

CheeseData

Rohit Arora

1. Setup

The data consists of quantities sold and the prices for 88 stores selling cheese. In addition we have a store specific display variable indicating if the cheese was on display. We start of with the following model

$$y_i = X_i\beta + Z_i\gamma_i + \epsilon_i$$

where the coefficients β correspond to the fixed effect across stores and γ_i correspond to the store specific random effects. To setup a fully bayesian model we assume the following set of priors on the parameters in the model

$$\begin{aligned} y_i|\beta_i, \gamma_i, \sigma^2, \Sigma &\sim \mathcal{N}(X\beta + Z_i\gamma_i, \sigma^2 I) \\ \beta &\propto 1 \\ \gamma_i|\Sigma &\sim \mathcal{N}(0, \Sigma) \\ \Sigma &\sim \mathcal{IW}(d_0, C_0) \\ \sigma^2 &\propto \frac{1}{\sigma^2} \end{aligned}$$

β is assumed to have a flat prior. γ_i are considered to have a Jeffrey's prior. Other prior's are standard.

2. Gibbs Updates

Below we specify the posterior updates for the parameters in our model. n_s is the number of stores, n_t is the number of observations per store and n is the total number of stores

- posterior for fixed effect β

$$\begin{aligned} p(\beta|\mathbf{y}, \gamma, \sigma^2, \Sigma; \Theta) &\propto p(\beta) \prod_{i=1}^n p(\mathbf{y}_i|\beta, \gamma_i, \sigma^2) \\ &\propto 1 \exp\left(-\frac{1}{2\sigma^2} \sum_{i=1}^{n_s} \sum_{t=1}^{n_t} (y_{it} - x_{it}^\top \beta - z_{it}^\top \gamma_i)^2\right) \\ &= \exp\left(-\frac{1}{2} \sum_{i=1}^{n_s} (\mathbf{y}_i - X_i\beta - Z_i\gamma_i)^\top (\sigma^2 I)^{-1} (\mathbf{y}_i - X_i\beta - Z_i\gamma_i)\right) \\ &= \exp\left(-\frac{1}{2} \left(\beta^\top \left(\sigma^{-2} \sum_{i=1}^{n_s} X_i^\top X_i \right) \beta - 2 \left(\sum_{i=1}^{n_s} \sigma^{-2} (\mathbf{y}_i - Z_i\gamma_i)^\top X_i \right) \beta \right)\right) \\ &\sim \mathcal{N}\left(\left(\sum_{i=1}^{n_s} X_i^\top X_i \right)^{-1} \left(\sum_{i=1}^{n_s} (\mathbf{y}_i - Z_i\gamma_i)^\top X_i \right), \left(\sigma^{-2} \sum_{i=1}^{n_s} X_i^\top X_i \right)^{-1}\right) \end{aligned}$$

- posterior update for random effect γ_i

$$\begin{aligned}
p(\gamma_i | \mathbf{y}_i, \sigma^2, \Sigma; \Theta) &\propto p(\mathbf{y}_i | \gamma_i, \sigma^2) p(\gamma_i | \Sigma) \\
&\propto \exp \left(-\frac{1}{2\sigma^2} \sum_{t=1}^{n_i} (y_{it} - x_{it}^\top \beta - z_{it}^\top \gamma_i)^\top (y_{it} - x_{it}^\top \beta - z_{it}^\top \gamma_i) \right) \exp \left(-\frac{1}{2} \gamma_i^\top \Sigma^{-1} \gamma_i \right) \\
&= \exp \left(-\frac{1}{2\sigma^2} (\mathbf{y}_i - X_i \beta - Z_i \gamma_i)^\top (\mathbf{y}_i - X_i \beta - Z_i \gamma_i) \right) \exp \left(-\frac{1}{2} \gamma_i^\top \Sigma^{-1} \gamma_i \right) \\
&= \exp \left(-\frac{1}{2} \left((\mathbf{y}_i - X_i \beta - Z_i \gamma_i)^\top (\sigma^2 I)^{-1} (\mathbf{y}_i - X_i \beta - Z_i \gamma_i) + \gamma_i^\top \Sigma^{-1} \gamma_i \right) \right) \\
&\propto \exp \left(-\frac{1}{2} \left(-2\sigma^{-2} (\mathbf{y}_i - X_i \beta)^\top Z_i \gamma_i + \sigma^{-2} \gamma_i^\top Z_i^\top Z_i \gamma_i + \gamma_i^\top \Sigma^{-1} \gamma_i \right) \right) \\
&= \exp \left(-\frac{1}{2} \left(\gamma_i^\top (\sigma^{-2} Z_i^\top Z_i + \Sigma^{-1}) \gamma_i - 2\sigma^{-2} (\mathbf{y}_i - X_i \beta)^\top Z_i \gamma_i \right) \right) \\
&\sim \mathcal{N} \left((\sigma^{-2} Z_i^\top Z_i + \Sigma^{-1})^{-1} \sigma^{-2} (\mathbf{y}_i - X_i \beta)^\top Z_i, (\sigma^{-2} Z_i^\top Z_i + \Sigma^{-1})^{-1} \right)
\end{aligned}$$

- posterior update for σ^2

$$\begin{aligned}
p(\sigma^2 | \mathbf{y}, \gamma, \Sigma; \Theta) &\propto p(\mathbf{y} | \gamma, \sigma^2) p(\sigma^2) = p(\sigma^2) \prod_{i=1}^{n_s} p(\mathbf{y}_i | \gamma_i, \sigma^2) \\
&\propto \sigma^{-2 \frac{(n+1)}{2}} \exp \left(-\frac{1}{2\sigma^2} \sum_{i=1}^{n_s} \sum_{t=1}^{n_t} (y_{it} - z_{it}^\top \gamma_i)^\top (y_{it} - z_{it}^\top \gamma_i) \right) \\
&\sim \Pi \left(\frac{n}{2}, \frac{1}{2} \sum_{i=1}^{n_s} \sum_{t=1}^{n_t} (y_{it} - x_{it}^\top \beta - z_{it}^\top \gamma_i)^\top (y_{it} - x_{it}^\top \beta - z_{it}^\top \gamma_i) \right)
\end{aligned}$$

- posterior update for Σ

$$p(\Sigma | \mathbf{y}, \gamma, \sigma^2; \Theta) \propto \mathcal{IW} \left(d_0 + n_s, C_0 + \sum_{i=1}^{n_s} \gamma_i \gamma_i^\top \right)$$