第7次作业

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解

[1] N个样本点的联合概率为

 $L(\alpha, M, \Sigma) = \frac{1}{12} \frac{1}{$

 $\lfloor (X, \mu, \Sigma) = \bigcap_{i=1}^{N} \lfloor \ln \left(\sum_{j=1}^{k} X_{j} N(X) M_{j}, \Sigma_{j} \rfloor \right)$ 引入降变量 $Y = \{Y_{1}, Y_{1}, \dots, Y_{k}\}$. $Z \times \mathbb{R}_{3}$ 第 $i \not\in \mathbb{R}_{3}$ \mathbb{R}_{3} \mathbb{R}_{3} \mathbb{R}_{3}

 $\mathbb{R}^{||} P(\chi, \gamma | \lambda, \mu, \Sigma) = \prod_{i=1}^{N} \prod_{j=1}^{k} (\alpha_{j} N(\chi_{i} | \mu_{j}, \Sigma_{j}))^{\gamma_{i,j}}$

 $= \frac{1}{16} \left(\alpha_j \stackrel{\text{def}}{=} \chi_{i,j} + \left(N(\chi_i | M_j, \Sigma_j) \right)^{\chi_{i,j}} \right) \quad \text{with} \quad \phi$ $\ln P(X, Y | d, \mu, \bar{z}) = \sum_{j=1}^{k} \left(\left(\sum_{i=1}^{k} Y_{i,j} \right) \ln d_j + \sum_{j=1}^{k} Y_{i,j} \left(-\ln \bar{x} - \frac{1}{z} \ln |\bar{z}_j| - \frac{1}{z} (X_i - \mu_j)^T (X_i - \mu_j) \right) \right)$

面色期望为

Ex[[np(X,8|d,1,2) | X,d,1,2]

 $= \sum_{i=1}^{k} \left(\sum_{i=1}^{N} E(Y_{i,j} | X_{i,d}, \mu, \Sigma) | \mu d_j + \sum_{i=1}^{N} E(Y_{i,j} | X_{i,d}, \mu, \Sigma) | -|\mu X_{i,d} - \frac{1}{2} | \mu | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \mu | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \mu | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \mu | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \mu | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | \Sigma_j | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (\Sigma_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i - \mu_j)^T (X_i - \mu_j) | - \frac{1}{2} (X_i$ $\sharp + E(Y_{i,j} \mid X_i, \lambda, \mu, \Sigma) = \rho(Y_{i,j} = 1 \mid X_i, \lambda, \mu, \Sigma) = \frac{P(Y_{ij} = 1, X_i \mid \alpha, \mu, \Sigma)}{\rho(Y_i)}$

 $= \frac{\alpha_{j} N(X_{i} | M_{j}, \Sigma_{j})}{\sum_{i=1}^{k} \alpha_{j} N(X_{i} | M_{j}, \Sigma_{j})}$

(L)为3将上式局些期望最大、我们)随机和放化一起(d,M,E)记为(d,M',E°) MAGA d^{t+1} , μ^{t+1} , $\Sigma^{t+1} = \operatorname{argmax} E_{\gamma}(|\Delta P(X, X|d, \mu, \Sigma)|X, d^{t}, \mu^{t}, \Sigma^{t})$

$$\frac{2}{2} \frac{1}{2} \frac{1$$