

- Joint approximate conditional posterior in blocked Gibbs
  - Likelihood is approximated

$$p(\alpha_{j}, \gamma_{j} \mid \text{rest}) = \prod_{i=1}^{I} N\left(\alpha_{j} + \beta_{j} x_{1,ij} + \gamma_{j} x_{2j} \mid \log y_{ij}, \frac{1}{y_{ij}}\right) N\left(\alpha_{j} \mid \mu_{\alpha}, \sigma_{\alpha}^{2}\right) N\left(\gamma_{j} \mid \mu_{\gamma}, \sigma_{\gamma}^{2}\right)$$

More approximation

$$p(\alpha_{j}, \gamma_{j} \mid \text{rest}) \sim N(\alpha_{z} \mid \mu_{z}, \sigma_{z}^{2}), z = \alpha_{j} \gamma_{j} \prod_{i=1}^{I} (\alpha_{j} + \beta_{j} x_{1,ij} + \gamma_{j} x_{2j})$$

$$\mu_{z} = \mu_{\alpha} \mu_{\gamma} \prod_{i=1}^{I} \log y_{ij}$$

$$\sigma_{z}^{2} = (\mu_{\alpha}^{2} + \sigma_{\alpha}^{2}) (\mu_{\gamma}^{2} + \sigma_{\gamma}^{2}) \prod_{i=1}^{I} (\log^{2} y_{ij} + \frac{1}{y_{ij}}) - \mu_{z}^{2}$$

Refs.: <u>Approximating the Distribution for Sums of Products of Normal Variables</u>, Robert Ware and Frank Lad, 2003.