM  
C. Fuzzy clustering  
D. Spectral clustering

(2 pts) **1.2** Which one of the following methods can learn parts-based object representation:

- A. PCA  
B. K-means  
C. NMF  
D. SVD

(2 pts) **1.4.** Assume that in MDS, the eigen decomposition on  $B_{\Delta} = -\frac{1}{2}JD^{(2)}J$  gives the following eigenvalue matrix:

$$\Lambda = \begin{bmatrix} 10 & 0 & 0 & 0 \\ 0 & -5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

Up to how many dimensions can we keep for data sample coordinates?

- A. 1  
B. 2  
C. 3  
D. 4  
E. None of the above

(2 pts) **1.4.** Assume that in MDS, the eigen decomposition on  $B_{\Delta} = -\frac{1}{2}JD^{(2)}J$  gives the following eigenvalue matrix:

$$\Lambda = \begin{bmatrix} 10 & 0 & 0 & 0 \\ 0 & -5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

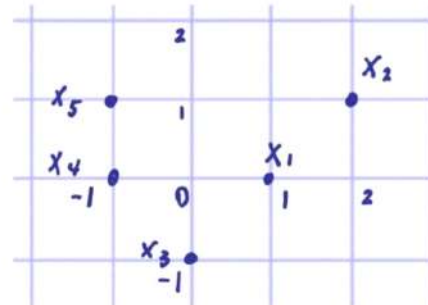
Is the distance matrix  $D$  a Euclidean distance matrix?

- A. No  
B. Yes  
C. None of the above

(21 pts) Problem 2

2.1 Referring to the figure on the left.

Compute  $d_1(x_1, x_5)$ ,  $d_\infty(x_1, x_5)$ ,  $S_{\cosine}(x_2, x_3)$ .



$$|x_{11} - x_{51}| + |x_{12} - x_{52}|$$

$$= 2 + 1 = 3$$

(4 pts)  $d_1(x_1, x_5) =$  3

$$\max \{ |x_{11} - x_{51}|, |x_{12} - x_{52}| \}$$

$$= 2$$

(4 pts)  $d_\infty(x_1, x_5) =$  2

$$x_2^T x_3 = \begin{bmatrix} 2 & 1 \end{bmatrix} \begin{bmatrix} 0 \\ -1 \end{bmatrix} = -1$$

$$\|x_2\| = \sqrt{5}, \quad \|x_3\| = 1$$

(4 pts)  $S_{\cosine}(x_2, x_3) =$   $\frac{-1}{\sqrt{5} \times 1} = -1/\sqrt{5}$

2.2 Assume that the five data samples are partitioned into two sets as  $C_1 = \{x_1, x_2\}$  and  $C_2 = \{x_3, x_4, x_5\}$ . Compute  $d_{min}^{ps}(x_2, C_2)$ , and  $d_{max}^{ss}(C_1, C_2)$  by using  $d_2(x_i, x_j)$  for sample-pair distance.

$$\sqrt{2^2 + 2^2} = \sqrt{8}$$

(4 pts)  $d_{min}^{ps}(x_2, C_2) =$   $\sqrt{8}$

$$\sqrt{1 + 3^2} = \sqrt{10}$$

(5 pts)  $d_{max}^{ss}(C_1, C_2) =$   $\sqrt{10}$

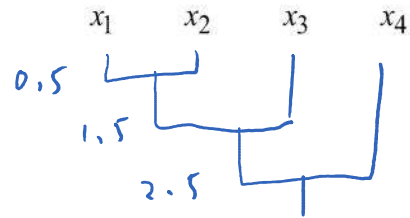
(24 pts) **Problem 3** Consider the following dissimilarity matrix P.

- (a) (10 pts) Use the complete-linkage clustering algorithm to generate the dendrogram.

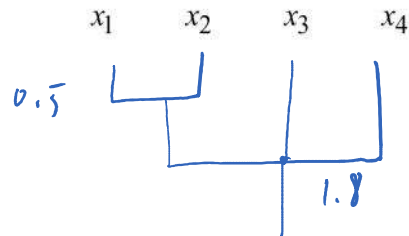
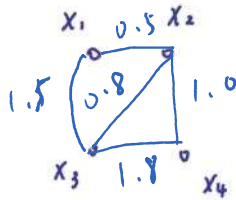
$$P = \begin{bmatrix} 0 & 0.5 & 1.5 & 2.5 \\ 0.5 & 0 & 0.8 & 1.0 \\ 1.5 & 0.8 & 0 & 1.8 \\ 2.5 & 1.0 & 1.8 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 1.5 & 2.5 \\ 1.5 & 0 & 1.8 \\ 2.5 & 1.8 & 0 \end{bmatrix}$$

$$\begin{bmatrix} 0 & 2.5 \\ 2.5 & 0 \end{bmatrix}$$



- (b) (10 pts) Use the graph theory based agglomerative clustering algorithm with  $h(k)$  the node degree of  $k = 2$  to generate the dendrogram.



- (c) (4 pts) What is the best choice of clustering in the criterion of “the longest survival time” based on the dendrogram you’ve derived in (b)?

Clustering of (b)  $\{x_1, x_2\}, \{x_3\}, \{x_4\}$

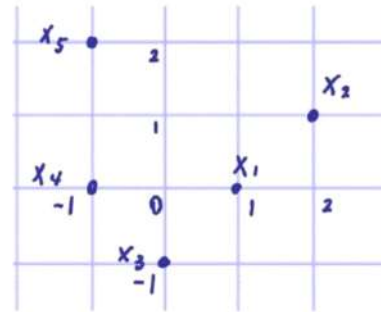
(27 pts) Problem 4

4.1. Five samples are shown in the figure:

$$\begin{bmatrix} x_1 & x_2 & x_3 & x_4 & x_5 \end{bmatrix} = \begin{bmatrix} 1 & 2 & 0 & -1 & -1 \\ 0 & 1 & -1 & 0 & 2 \end{bmatrix},$$

from which the EM algorithm is used to estimate parameters of a Gaussian mixture density (GMD) with two Gaussian components.

Assume that at one step of Expectation, the posterior probabilities of the Gaussian sources are obtained as:



	$t = 1$	$t = 2$	$t = 3$	$t = 4$	$t = 5$
$P(z_t = 1   x_t)$	0.8	1.0	0.4	0.2	0.0
$P(z_t = 2   x_t)$	0.2	0.0	0.6	0.8	1.0

- (a) (10 pts) For the subsequent Maximization step, compute the mixture weights of the two Gaussian sources,  $\pi_1^{new}$  and  $\pi_2^{new}$ , and the mean vector  $\mu_2^{new}$  (show your calculations clearly).

$$\pi_1^{new} = \frac{1}{5} (0.8 + 1.0 + 0.4 + 0.2 + 0.0) = \frac{2.4}{5} = 0.48$$

$$\pi_2^{new} = 0.52$$

$$N_2 = 2.6, \quad \mu_2 = \frac{1}{N_2} \left( 0.2 \begin{bmatrix} 1 \\ 0 \end{bmatrix} + 0.6 \begin{bmatrix} 0 \\ -1 \end{bmatrix} + 0.8 \begin{bmatrix} -1 \\ 0 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix} \right)$$

$$= \frac{1}{2.6} \begin{bmatrix} -1.6 \\ 1.4 \end{bmatrix} = \frac{1}{13} \begin{bmatrix} -8 \\ 7 \end{bmatrix}$$

$$\pi_1^{new} = 0.48$$

$$\pi_2^{new} = 0.52$$

$$\mu_2^{new} = \frac{1}{13} \begin{bmatrix} -8 \\ 7 \end{bmatrix}$$

- (b) (5 pts) Cluster the data samples to the Gaussian sources (C1 or C2) according to the maximum posterior probability criterion.

Data samples	$x_1$	$x_2$	$x_3$	$x_4$	$x_5$
Gaussian sources	C1	C1	C2	C2	C2

4.2 Refer to the same figure in 4.1 but perform k-means clustering on the five data samples instead. Assume that at the current iteration the cluster assignment is

$$C_1 = \{x_1, x_2, x_3\} \quad C_2 = \{x_4, x_5\}$$

(c) (6 pts) Compute the cluster centroids  $\mu_1$  and  $\mu_2$ :

$$\mu_1 = \frac{1}{3} \left( \begin{bmatrix} 1 \\ 0 \end{bmatrix} + \begin{bmatrix} 2 \\ 1 \end{bmatrix} + \begin{bmatrix} 0 \\ -1 \end{bmatrix} \right) = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$\mu_2 = \frac{1}{2} \left( \begin{bmatrix} -1 \\ 0 \end{bmatrix} + \begin{bmatrix} -1 \\ 2 \end{bmatrix} \right) = \begin{bmatrix} -1 \\ 1 \end{bmatrix}$$

(d) (6 pts) Compute the total distortion  $J$  based on the current cluster assignment and the cluster centroids.

$$\begin{aligned} J &= \left\| \begin{bmatrix} 1 \\ 0 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\|^2 + \left\| \begin{bmatrix} 2 \\ 1 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\|^2 + \left\| \begin{bmatrix} 0 \\ -1 \end{bmatrix} - \begin{bmatrix} 1 \\ 0 \end{bmatrix} \right\|^2 + \left\| \begin{bmatrix} -1 \\ 0 \end{bmatrix} - \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\|^2 + \left\| \begin{bmatrix} -1 \\ 2 \end{bmatrix} - \begin{bmatrix} -1 \\ 1 \end{bmatrix} \right\|^2 \\ &= 0 + 2 + 2 + 1 + 1 \\ &= 6 \end{aligned}$$

**(20 pts) Problem 5**

Assume that for latent semantic analysis a word-document matrix  $W$  has been decomposed by SVD as

$W = USV^T$ . Let

$$S = \begin{bmatrix} 10 & 0 & 0 & 0 \\ 0 & 5 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0.1 \end{bmatrix}, \quad V = \begin{bmatrix} v_{11} & v_{12} & v_{13} & v_{14} \\ v_{21} & v_{22} & v_{23} & v_{24} \\ v_{31} & v_{32} & v_{33} & v_{34} \\ v_{41} & v_{42} & v_{43} & v_{44} \end{bmatrix}$$

and  $U$  be represented by its column vectors as  $U = [u_1 \ u_2 \ u_3 \ u_4]$  with its size being  $10 \times 4$ .

(8 pts) (a) Consider a rank-1 approximation for  $W$ . Determine the approximated document-pair distance  $\|d_1 - d_3\|_2^2$  in terms of the relevant elements of  $V$ :

$$\begin{aligned} W &\approx u_1 \times 10 \times [v_{11} \ v_{21} \ v_{31} \ v_{41}] \\ &= u_1 \times [10v_{11} \ 10v_{21} \ 10v_{31} \ 10v_{41}] \end{aligned}$$

$$\|d_1 - d_3\|^2 \approx (10v_{11} - 10v_{31})^2 = 100(v_{11} - v_{31})^2$$

The approximation error  $\|W - \hat{W}\|_F^2 = 5^2 + 1^2 + 0.1^2 = 26.01$

(12 pts) (b) Consider a rank-2 approximation for  $W$ . Determine the approximated document-pair distance  $\|d_1 - d_3\|_2^2$  in terms of the relevant elements of  $V$ :

$$\begin{aligned} W &\approx [u_1 \ u_2] \begin{bmatrix} 10 & 0 \\ 0 & 5 \end{bmatrix} \begin{bmatrix} v_{11} & v_{21} & v_{31} & v_{41} \\ v_{12} & v_{22} & v_{32} & v_{42} \end{bmatrix} \\ &= [u_1 \ u_2] \begin{bmatrix} 10v_{11} & 10v_{21} & 10v_{31} & 10v_{41} \\ 5v_{12} & 5v_{22} & 5v_{32} & 5v_{42} \end{bmatrix} \end{aligned}$$

$$\|d_1 - d_3\|^2 \approx 100(v_{11} - v_{31})^2 + 25(v_{12} - v_{32})^2$$

The approximation error  $\|W - \hat{W}\|_F^2 = 1^2 + 0.1^2 = 1.01$