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goal: 半监督的高斯混合模型的EM算法推导过程

$$L = \prod_{D_l} P(X_j, Y_j | \theta) \prod_{D_u} P(X_j | \theta) = \prod_{D_l} P(X_j | \theta) P(Y_j | \theta, X_j) \prod_{D_u} P(X_j | \theta) \\ = \prod_{D_l} [\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i) P(Y_j | \mu_i, \Sigma_i, X_j)] \prod_{D_u} [\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)]$$

$$P(X_j | \mu_i, \Sigma_i) = \frac{1}{(2\pi)^{\frac{p}{2}} |\Sigma_i|^{\frac{1}{2}}} \exp(-\frac{1}{2} (X_j - \mu_i)^T \Sigma_i^{-1} (X_j - \mu_i))$$

$$P(Y_j | \mu_i, \Sigma_i, X_j) = \begin{cases} 0 & i = Y_j \\ 1 & i \neq Y_j \end{cases}$$

$$l = \log L = \sum_{D_l} \log(\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i) I(Y_j = i)) + \sum_{D_u} \log(\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i))$$

$$\text{记 } A = \sum_{D_l} \log(\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i) I(Y_j = i))$$

$$B = \sum_{D_u} \log(\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i))$$

$$\frac{\partial A}{\partial \mu_i} = \sum_{D_l} \frac{\alpha_i \frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \mu_i} I(Y_j = i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i) I(Y_j = i)} = \sum_{D_l} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) (X_j - \mu_i) I(Y_j = i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i) I(Y_j = i)}$$

$$\begin{cases} \sum_{D_l, Y_j=i} (X_j - \mu_i) & i = Y_j \\ 0 & i \neq Y_j \end{cases}$$

$$\frac{\partial A}{\partial \mu_i} = \sum_{D_l, Y_j=i} (X_j - \mu_i)$$

$$\frac{\partial B}{\partial \mu_i} = \sum_{D_u} \frac{\alpha_i \frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \mu_i}}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)} = \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) (X_j - \mu_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}$$

$$\frac{\partial l}{\partial \mu_i} = \frac{\partial A}{\partial \mu_i} + \frac{\partial B}{\partial \mu_i} = \sum_{D_l, Y_j=i} (X_j - \mu_i) + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) (X_j - \mu_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)} = 0$$

$$(1) \quad \mu_i = \frac{\sum_{D_l, Y_j=i} X_j + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) X_j}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}}{\sum_{D_l, Y_j=i} 1 + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}}$$

$\partial \Sigma_i(m, n)$ 为 Σ_i 的 m 行 n 列的元素

$$\frac{\partial B}{\partial \Sigma_i(m, n)} = \sum_{D_u} \frac{\alpha_i \frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \Sigma_i(m, n)}}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}$$

$$\frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \Sigma_i(m, n)} = \frac{1}{(2\pi)^{\frac{p}{2}} |\Sigma_i|^{\frac{1}{2}}} \left(-\frac{1}{2} \frac{\partial |\Sigma_i|}{\partial \Sigma_i(m, n)} \right) \exp(-\frac{1}{2} (X_j - \mu_i)^T \Sigma_i^{-1} (X_j - \mu_i)) + \\ \frac{1}{(2\pi)^{\frac{p}{2}} |\Sigma_i|^{\frac{1}{2}}} \exp(-\frac{1}{2} (X_j - \mu_i)^T \Sigma_i^{-1} (X_j - \mu_i)) \left(-\frac{1}{2} \frac{\partial D}{\partial \Sigma_i(m, n)} \right), \text{ where } D = (X_j - \mu_i)^T \Sigma_i^{-1} (X_j - \mu_i)$$

先看前半部分, 由于 $|\Sigma_i| = \sum_{(m \text{ 或 } n)} \Sigma_i(m, n) * A_{i(m, n)}$, $A_{i(m, n)}$ 为对应元素的代数余子式

所以 $\frac{\partial |\Sigma_i|}{\partial \Sigma_i(m, n)} = A_{i(m, n)}$, 所以前半部分为: $-\frac{1}{2} P(X_j | \mu_i, \Sigma_i) \frac{A_{i(m, n)}}{|\Sigma_i|}$

$$\text{再看后半部分 } \frac{\partial D}{\partial \Sigma_i(m, n)} = (X_j - \mu_i)^T \frac{\partial \Sigma_i^{-1}}{\partial \Sigma_i(m, n)} (X_j - \mu_i)$$

$$\text{由于 } \frac{\partial I}{\partial \Sigma_i(m, n)} = 0, \text{ 得到 } \frac{\partial \Sigma_i \Sigma_i^{-1}}{\partial \Sigma_i(m, n)} = \frac{\partial \Sigma_i}{\partial \Sigma_i(m, n)} \Sigma_i^{-1} + \Sigma_i \frac{\partial \Sigma_i^{-1}}{\partial \Sigma_i(m, n)} = 0$$

$$\text{移项得 } \frac{\partial \Sigma_i^{-1}}{\partial \Sigma_i(m, n)} = -\Sigma_i^{-1} \frac{\partial \Sigma_i}{\partial \Sigma_i(m, n)} \Sigma_i^{-1}, \text{ 代入上式得}$$

$$\frac{\partial D}{\partial \Sigma_i(m,n)} = -(X_j - \mu_i)^T \Sigma_i^{-1} \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} \Sigma_i^{-1} (X_j - \mu_i) = -[\Sigma_i^{-1} (X_j - \mu_i)]^T \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} [\Sigma_i^{-1} (X_j - \mu_i)],$$

where, $\frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)}$ 为在m,n处为1, 其余为0的矩阵

$$\text{所以后半部分为 } \frac{1}{2} P(X_j | \mu_i, \Sigma_i) [\Sigma_i^{-1} (X_j - \mu_i)]^T \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} [\Sigma_i^{-1} (X_j - \mu_i)]$$

$$\frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \Sigma_i(m,n)} = -\frac{1}{2} P(X_j | \mu_i, \Sigma_i) \frac{A_i(m,n)}{|\Sigma_i|} + \frac{1}{2} P(X_j | \mu_i, \Sigma_i) [\Sigma_i^{-1} (X_j - \mu_i)]^T \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} [\Sigma_i^{-1} (X_j - \mu_i)]$$

$$\text{所以, } \frac{\partial B}{\partial \Sigma_i(m,n)} = \sum_{D_u} \frac{\alpha_i (-\frac{1}{2} P(X_j | \mu_i, \Sigma_i) \frac{A_i(m,n)}{|\Sigma_i|} + \frac{1}{2} P(X_j | \mu_i, \Sigma_i) [\Sigma_i^{-1} (X_j - \mu_i)]^T \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} [\Sigma_i^{-1} (X_j - \mu_i)])}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}$$

可以发现前半部分 $\frac{A_i(m,n)}{|\Sigma_i|}$ 为 Σ_i^{-1} 的 m, n 位置的元素; 且由于 $\frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)}$ 为在 m, n 处为1, 其余为0的矩阵, 所以后半部分为 $\Sigma_i^{-1} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1}$ 的 m, n 元素

$$\text{所以 } \frac{\partial B}{\partial \Sigma_i} = \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) [-\frac{1}{2} \Sigma_i^{-1} + \frac{1}{2} (\Sigma_i^{-1} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1})]}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}$$

同理, 可以证明得到如下:

$$\frac{\partial A}{\partial \Sigma_i(m,n)} = \sum_{D_l, Y_j=i} \frac{\alpha_i \frac{\partial P(X_j | \mu_i, \Sigma_i)}{\partial \Sigma_i(m,n)}}{\alpha_i P(X_j | \mu_i, \Sigma_i)} = \sum_{D_l, Y_j=i} -\frac{1}{2} \frac{A_i(m,n)}{|\Sigma_i|} + \frac{1}{2} [\Sigma_i^{-1} (X_j - \mu_i)]^T \frac{\partial \Sigma_i}{\partial \Sigma_i(m,n)} [\Sigma_i^{-1} (X_j - \mu_i)]$$

$$\text{所以 } \frac{\partial A}{\partial \Sigma_i} = \sum_{D_l, Y_j=i} [-\frac{1}{2} \Sigma_i^{-1} + \frac{1}{2} (\Sigma_i^{-1} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1})]$$

$$\frac{\partial l}{\partial \Sigma_i} = \frac{\partial A}{\partial \Sigma_i} + \frac{\partial B}{\partial \Sigma_i} = \sum_{D_l, Y_j=i} [-\frac{1}{2} \Sigma_i^{-1} + \frac{1}{2} (\Sigma_i^{-1} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1})] + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) [-\frac{1}{2} \Sigma_i^{-1} + \frac{1}{2} (\Sigma_i^{-1} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1})]}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}$$

$$\sum_{D_l, Y_j=i} (-1) + \sum_{D_l, Y_j=i} (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1} - \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)} + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) (X_j - \mu_i) (X_j - \mu_i)^T \Sigma_i^{-1}}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)} = 0$$

$$(2) \Sigma_i = \frac{\sum_{D_l, Y_j=i} (X_j - \mu_i) (X_j - \mu_i)^T + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i) (X_j - \mu_i) (X_j - \mu_i)^T}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}}{\sum_{D_l, Y_j=i} 1 + \sum_{D_u} \frac{\alpha_i P(X_j | \mu_i, \Sigma_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}}$$

由于 $\sum \alpha_i = 1, \alpha_i \geq 0$, 所以使用拉格朗日乘子法求解:

$$U = l + \lambda (\sum_{i=1}^N \alpha_i - 1)$$

$$\frac{\partial U}{\partial \alpha_i} = \sum_{D_l, Y_j=i} \frac{1}{\alpha_i} + \sum_{D_u} \frac{P(X_j | \mu_i, \Sigma_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)} + \lambda = 0$$

将上式乘以 α_i , 并从 $i=1$ 累加到 $i=N$, 可以计算得到 $\lambda = l_o + u_o$, l_o 为 D_l 的个数, u_o 为 D_u 的个数

$$(3) \text{ 所以 } \alpha_i = \frac{\sum_{D_l, Y_j=i} 1 + \sum_{D_u} \frac{P(X_j | \mu_i, \Sigma_i)}{\sum_{i=1}^N \alpha_i P(X_j | \mu_i, \Sigma_i)}}{l_o + u_o}$$