

ZERO-INFLATED POISSON GENERALIZED LINEAR MIXED MODEL FOR COUNT DATA

Brian M. Brost

06 April 2016

Description

A zero-inflated generalized linear mixed model with varying coefficients for grouped count data.

Implementation

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

Model statement

Let z_{ij} , for $i = 1, \dots, N$ and $j = 1, \dots, n_i$, denote observed count data (i.e., z_{ij} are integers greater than or equal to 0), where the index j denotes replicate observations (level-1 units) within group i (level-2 units), and n_i is the number of observations in group i . Furthermore, let \mathbf{x}_{ij} be a vector of p covariates (including the intercept) associated with y_{ij} and $\boldsymbol{\beta}_i$ be the corresponding vector of coefficients for group i .

$$\begin{aligned} y_{ij} &\sim \begin{cases} \text{Pois}(\lambda_{ij}), & z_{ij} = 1 \\ 0, & z_{ij} = 0 \end{cases} \\ z_{ij} &\sim \text{Bern}(p_i) \\ \log(\lambda_{ij}) &= \mathbf{x}_{ij}'\boldsymbol{\beta}_i \\ \boldsymbol{\beta}_i &\sim \mathcal{N}(\boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}) \\ \boldsymbol{\mu}_\beta &\sim \mathcal{N}(\mathbf{0}, \sigma_\beta^2 \mathbf{I}) \\ \boldsymbol{\Sigma}^{-1} &\sim \text{Wish}(\mathbf{S}_0^{-1}, \nu) \\ p_i &\sim \text{Beta}(\alpha_1, \alpha_2) \end{aligned}$$

Full conditional distributions

Latent mixture component indicator variable (z_{ij}):

$$\begin{aligned} [z_{ij} \mid \cdot] &\propto [y_{ij} \mid \lambda_{ij}, z_{ij}] [z_{ij} \mid p_i] \\ &\propto \text{Pois}(y_{ij} \mid \lambda_{ij})^{z_{ij}} 1_{\{y_{ij}=0\}}^{1-z_{ij}} p_i^{z_{ij}} (1-p_i)^{1-z_{ij}} \\ &\propto \left(\frac{\lambda_{ij}^{y_{ij}} \exp(-\lambda_{ij})}{y_{ij}!} \right)^{z_{ij}} p_i^{z_{ij}} (1-p_i)^{1-z_{ij}} \\ &\propto (\exp(-\lambda_{ij}))^{z_{ij}} p_i^{z_{ij}} (1-p_i)^{1-z_{ij}} \\ &\propto (p_i \times \exp(-\lambda_{ij}))^{z_{ij}} (1-p_i)^{1-z_{ij}} \\ &= \text{Bern}(\tilde{p}), \end{aligned}$$

where

$$\tilde{p} = \frac{p_i \times \exp(-\lambda_{ij})}{p_i \times \exp(-\lambda_{ij}) + 1 - p_i}.$$

Note that z_{ij} is only updated for instances where $y_{ij} = 0$ ($z_{ij} = 1$ when $y_{ij} > 0$).

Probability associated with the mixture component indicator variables (p_i):

$$\begin{aligned}
[p_i \mid \cdot] &\propto \prod_{j=1}^{n_i} [z_{ij} \mid p_i] [p_i] \\
&\propto \prod_{j=1}^{n_i} p_i^{z_{ij}} (1-p_i)^{1-z_{ij}} p_i^{\alpha_1-1} (1-p_i)^{\alpha_2-1} \\
&\propto p_i^{\sum_{j=1}^{n_i} z_{ij}} (1-p_i)^{n_i - \sum_{j=1}^{n_i} z_{ij}} p_i^{\alpha_1-1} (1-p_i)^{\alpha_2-1} \\
&= \text{Beta} \left(\sum_{j=1}^{n_i} z_{ij} + \alpha_1, n_i - \sum_{j=1}^{n_i} z_{ij} + \alpha_2 \right)
\end{aligned}$$

Regression coefficients (β_i):

$$\begin{aligned}
[\beta_i \mid \cdot] &\propto \prod_{j=1}^{n_i} [y_{ij} \mid \mathbf{x}'_{ij} \beta_i] [\beta_i \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}] \\
&\propto \prod_{j=1}^{n_i} \text{Pois}(y_{ij} \mid \mathbf{x}'_{ij} \beta_i) \mathcal{N}(\beta_i \mid \boldsymbol{\mu}_\beta, \boldsymbol{\Sigma}).
\end{aligned}$$

The update for β_i proceeds using Metropolis-Hastings.

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

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

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

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

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