

# EXHIBIT 4 PART

## 260 UNSUPERVISED LEARNING AND CLUSTERING

(b) Show that the transfer of a sample  $\hat{\mathbf{x}}$  from  $\mathcal{X}_i$  to  $\mathcal{X}_j$  causes  $J_T$  to change to

$$J_T^* = J_T - \left[ \frac{n_j}{n_j + 1} (\hat{\mathbf{x}} - \mathbf{m}_j)^t S_T^{-1} (\hat{\mathbf{x}} - \mathbf{m}_j) - \frac{n_i}{n_i - 1} (\hat{\mathbf{x}} - \mathbf{m}_i)^t S_T^{-1} (\hat{\mathbf{x}} - \mathbf{m}_i) \right].$$

(c) Suggest an iterative procedure for minimizing  $J_T$ .

17. Use the facts that  $S_T = S_W + S_B$ ,  $J_e = \text{tr } S_W$ , and  $\text{tr } S_B = \sum n_i \|\mathbf{m}_i - \mathbf{m}\|^2$  to derive the equations given in Section 6.9 for the change in  $J_e$  resulting from transferring a sample  $\hat{\mathbf{x}}$  from cluster  $\mathcal{X}_i$  to cluster  $\mathcal{X}_j$ .

18. Let cluster  $\mathcal{X}_i$  contain  $n_i$  samples, and let  $d_{ij}$  be some measure of the distance between two clusters  $\mathcal{X}_i$  and  $\mathcal{X}_j$ . In general, one might expect that if  $\mathcal{X}_i$  and  $\mathcal{X}_j$  are merged to form a new cluster  $\mathcal{X}_k$ , then the distance from  $\mathcal{X}_k$  to some other cluster  $\mathcal{X}_h$  is not simply related to  $d_{hi}$  and  $d_{hj}$ . However, consider the equation

$$d_{hk} = \alpha d_{hi} + \alpha_j d_{hj} + \beta d_{ij} + \gamma |d_{hi} - d_{hj}|.$$

Show that the following choices for the coefficients  $\alpha_i$ ,  $\alpha_j$ ,  $\beta$ , and  $\gamma$  lead to the distance functions indicated. (For other cases, see Lance and Williams, 1967.)

(a)  $d_{\min}$ :  $\alpha_i = \alpha_j = 0.5$ ,  $\beta = 0$ ,  $\gamma = -0.5$ .

(b)  $d_{\max}$ :  $\alpha_i = \alpha_j = 0.5$ ,  $\beta = 0$ ,  $\gamma = 0.5$ .

(c)  $d_{\text{avg}}$ :  $\alpha_i = \frac{n_i}{n_i + n_j}$ ,  $\alpha_j = \frac{n_j}{n_i + n_j}$ ,  $\beta = \gamma = 0$ .

(d)  $d_{\text{mean}}^2$ :  $\alpha_i = \frac{n_i}{n_i + n_j}$ ,  $\alpha_j = \frac{n_j}{n_i + n_j}$ ,  $\beta = -\alpha_i \alpha_j$ ,  $\gamma = 0$ .

19. Consider a hierarchical clustering procedure in which clusters are merged so as to produce the smallest increase in the sum-of-squared error at each step. If the  $i$ th cluster contains  $n_i$  samples with sample mean  $\mathbf{m}_i$ , show that the smallest increase results from merging the pair of clusters for which

$$\frac{n_i n_j}{n_i + n_j} \|\mathbf{m}_i - \mathbf{m}_j\|^2$$

is minimum.

20. Consider the representation of the points  $\mathbf{x}_1 = (1 \ 0)^t$ ,  $\mathbf{x}_2 = (0 \ 0)^t$  and  $\mathbf{x}_3 = (0 \ 1)^t$  by a one-dimensional configuration. To obtain a unique solution, assume that the image points satisfy  $0 = y_1 < y_2 < y_3$ .

(a) Show that the criterion function  $J_{ee}$  is minimized by the configuration with

$$y_2 = (1 + \sqrt{2})/3 \text{ and } y_3 = 2y_2.$$

(b) Show that the criterion function  $J_{ff}$  is minimized by the configuration with

$$y_2 = (2 + \sqrt{2})/4 \text{ and } y_3 = 2y_2.$$