

# BINOMIAL GENERALIZED LINEAR MIXED MODEL

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## Description

A generalized linear mixed model with varying coefficients for binomially distributed data.

## Implementation

The file `binomial.varying.coef.sim.R` simulates data according to the model statement presented below, and `binomial.varying.coef.mcmc.R` contains the MCMC algorithm for model fitting.

## Model statement

Let  $z_{ij}$  be the number of “successes” (i.e.,  $z_{ij}$  are integers greater than or equal to 0) out of  $N_{ij}$  “trials”, for  $i = 1, \dots, n_j$  and  $j = 1, \dots, J$ , where the index  $i$  denotes replicate observations within group  $j$ . Furthermore, let  $\mathbf{x}_{ij}$  be a vector of covariates associated with  $z_{ij}$  and  $\boldsymbol{\beta}_j$  be the corresponding vector of coefficients for group  $j$ .

$$\begin{aligned} z_{ij} &\sim \text{Binom}(N_{ij}, p_{ij}) \\ \text{logit}(p_{ij}) &= \mathbf{x}_{ij}' \boldsymbol{\beta}_j \\ \boldsymbol{\beta}_j &\sim \mathcal{N}(\boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}) \\ \boldsymbol{\mu}_\beta &\sim \mathcal{N}(\mathbf{0}, \sigma_\beta^2 \mathbf{I}) \\ \boldsymbol{\Sigma}^{-1} &\sim \text{Wish}(\mathbf{S}_0^{-1}, \nu) \end{aligned}$$

## Full conditional distributions

*Regression coefficients ( $\boldsymbol{\beta}_j$ ):*

$$\begin{aligned} [\boldsymbol{\beta}_j \mid \cdot] &\propto \prod_{i=1}^{n_j} [z_{ij} \mid N_{ij}, \mathbf{x}_{ij}' \boldsymbol{\beta}_j] [\boldsymbol{\beta}_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}] \\ &\propto \prod_{i=1}^{n_j} \text{Binom}(z_{ij} \mid N_{ij}, \mathbf{x}_{ij}' \boldsymbol{\beta}_j) \mathcal{N}(\boldsymbol{\beta}_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}). \end{aligned}$$

The update for  $\boldsymbol{\beta}_j$  proceeds using Metropolis-Hastings.

*Mean of regression coefficients ( $\boldsymbol{\mu}_\beta$ ):*

$$\begin{aligned} [\boldsymbol{\mu}_\beta \mid \cdot] &\propto \prod_{j=1}^J [\boldsymbol{\beta}_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}] [\boldsymbol{\mu}_\beta \mid \mathbf{0}, \sigma_\beta^2 \mathbf{I}] \\ &\propto \prod_{j=1}^J \mathcal{N}(\boldsymbol{\beta}_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}) \mathcal{N}(\boldsymbol{\mu}_\beta \mid \mathbf{0}, \sigma_\beta^2 \mathbf{I}) \\ &\propto \exp \left\{ \sum_{j=1}^J \left( -\frac{1}{2} (\boldsymbol{\beta}_j - \boldsymbol{\mu}_\beta)' \boldsymbol{\Sigma}^{-1} (\boldsymbol{\beta}_j - \boldsymbol{\mu}_\beta) \right) \right\} \end{aligned}$$

$$\begin{aligned}
& \times \exp \left\{ -\frac{1}{2} (\boldsymbol{\mu}_\beta - \mathbf{0})' (\sigma_\beta^2 \mathbf{I})^{-1} (\boldsymbol{\mu}_\beta - \mathbf{0}) \right\} \\
& \propto \exp \left\{ -\frac{1}{2} \left( -2 \left( \sum_{j=1}^J \beta_j' \boldsymbol{\Sigma}^{-1} \right) \boldsymbol{\mu}_\beta + \boldsymbol{\mu}_\beta' (J \boldsymbol{\Sigma}^{-1}) \boldsymbol{\mu}_\beta \right) \right\} \\
& \quad \times \exp \left\{ -\frac{1}{2} (\boldsymbol{\mu}_\beta' (\sigma_\beta^2 \mathbf{I})^{-1} \boldsymbol{\mu}_\beta) \right\} \\
& \propto \exp \left\{ -\frac{1}{2} \left( -2 \left( \sum_{j=1}^J \beta_j' \boldsymbol{\Sigma}^{-1} \right) \boldsymbol{\mu}_\beta + \boldsymbol{\mu}_\beta' (J \boldsymbol{\Sigma}^{-1} + (\sigma_\beta^2 \mathbf{I})^{-1}) \boldsymbol{\mu}_\beta \right) \right\} \\
& = \mathcal{N}(\mathbf{A}^{-1} \mathbf{b}, \mathbf{A}^{-1}),
\end{aligned}$$

where  $\mathbf{A} = J \boldsymbol{\Sigma}^{-1} + (\sigma_\beta^2 \mathbf{I})^{-1}$  and  $\mathbf{b}' = \boldsymbol{\beta}' \boldsymbol{\Sigma}^{-1}$ , where  $\boldsymbol{\beta}$  is the vector sum  $\sum_{j=1}^J \beta_j$ .

*Variance-covariance of regression coefficients ( $\boldsymbol{\Sigma}$ ):*

$$\begin{aligned}
[\boldsymbol{\Sigma} \mid \cdot] & \propto \prod_{j=1}^J [\beta_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}] [\boldsymbol{\Sigma} \mid \mathbf{S}_0, \nu] \\
& \propto \prod_{j=1}^J \mathcal{N}(\beta_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}) \text{Wish}(\boldsymbol{\Sigma} \mid \mathbf{S}_0, \nu) \\
& \propto |\boldsymbol{\Sigma}|^{-\frac{J}{2}} \exp \left\{ -\frac{1}{2} \sum_{j=1}^J (\beta_j - \boldsymbol{\mu}_\beta)' \boldsymbol{\Sigma}^{-1} (\beta_j - \boldsymbol{\mu}_\beta) \right\} \\
& \quad \times |\mathbf{S}_0|^{-\frac{\nu}{2}} |\boldsymbol{\Sigma}|^{-\frac{\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \text{tr}(\mathbf{S}_0 \boldsymbol{\Sigma}^{-1}) \right\} \\
& \propto |\boldsymbol{\Sigma}|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \sum_{j=1}^J \text{tr}((\beta_j - \boldsymbol{\mu}_\beta)' \boldsymbol{\Sigma}^{-1} (\beta_j - \boldsymbol{\mu}_\beta)) + \text{tr}(\mathbf{S}_0 \boldsymbol{\Sigma}^{-1}) \right] \right\} \\
& \propto |\boldsymbol{\Sigma}|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \sum_{j=1}^J \text{tr}((\beta_j - \boldsymbol{\mu}_\beta) (\beta_j - \boldsymbol{\mu}_\beta)' \boldsymbol{\Sigma}^{-1}) + \text{tr}(\mathbf{S}_0 \boldsymbol{\Sigma}^{-1}) \right] \right\} \\
& \propto |\boldsymbol{\Sigma}|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \text{tr} \left( \sum_{j=1}^J ((\beta_j - \boldsymbol{\mu}_\beta) (\beta_j - \boldsymbol{\mu}_\beta)') \boldsymbol{\Sigma}^{-1} + \mathbf{S}_0 \boldsymbol{\Sigma}^{-1} \right) \right] \right\} \\
& \propto |\boldsymbol{\Sigma}|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \text{tr} \left( \sum_{j=1}^J ((\beta_j - \boldsymbol{\mu}_\beta) (\beta_j - \boldsymbol{\mu}_\beta)') + \mathbf{S}_0 \right) \boldsymbol{\Sigma}^{-1} \right] \right\} \\
& = \text{Wish} \left( \left( \sum_{j=1}^J ((\beta_j - \boldsymbol{\mu}_\beta) (\beta_j - \boldsymbol{\mu}_\beta)') + \mathbf{S}_0 \right)^{-1}, J + \nu \right).
\end{aligned}$$