# GENERALIZED LINEAR MIXED EFFECTS MODEL

## FOR BINARY DATA USING THE PROBIT LINK

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08 March 2016

## Description

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

#### Implementation

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

#### Model statement

Let  $y_{ij}$ , for  $i = 1, ..., n_j$  and j = 1, ..., 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.

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

### Full conditional distributions

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

$$[v_{ij} \mid \cdot] \propto [y_{ij} \mid v_{ij}] [v_{ij} \mid \mathbf{x}'_{ij}\boldsymbol{\beta}_{j}, \mathbf{1}]$$

$$\propto (1_{\{y_{ij}=0\}}1_{\{v_{ij}\leq 0\}} + 1_{\{y_{ij}=1\}}1_{\{v_{ij}>0\}}) \times \mathcal{N}(v_{ij} \mid \mathbf{x}'_{ij}\boldsymbol{\beta}_{j}, \mathbf{1})$$

$$= \begin{cases} \mathcal{T}\mathcal{N}(\mathbf{x}'_{ij}\boldsymbol{\beta}_{j}, \mathbf{1})_{-\infty}^{0}, & y_{ij} = 0\\ \mathcal{T}\mathcal{N}(\mathbf{x}'_{ij}\boldsymbol{\beta}_{j}, \mathbf{1})_{0}^{\infty}, & y_{ij} = 1 \end{cases}$$

Regression coefficients ( $\beta_i$ ):

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

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  $(\mu_{\beta})$ :

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

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

Variance-covariance of regression coefficients ( $\Lambda$ ):

$$\begin{split} \left[ \boldsymbol{\Lambda} \mid \cdot \right] & \propto \prod_{j=1}^{J} \left[ \boldsymbol{\beta}_{j} \mid \boldsymbol{\mu}_{\beta}, \boldsymbol{\Lambda} \right] \left[ \boldsymbol{\Lambda} \mid \mathbf{S}_{0}, \boldsymbol{\nu} \right] \\ & \propto \prod_{j=1}^{J} \mathcal{N} \left( \boldsymbol{\beta}_{j} \mid \boldsymbol{\mu}_{\beta}, \boldsymbol{\Lambda} \right) \operatorname{Wish} \left( \boldsymbol{\Lambda} \mid \mathbf{S}_{0}, \boldsymbol{\nu} \right) \\ & \propto \left| \boldsymbol{\Lambda} \right|^{-\frac{J}{2}} \exp \left\{ -\frac{1}{2} \sum_{j=1}^{J} \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \boldsymbol{\Lambda}^{-1} \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \right\} \\ & \times \left| \mathbf{S}_{0} \right|^{-\frac{\nu}{2}} \left| \boldsymbol{\Lambda} \right|^{-\frac{\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \operatorname{tr} \left( \mathbf{S}_{0} \boldsymbol{\Lambda}^{-1} \right) \right\} \\ & \propto \left| \boldsymbol{\Lambda} \right|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \sum_{j=1}^{J} \operatorname{tr} \left( \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \boldsymbol{\Lambda}^{-1} \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \right) + \operatorname{tr} \left( \mathbf{S}_{0} \boldsymbol{\Lambda}^{-1} \right) \right] \right\} \\ & \propto \left| \boldsymbol{\Lambda} \right|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \operatorname{tr} \left( \sum_{j=1}^{J} \left( \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \right) \boldsymbol{\Lambda}^{-1} + \mathbf{S}_{0} \boldsymbol{\Lambda}^{-1} \right) \right] \right\} \\ & \propto \left| \boldsymbol{\Lambda} \right|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \operatorname{tr} \left( \sum_{j=1}^{J} \left( \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \right) \boldsymbol{\Lambda}^{-1} + \mathbf{S}_{0} \boldsymbol{\Lambda}^{-1} \right) \right] \right\} \\ & \propto \left| \boldsymbol{\Lambda} \right|^{-\frac{J+\nu-p-1}{2}} \exp \left\{ -\frac{1}{2} \left[ \operatorname{tr} \left( \sum_{j=1}^{J} \left( \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \right) + \mathbf{S}_{0} \right) \boldsymbol{\Lambda}^{-1} \right] \right\} \\ & = \operatorname{Wish} \left( \left( \sum_{j=1}^{J} \left( \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right) \left( \boldsymbol{\beta}_{j} - \boldsymbol{\mu}_{\beta} \right)' \right) + \mathbf{S}_{0} \right)^{-1}, J + \nu \right). \end{split}$$