

$$\begin{cases} \logit [P(Y_{ij}=1)] = \beta_0 + \beta_1 x_i + u_i & i=1, \dots, I \\ u_i \sim N(0, 1/\tau_u) & j=1, \dots, J \end{cases}$$

$$\text{let } \underline{\theta} = (\beta_0, \beta_1, \tau_u)$$

$$P(\underline{\theta} | \underline{u}, \text{data}) = \prod_{i=1}^I \prod_{j=1}^J \frac{[\exp(\beta_0 + \beta_1 x_i + u_i)]^{I(Y_{ij}=1)}}{1 + \exp(\beta_0 + \beta_1 x_i + u_i)}$$

$$P(\underline{\theta}, \underline{u} | \text{data}) \propto \prod_{i=1}^I \left\{ \prod_{j=1}^J \frac{[\exp(\beta_0 + \beta_1 x_i + u_i)]^{I(Y_{ij}=1)}}{1 + \exp(\beta_0 + \beta_1 x_i + u_i)} \right\} \cdot \tau_u^{\frac{1}{2}} \exp\left\{-\frac{u_i^2 \cdot \tau_u}{2}\right\}$$

$$\begin{aligned} \log P(\underline{\theta}, \underline{u} | \text{data}) &\propto \sum_{i=1}^I \left( \sum_{j=1}^J \left\{ I(Y_{ij}=1)(\beta_0 + \beta_1 x_i + u_i) - \log[1 + \exp(\beta_0 + \beta_1 x_i + u_i)] \right\} \right. \\ &\quad \left. + \frac{1}{2} \log \tau_u - \frac{u_i^2 \tau_u}{2} \right) \\ &\propto \sum_{i=1}^I \left( \sum_{j=1}^J \left\{ I(Y_{ij}=1)(\beta_0 + \beta_1 x_i + u_i) - \log[1 + \exp(\beta_0 + \beta_1 x_i + u_i)] \right\} \right) \\ &\quad + \frac{I}{2} \log \tau_u - \frac{\sum_{i=1}^I u_i^2}{2} \tau_u. \end{aligned}$$

Full conditionals

GLMM-②

$$\beta_0, \beta_1 | \text{others} \propto \sum_{i=1}^I \left( \sum_{j=1}^J \left\{ I(y_{ij}=1) (\beta_0 + \beta_1 x_i + u_i) - \log[1 + \exp(\beta_0 + \beta_1 x_i + u_i)] \right\} \right)$$

$$\tau_u | \text{others} \propto \frac{I}{2} \log \tau_u - \frac{\sum_{i=1}^I u_i^2}{2} \tau_u$$

$$\propto \left( \frac{I}{2} + 1 - 1 \right) \log \tau_u - \frac{\sum_{i=1}^I u_i^2}{2} \tau_u$$

$$\therefore \tau_u | \text{others} \sim \text{Gam}\left(\frac{I}{2} + 1, 0.5 \sum_{i=1}^I u_i^2\right)$$

$$u_i | \text{others} \propto \sum_{j=1}^J \left\{ I(y_{ij}=1) (\beta_0 + \beta_1 x_i + u_i) - \log[1 + \exp(\beta_0 + \beta_1 x_i + u_i)] \right\}$$

$$= \frac{u_i^2}{2} \tau_u \quad \text{for } i=1, \dots, I.$$