# SEMIPARAMETRIC PROBIT REGRESSION MODEL FOR BINARY DATA

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## Description

A semiparametric, probit regression model for binary data.

### Implementation

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

#### Model statement

Let  $y_t$ , for t = 1, ..., T, be observed data that take on the values  $\{0,1\}$ . Also let **X** be a design matrix containing covariates for which inference is desired, and **Z** be a design matrix containing some basis expansion. The vectors  $\boldsymbol{\beta}$  and  $\boldsymbol{\alpha}$  are the corresponding 'fixed' and 'random' effects, respectively. Note that  $\mathbf{Z}\boldsymbol{\alpha}$  models non-linear patterns or dependence non-parametrically.

$$y_t \sim \begin{cases} 0, & u_t \leq 0 \\ 1, & u_t > 1 \end{cases}$$

$$u_t \sim \mathcal{N}(\mathbf{x}_t'\boldsymbol{\beta} + \mathbf{z}_t'\boldsymbol{\alpha}, \mathbf{1})$$

$$\boldsymbol{\beta} \sim \mathcal{N}(\mathbf{0}, \sigma_{\beta}^2 \mathbf{I})$$

$$\boldsymbol{\alpha} \sim \mathcal{N}(\mathbf{0}, \sigma_{\alpha}^2 \mathbf{I})$$

Models of this type are typically fit using a large number of basis vectors, more than necessary to approximate non-linear trends or dependence. Regularization (e.g., a ridge penalty) is subsequently conducted to shrink the coefficients  $\alpha$  toward 0 where appropriate. Therefore, the parameter  $\sigma_{\alpha}^2$  must be selected using cross-validation or some model selection criterion. Model-based estimation of  $\sigma_{\alpha}^2$ , i.e.,  $\sigma_{\alpha}^2 \sim \mathrm{IG}(r,q)$ , results in a mixed effects model similar to that implemented by the function glmer (???) in the R package lme4.

#### Full conditional distributions

Observation model auxiliary variable  $(u_t)$ :

$$[u_t|\cdot] \propto [y_t|u_t][u_t]$$

$$\propto (1_{\{y_t=0\}}1_{\{u_t\leq 0\}} + 1_{\{y_t=1\}}1_{\{u_t>0\}}) \times \mathcal{N}(u_t \mid \mathbf{x}_t'\boldsymbol{\beta} + \mathbf{z}_t'\boldsymbol{\alpha}, \mathbf{1})$$

$$= \begin{cases} \mathcal{T}\mathcal{N}(\mathbf{x}_t'\boldsymbol{\beta} + \mathbf{z}_t'\boldsymbol{\alpha}, \mathbf{1})_{-\infty}^0, & y_t = 0 \\ \mathcal{T}\mathcal{N}(\mathbf{x}_t'\boldsymbol{\beta} + \mathbf{z}_t'\boldsymbol{\alpha}, \mathbf{1})_{0}^\infty, & y_t = 1 \end{cases}$$

Fixed effects  $(\beta)$ :

$$\begin{split} [\boldsymbol{\beta}|\cdot] & \propto & [\mathbf{u}|\boldsymbol{\beta},\boldsymbol{\alpha},\sigma^2][\boldsymbol{\beta}] \\ & \propto & \mathcal{N}(\mathbf{u}|\mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\boldsymbol{\alpha},\mathbf{1})\mathcal{N}(\boldsymbol{\beta}|\mathbf{0},\sigma_{\boldsymbol{\beta}}^2\mathbf{I}) \\ & \propto & \exp\left\{-\frac{1}{2}\left(\mathbf{u} - (\mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\boldsymbol{\alpha})\right)'\left(\mathbf{u} - (\mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\boldsymbol{\alpha})\right)\right\} \\ & \exp\left\{-\frac{1}{2}\left(\boldsymbol{\beta} - \mathbf{0}\right)'\left(\sigma_{\boldsymbol{\beta}}^2\mathbf{I}\right)^{-1}\left(\boldsymbol{\beta} - \mathbf{0}\right)\right\} \\ & \propto & \exp\left\{-\frac{1}{2}\left((\mathbf{u} - \mathbf{Z}\boldsymbol{\alpha}) - \mathbf{X}\boldsymbol{\beta}\right)'\left((\mathbf{u} - \mathbf{Z}\boldsymbol{\alpha}) - \mathbf{X}\boldsymbol{\beta}\right)\right\} \\ & \exp\left\{-\frac{1}{2}\left(\boldsymbol{\beta} - \mathbf{0}\right)'\left(\sigma_{\boldsymbol{\beta}}^2\mathbf{I}\right)^{-1}\left(\boldsymbol{\beta} - \mathbf{0}\right)\right\} \\ & \propto & \exp\left\{-\frac{1}{2}\left(-2(\mathbf{u} - \mathbf{Z}\boldsymbol{\alpha})'\mathbf{X}\boldsymbol{\beta} + \boldsymbol{\beta}'\mathbf{X}'\mathbf{X}\boldsymbol{\beta}\right)\right\} \times \\ & \exp\left\{-\frac{1}{2}\left(\boldsymbol{\beta}'\left(\sigma_{\boldsymbol{\beta}}^2\mathbf{I}\right)^{-1}\boldsymbol{\beta}\right)\right\} \\ & \propto & \exp\left\{-\frac{1}{2}\left(-2\left((\mathbf{u} - \mathbf{Z}\boldsymbol{\alpha})'\mathbf{X}\right)\boldsymbol{\beta} + \boldsymbol{\beta}'\left(\mathbf{X}'\mathbf{X} + \left(\sigma_{\boldsymbol{\beta}}^2\mathbf{I}\right)^{-1}\right)\boldsymbol{\beta}\right)\right\} \times \\ & = & \mathcal{N}(\mathbf{A}^{-1}\mathbf{b}, \mathbf{A}^{-1}) \end{split}$$

where  $\mathbf{A} = \mathbf{X}'\mathbf{X} + \left(\sigma_{\beta}^2\mathbf{I}\right)^{-1}$  and  $\mathbf{b}' = (\mathbf{u} - \mathbf{Z}\boldsymbol{\alpha})'\mathbf{X}$ .

Random effects  $(\alpha)$ :

$$\begin{aligned} [\boldsymbol{\alpha}|\cdot] & \propto & [\mathbf{u}|\boldsymbol{\beta},\boldsymbol{\alpha},\sigma^2][\boldsymbol{\alpha}] \\ & \propto & \mathcal{N}(\mathbf{u}|\mathbf{X}\boldsymbol{\beta} + \mathbf{Z}\boldsymbol{\alpha},\mathbf{1})\mathcal{N}(\boldsymbol{\alpha}|\mathbf{0},\sigma_{\alpha}^2\mathbf{I}) \\ & = & \mathcal{N}(\mathbf{A}^{-1}\mathbf{b},\mathbf{A}^{-1}) \end{aligned}$$

where  $\mathbf{A} = \mathbf{Z}'\mathbf{Z} + \left(\sigma_{\alpha}^2\mathbf{I}\right)^{-1}$  and  $\mathbf{b}' = (\mathbf{u} - \mathbf{X}\boldsymbol{\beta})'\mathbf{Z}$ .