

1 GLMER E-Step

$$\begin{aligned}
P(Z_i = 1|y, \theta) &= \frac{\int \prod_j f(y_{ij}|X, \beta_1)g(v)dv}{\int \prod_j f(y_{ij}|X, \beta_1)f(v)dv + \int \prod_j f(y_{ij}|X, \beta_0)g(v)dv} \\
&\approx \frac{\sum_{m=1}^M \frac{g(v)}{\hat{g}_1(v)} \prod_j f(y_{ij}|X, \beta_1, v_m)}{\sum_{m=1}^M \frac{g(v)}{\hat{g}_1(v)} \prod_j f(y_{ij}|X, \beta_1, v_m) + \frac{g(v)}{\hat{g}_0(v)} \prod_j f(y_{ij}|X, \beta_0, v_m)} \\
&= \left(1 + \frac{\sum_{m=1}^M \frac{g(v)}{\hat{g}_0(v)} \prod_j f(y_{ij}|X, \beta_0, v_m)}{\sum_{m=1}^M \frac{g(v)}{\hat{g}_1(v)} \prod_j f(y_{ij}|X, \beta_1, v_m)} \right)^{-1}
\end{aligned}$$

2 Some Computation

$$\begin{aligned}\mathcal{L}(\theta) &= \int \prod_{k=1}^K f_{\theta}(x_k, v) \pi(v) dv \\ l(\theta) &= \log \int \prod_{k=1}^K f_{\theta}(x_k, v) \pi(v) dv \\ \frac{\partial}{\partial \theta} l(\theta) &= \frac{1}{\mathcal{L}(\theta)} \int \frac{\partial}{\partial \theta} \prod_{k=1}^K f_{\theta}(x_k) \pi(v) dv \\ &= \frac{1}{\mathcal{L}(\theta)} \int \prod_{k=1}^K f_{\theta}(x_k, v) \left(\sum_{k=1}^K \frac{\partial}{\partial \theta} \log f_{\theta}(x_k, v) \right) \pi(v) dv \\ &\approx \left(\sum_{j=1}^J \prod_{k=1}^K f_{\theta}(x_k, v_j) \right)^{-1} \left(\sum_{j=1}^J \prod_{k=1}^K f_{\theta}(x_k, v_j) \left(\sum_{k=1}^K \frac{\partial}{\partial \theta} \log f_{\theta}(x_k, v_j) \right) \right)\end{aligned}$$

3 Constrained Random Effect

$$\begin{aligned} \min & \frac{1}{2} \sum_{i=1}^n \sum_{k=1}^K \sum_{j=1}^J w_{ijk} (x_{ijk} - \hat{\mu}_{ik})^2 \\ & w_{ik} := \sum_j w_{ijk} \\ \text{s.t. } & \sum_i w_{ik} \mu_{ik} = 0 \quad \forall k \end{aligned}$$

For a set k :

$$L_k = \frac{1}{2} \sum_i \sum_j w_{ij} (x_{ij} - \hat{\mu}_i)^2 + \lambda \sum_i w_i \hat{\mu}_i$$

Solve a system of equations:

$$\begin{aligned} - \sum_j w_{ij} (x_{ij} - \hat{\mu}_i) + \lambda w_i &= 0 \quad \forall i \\ \sum_i w_i \hat{\mu}_i &= 0 \\ \Rightarrow \lambda &= \frac{\sum_{ij} w_{ij} x_{ij}}{\sum_i w_i} \\ \Rightarrow \hat{\mu}_i &= \frac{\sum_j w_{ij} x_{ij} - \lambda w_i}{w_i} = \frac{\sum_j w_{ij} x_{ij}}{\sum_j w_{ij}} - \lambda := \bar{x}_i - \bar{\mu} \end{aligned}$$