

EM – algorithm 求取 $\alpha_j^{(t+1)}$

$$Q(\theta|\theta^{(t)}) \equiv E_y[\log P(x, y|\theta) | x, \theta^{(t)}] = \sum_i^n \sum_j^k \left[\log(\alpha_j \phi(x_i | \mu_j, \sigma_j)) P(y_i = j | x_i, \theta^{(t)}) \right]$$

給定限制式： $\sum_j^k \alpha_j = 1$

找 α_j 使得 $Q(\theta|\theta^{(t)})$ 最大

使用 Lagrange 乘數

$$L = \sum_i^n \sum_j^k \left[\log(\alpha_j \phi(x_i | \mu_j, \sigma_j)) P(y_i = j | x_i, \theta^{(t)}) \right] + \lambda (\sum_j^k \alpha_j - 1)$$

$$\frac{\partial}{\partial \alpha_j} L = \sum_i^n \frac{\phi(x_i | \mu_j, \sigma_j)}{\alpha_j \phi(x_i | \mu_j, \sigma_j)} P(y_i = j | x_i, \theta^{(t)}) + \lambda$$

$$\frac{\partial}{\partial \lambda} L = \sum_j^k \alpha_j - 1$$

令 $\frac{\partial}{\partial \alpha_j} L$ 以及 $\frac{\partial}{\partial \lambda} L$ 為 0

$$\frac{\partial}{\partial \alpha_j} L = 0 \Rightarrow \lambda = - \sum_i^n \frac{\phi(x_i | \mu_j, \sigma_j)}{\alpha_j \phi(x_i | \mu_j, \sigma_j)} P(y_i = j | x_i, \theta^{(t)}) \rightarrow \alpha_j = - \sum_i^n \frac{P(y_i = j | x_i, \theta^{(t)})}{\lambda} \quad \text{--- (1)}$$

$$\frac{\partial}{\partial \lambda} L = 0 \Rightarrow \sum_j^k \alpha_j = 1 \quad \text{--- (2)}$$

將(1)代入(2)

$$- \sum_j^k \sum_i^n \frac{P(y_i = j | x_i, \theta^{(t)})}{\lambda} = 1 \Rightarrow \lambda = - \sum_j^k \sum_i^n P(y_i = j | x_i, \theta^{(t)})$$

$$\because \sum_j^k P(y_i = j | x_i, \theta^{(t)}) = 1 \therefore \lambda = - \sum_i^n 1 = -n \quad \text{--- (3)}$$

將(3)代回(1)得到 α_j

$$\alpha_j^{(t+1)} = - \sum_i^n \frac{P(y_i = j | x_i, \theta^{(t)})}{-n} = \frac{\sum_i^n P(y_i = j | x_i, \theta^{(t)})}{n}$$