

# GENERALIZED LINEAR MIXED EFFECTS MODEL FOR BINARY DATA USING THE PROBIT LINK

Brian M. Brost

08 March 2016

---

## Description

A mixed effects regression model for binary data using the probit link.

## Implementation

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

## Model statement

Let  $y_{ij}$ , for  $i = 1, \dots, n_j$  and  $j = 1, \dots, J$ , denote observed data that takes on the values  $\{0, 1\}$ , where the index  $i$  denotes replicate observations within group  $j$ , and  $n_j$  is the number of observations in group  $j$ . Furthermore, let  $\mathbf{x}_{ij}$  be a vector of  $p$  covariates (including the intercept) associated with  $y_{ij}$  and  $\boldsymbol{\beta}_j$  be the corresponding vector of coefficients for group  $j$ .

$$\begin{aligned} y_{ij} &\sim \begin{cases} 0, & v_{ij} \leq 0 \\ 1, & v_{ij} > 0 \end{cases} \\ v_{ij} &\sim \mathcal{N}(\mathbf{x}_{ij}'\boldsymbol{\beta}_j, 1) \\ \boldsymbol{\beta}_j &\sim \mathcal{N}(\boldsymbol{\mu}_\beta, \boldsymbol{\Lambda}) \\ \boldsymbol{\mu}_\beta &\sim \mathcal{N}(\mathbf{0}, \sigma_\beta^2 \mathbf{I}) \\ \boldsymbol{\Lambda}^{-1} &\sim \text{Wish}(\mathbf{S}_0^{-1}, \nu) \end{aligned}$$

## Full conditional distributions

*Observation model auxiliary variable ( $v_{ij}$ ):*

$$\begin{aligned} [v_{ij} \mid \cdot] &\propto [y_{ij} \mid v_{ij}] [v_{ij} \mid \mathbf{x}_{ij}'\boldsymbol{\beta}_j, 1] \\ &\propto (1_{\{y_{ij}=0\}} 1_{\{v_{ij} \leq 0\}} + 1_{\{y_{ij}=1\}} 1_{\{v_{ij} > 0\}}) \times \mathcal{N}(v_{ij} \mid \mathbf{x}_{ij}'\boldsymbol{\beta}_j, 1) \\ &= \begin{cases} \mathcal{TN}(\mathbf{x}_{ij}'\boldsymbol{\beta}_j, 1)_{-\infty}^0, & y_{ij} = 0 \\ \mathcal{TN}(\mathbf{x}_{ij}'\boldsymbol{\beta}_j, 1)_0^{\infty}, & y_{ij} = 1 \end{cases} \end{aligned}$$

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

$$\begin{aligned}
[\beta_j \mid \cdot] &\propto [\mathbf{v}_j \mid \mathbf{X}_j \beta_j, \mathbf{1}] [\beta_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Lambda}] \\
&\propto \mathcal{N}(\mathbf{v}_j \mid \mathbf{X}_j \beta_j, \mathbf{1}) \mathcal{N}(\beta_j \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Lambda}) \\
&\propto \exp \left\{ -\frac{1}{2} (\mathbf{v}_j - \mathbf{X}_j \beta_j)' (\mathbf{v}_j - \mathbf{X}_j \beta_j) \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2} (\beta_j - \boldsymbol{\mu}_\beta)' \boldsymbol{\Lambda}^{-1} (\beta_j - \boldsymbol{\mu}_\beta) \right\} \\
&\propto \exp \left\{ -\frac{1}{2} (-2 (\mathbf{v}_j' \mathbf{X}_j) \beta_j + \beta_j' \mathbf{X}_j' \mathbf{X}_j \beta_j) \right\} \\
&\quad \times \exp \left\{ -\frac{1}{2} (-2 (\boldsymbol{\mu}_\beta' \boldsymbol{\Lambda}^{-1}) \beta_j + \beta_j' \boldsymbol{\Lambda}^{-1} \beta_j) \right\} \\
&\propto \exp \left\{ -\frac{1}{2} (-2 (\mathbf{v}_j' \mathbf{X}_j + \boldsymbol{\mu}_\beta' \boldsymbol{\Lambda}^{-1}) \beta_j + \beta_j' (\mathbf{X}_j' \mathbf{X}_j + \boldsymbol{\Lambda}^{-1}) \beta_j) \right\} \\
&= \mathcal{N}(\mathbf{A}^{-1} \mathbf{b}, \mathbf{A}^{-1}),
\end{aligned}$$

where  $\mathbf{A} = \mathbf{X}_j' \mathbf{X}_j + \boldsymbol{\Lambda}^{-1}$ ,  $\mathbf{b}' = \mathbf{v}_j' \mathbf{X}_j + \boldsymbol{\mu}_\beta' \boldsymbol{\Lambda}^{-1}$ ,  $\mathbf{X}_j$  is an  $n_j \times p$  matrix collecting the vectors  $\mathbf{x}_{1,j}, \dots, \mathbf{x}_{n_j,j}$ , and  $\mathbf{v}_j' = \{v_{1,j}, \dots, v_{n_j,j}\}$ .

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

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

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

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

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