

# Probability Model For A Multivariate Hierarchical Linear Regression Model

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We consider a two level hierarchical linear regression model in which the data  $y$  are organized into  $J$  distinct groups. Each individual observation  $y_n$  has a predictor row vector  $x_n$  of size  $K$  with  $x_{n,1} = 1$  to unify the notation. Each group  $j$  gets its own coefficient vector  $\beta_j$  of size  $K$ , which is predictable from its group-level predictors  $u_j$  of size  $L$ .

## 1 Multivariate Hierarchical Linear Regression Model

The two level hierarchical linear regression model is described by the following two level regression equations.

$$y_n = \sum_{k=1}^K \beta_{j_n k} x_{nk} + e_n \quad (1)$$

$$\beta_j = \sum_{l=1}^L \gamma_l u_{jl} + \epsilon_j \quad (2)$$

where  $x \in \mathbb{R}^{N \times K}$  is the individual-level predictor matrix;  $u \in \mathbb{R}^{J \times L}$  is the group-level predictor matrix;  $\beta_j \in \mathbb{R}^{1 \times K}$  is the individual-level coefficient vector by group;  $\gamma_l \in \mathbb{R}^{1 \times K}$  is the group-level coefficient vector;  $\epsilon_j \in \mathbb{R}^{1 \times K}$  is the group-level residual vector;  $j_n \in \{1, \dots, J\}$ ;  $n = 1, \dots, N$ ; and  $j = 1, \dots, J$ .

## 2 Probability Model

Assume  $e_n \sim \text{Normal}(0, \sigma)$ , and  $\epsilon_j \sim \text{MultiNormal}(0, \Sigma)$ , then we have

$$y_n \sim \text{Normal} \left( \sum_{k=1}^K \beta_{j_n k} x_{nk}, \sigma \right) \quad (3)$$

$$\beta_j \sim \text{MultiNormal} \left( \sum_{l=1}^L \gamma_l u_{jl}, \Sigma \right) \quad (4)$$

Then we must give priors to  $\sigma$ ,  $\gamma$ , and  $\Sigma$ . We may assume an improper prior for  $\sigma$ , and the following independent weakly informative prior for  $\gamma$ .

$$\gamma_{lk} \sim \text{Normal}(0, 5)$$

where  $l = 1, \dots, L$ , and  $k = 1, \dots, K$ .

We decompose  $\Sigma$  into a scale and a correlation matrix as below

$$\Sigma = \text{diag\_matrix}(\tau) \Omega \text{diag\_matrix}(\tau) \quad (5)$$

where  $\Sigma \in \mathbb{R}^{K \times K}$  is a correlation matrix and  $\tau \in \mathbb{R}^K$  is the vector of coefficient scales that

$$\begin{aligned} \tau_k &= \sqrt{\Sigma_{kk}} \\ \Omega_{ij} &= \frac{\Sigma_{ij}}{\tau_i \tau_j} \end{aligned}$$

We may assume the following priors for  $\tau$  and  $\Omega$ :

$$\begin{aligned} \tau_k &\sim \text{Cauchy}_+(0, 2.5) \\ \Omega &\sim \text{LKJCorr}(2) \end{aligned}$$

where  $k = 1, \dots, K$ .

Putting it all together, we have the probability model for this two level hierarchical linear regression model as:

$$\begin{aligned} y_n &\sim \text{Normal}\left(\sum_{k=1}^K \beta_{j_n k} x_{nk}, \sigma\right) \\ \sigma &\sim \text{Uniform}(0, \infty) \\ \beta_j &\sim \text{MultiNormal}\left(\sum_{l=1}^L \gamma_l u_{jl}, \Sigma\right) \\ \gamma_{lk} &\sim \text{Normal}(0, 5) \\ \Sigma &= \text{diag\_matrix}(\tau) \Omega \text{diag\_matrix}(\tau) \\ \tau_k &\sim \text{Cauchy}_+(0, 2.5) \\ \Omega &\sim \text{LKJCorr}(2) \end{aligned}$$

where  $j_n \in \{1, \dots, J\}$ .