## Computer experiment 4

(EM algorithm for Gaussian mixtures)

- 1. Consider a mixture of three Gaussians  $\sum_{i=1}^{3} N(\mu_i, \Sigma_i) P_i$  with mean vectors  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}, \begin{bmatrix} 4 \\ 4 \end{bmatrix}, \begin{bmatrix} 10 \\ 2 \end{bmatrix}$  and covariance matrices  $\begin{bmatrix} 1 & 0.4 \\ 0.4 & 1 \end{bmatrix}, \begin{bmatrix} 1 & -0.6 \\ -0.6 & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$ , respectively. Generate 500 samples according to the following rule. The first two samples are generated from the  $2^{nd}$  Gaussian, the  $3^{rd}$  sample from the  $1^{st}$  one, and the  $4^{th}$  sample from the last Gaussian. This rule repeats until all 500 samples are generated.
  - (a) Use EM algorithms and the generated samples to estimate the unknown parameters  $\mu_i$ ,  $\Sigma_i$ ,  $P_i$  (i = 1,2,3). Please specify your experimental settings (e.g., initialization, stopping criterion) in the report.
  - (b) Repeat the mixture density estimation by EM when the mean vectors are  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ ,  $\begin{bmatrix} 3 \\ 3 \end{bmatrix}$ ,  $\begin{bmatrix} 6 \\ 2 \end{bmatrix}$ .
  - (c) Repeat the mixture density estimation by EM when the mean vectors are  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ ,  $\begin{bmatrix} 2 \\ 2 \end{bmatrix}$ ,  $\begin{bmatrix} 3 \\ 1 \end{bmatrix}$ .
  - (d) Compare the results (in terms of confusion matrices or 2-D visualization) and draw your conclusion.

(K-means algorithm)

2. Use k-means algorithm on the data set of the previous experiment, for k = 2,3,4. Compare the results and draw your conclusion.

(Linear discrimination functions)

- 3. Consider a two-class and 2-D classification task, where the feature vectors in each class are Gaussian pdfs  $N(\mu_i, \Sigma_i)$ , (i = 1,2) with mean vectors  $\mu_{1=}\begin{bmatrix} 1\\1 \end{bmatrix}$ ,  $\mu_2 =$ 
  - $\begin{bmatrix} 0 \\ 0 \end{bmatrix}$  and covariance matrices  $\Sigma_1 = \Sigma_2 = \begin{bmatrix} 0.2 & 1 \\ 0 & 0.2 \end{bmatrix}$ . Generate 100 vectors from each class. To guarantee **linear separability** of the classes, disregard vectors with  $x_1 + x_2 < 1$  for the 1<sup>st</sup> class and vectors with  $x_1 + x_2 > 1$  for the 2<sup>nd</sup> class.
  - (a) Apply the <u>perceptron algorithm</u> and the <u>sum-of-squared-error classifier</u> on

the data set.

- (b) Plot the data set and the decision lines.
- (c) You may try various initial values, apply different variants of the algorithms, or include additional constraints in your implementation. Compare and discuss the results.