

$$\theta = (\kappa, \beta_\kappa, \sigma^2), \quad P(\theta|y) \propto P(y, \kappa, \beta_\kappa, \sigma^2) = P(y|\kappa, \beta_\kappa, \sigma^2) P\left(\frac{1}{\sigma^2}\right) P(\beta_\kappa|\kappa) P(\kappa) = *$$

$$* = N(y|X_\kappa \beta_\kappa, \sigma^2) \text{Ga}\left(\frac{1}{\sigma^2} | 1, 1\right) N(\beta_\kappa | \phi, 10 \cdot I_{2\kappa+1}) \text{Poi}^+(\kappa | \lambda)$$

$$\propto \left[\prod_{i=1}^n \left(\frac{1}{\sigma^2}\right)^{\frac{1}{2}} \exp\left\{-\frac{1}{2}\left(\frac{1}{\sigma^2}\right)(y_i - [X_\kappa \beta_\kappa]_i)^2\right\} \right] \left[\left(\frac{1}{\sigma^2}\right)^{1-1} \exp\left\{-\left(\frac{1}{\sigma^2}\right)\right\} \right]$$

$$\times \left[\prod_{j=1}^{2\kappa+1} \exp\left\{-\frac{1}{2} \frac{\beta_{\kappa j}^2}{10}\right\} \right] \left[\exp\{-\lambda\} \frac{\lambda^\kappa}{\kappa!} \right]$$

$$= (\sigma^2)^{-\frac{n}{2}} \exp\left\{-\frac{1}{2}(\sigma^2)^{-1} (y - X_\kappa \beta_\kappa)' (y - X_\kappa \beta_\kappa)\right\} \cdot \exp\left\{-(\sigma^2)^{-1}\right\} \exp\left\{-\frac{1}{2} \frac{\beta_\kappa' \beta_\kappa}{10}\right\} \exp\{-\lambda\} \frac{\lambda^\kappa}{\kappa!}$$

MCMC M2 : 2

Complete Conditionals

MD

$$\begin{aligned}
 P(\beta_k | \kappa, \sigma^2, y) &\propto \exp\left\{-\frac{\tau}{2} (y - x_k \beta_k)' (y - x_k \beta_k)\right\} \exp\left\{-\frac{1}{2} \frac{\beta_k' \beta_k}{10}\right\} \\
 &= \exp\left\{-\frac{1}{2} \left[\tau (y'y - y'x_k \beta_k - \beta_k' x_k' y + \beta_k' x_k' x_k \beta_k) + \frac{1}{10} (\beta_k' \beta_k) \right]\right\} \\
 &\propto \exp\left\{-\frac{1}{2} \left[\beta_k' \left(\frac{x_k' x_k}{\sigma^2} + \frac{I_{2k+1}}{10} \right) \beta_k - \beta_k' \left(\frac{x_k' y}{\sigma^2} + \dots \right) \right]\right\} \\
 &\sim N \left(\left(\frac{x_k' x_k}{\sigma^2} + \frac{I_{2k+1}}{10} \right)^{-1} \frac{x_k' y}{\sigma^2}, \left(\frac{x_k' x_k}{\sigma^2} + \frac{I_{2k+1}}{10} \right)^{-1} \right)
 \end{aligned}$$

Used "pattern matching" technique: $(y-m)' \Sigma^{-1} (y-m)$
 $= y' \Sigma^{-1} y - y' \Sigma^{-1} m + \dots$

$$\begin{aligned}
 P(\tau | \kappa, \beta_k, y) &\propto \tau^{\frac{n}{2}} \exp\left\{-\frac{\tau}{2} (y - x_k \beta_k)' (y - x_k \beta_k)\right\} \cdot \exp\{-\tau\} \\
 &= \tau^{\frac{n}{2}} \exp\left\{-\frac{\tau}{2} (y - x_k \beta_k)' (y - x_k \beta_k) - \tau\right\} \\
 &= \tau^{\frac{n}{2}} \exp\left\{-\left[\frac{(y - x_k \beta_k)' (y - x_k \beta_k)}{2} + 1 \right] \tau\right\} \\
 &\sim \text{Ga} \left(\frac{n}{2} + 1, \frac{(y - x_k \beta_k)' (y - x_k \beta_k)}{2} + 1 \right)
 \end{aligned}$$

$$\alpha_{\text{Birth}}(k, k+1) = \min \left\{ 1, \rho(k, k+1) \right\} \quad \text{where}$$

$$\begin{aligned} \rho(k, k+1) &= \frac{\text{target}_{k+1}}{\text{target}_k} \cdot \frac{\text{transition}_{k+1 \rightarrow k}}{\text{transition}_{k \rightarrow k+1}} \cdot \frac{1}{\text{auxiliary}(\mu)} \cdot |\text{Jacobian}| \\ &= \frac{P(k+1, \beta_{k+1}, \sigma^2 | y)}{P(k, \beta_k, \sigma^2 | y)} \cdot \frac{P(k | k+1)}{P(k+1 | k)} \cdot \frac{1}{P(\mu)} \cdot \frac{\partial T}{\partial x_k \partial \mu} \end{aligned}$$

(A) (B) (C) (D)

(A) is defined as the joint posterior probability, calculated in part (1).

(B) is the transition ratio, where probability of $P(k | k+1) = P(k+1 | k) = \frac{1}{2}$, unless $k=1 \Rightarrow P(k+1 | k) = 1$, or $k = k_{\max} \Rightarrow P(k | k+1) = 1$.

(C) is the density of the randomly sampled μ , in the birth move.

(D) is the Jacobian of the transformation $T(x_k, \mu) = x_{k+1}$

$$\alpha_{\text{Death}}(k+1, k) = \min \left\{ 1, \frac{1}{\rho(k, k+1)} \right\}$$

↑ As defined in (3).

MCMC M2: 5

Plots for RJ MCMC

MD

See attachment

MCMC M2: 6

Marginal of Y

Simple form of candidate's formula:

$$P(\beta|y) = \frac{P(y|\beta)P(\beta)}{P(y)} \Rightarrow P(y) = \frac{P(y|\beta)P(\beta)}{P(\beta|y)}$$

In this case:

$$P(\bar{\beta}_k, k | y) = \frac{P(y|\bar{\beta}_k, k) P(\bar{\beta}_k | k) P(k)}{P(y|k)}$$

$$\Rightarrow P(y|k) = \frac{P(y|\bar{\beta}_k, k) P(\bar{\beta}_k | k) P(k)}{P(\bar{\beta}_k, k | y)}$$

For a known k : Can find $P(k) \sim \text{Pois}^+(\lambda)$, and the first two numerator terms

$$\text{yield } P(\bar{\beta}_k | y, k) \sim N \left(\left(2X_k'X_k + \frac{I_{2k+1}}{10} \right)^{-1} \cdot 2X_k'y, \left(2X_k'X_k + \frac{I_{2k+1}}{10} \right)^{-1} \right)$$

$$\text{Therefore } P(y|k) \propto N(\bar{\beta}_k, S_k)$$

\downarrow
 $\bar{\beta}_k$

\downarrow
 S_k

$$\propto |S_k|^{-\frac{1}{2}} P(y|\bar{\beta}_k, k) P(\bar{\beta}_k)$$