distributions_math

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Full Posterior Distribution

$$\begin{split} Y_{i}(t+6) &= \beta_{0i} + \beta_{1i}Y_{i}(t) + \beta_{2i}\Delta_{i1}(t) + \beta_{3i}\Delta_{i2}(t) + \beta_{4i}\Delta_{i3}(t) + \varepsilon_{i}(t) \\ &= \eta_{i}(t) \\ Y &= (Y_{1},Y_{2},...,Y_{n}) \\ \text{Let } \eta &= (\eta_{1},\eta_{2},...,\eta_{n}), \text{ where } \eta_{i} = X_{i}\beta_{i}^{T} \\ Y_{i} &\sim MVN(\eta_{i},\sigma^{2}I_{n_{i}}) \\ f(Y_{i}|B,\mu,\Sigma,\sigma^{2}) &= (2\pi)^{\frac{n_{i}}{2}} \left|\sigma^{2}I_{n_{i}}\right|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}\left(Y_{i} - X_{i}\beta_{i}^{T}\right)(\sigma^{2}I_{n_{i}})^{-1}(Y_{i} - X_{i}\beta_{i}^{T})\right) \\ L(Y|B,\beta,\Sigma,\sigma^{2}) &= \prod_{i=1}^{N} f(Y_{i}|B,\beta,\Sigma,\sigma^{2}) \\ &= \prod_{i=1}^{N} (2\pi)^{-\frac{n_{i}}{2}} \left|\sigma^{2}I_{n_{i}}\right|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(Y_{i} - X_{i}\beta_{i}^{T})^{T}(\sigma^{2}I_{n_{i}})^{-1}(Y_{i} - X_{i}\beta_{i}^{T})^{T}\right) \\ B &= (\beta_{1}^{T},\beta_{2}^{T},...,\beta_{n}^{T})^{T} \\ \beta_{i} &\sim MVN(\mu,\Sigma) \\ f(B|\beta,\Sigma) &= \prod_{i=1}^{n} (2\pi)^{-\frac{5}{2}} \left|\sigma\right|^{-\frac{1}{2}} \exp\left(-\frac{1}{2}(\beta_{i} - \mu)^{T}\Sigma^{-1}(\beta_{i} - \mu)\right) \\ f(\mu) &\propto 1 \\ f(\Sigma) &\propto |\Sigma|^{-(d+1)} \exp\left(-\frac{1}{2}\Sigma^{-1}\right), \text{ where } d = 5 \\ f(\sigma^{2}) &\propto \frac{1}{\sigma^{2}} \end{split}$$

$$\begin{split} &\Longrightarrow f(B,\mu,\Sigma,\sigma^{2}|Y) \propto L(Y|B,\beta,\Sigma,\sigma^{2})f(B,\beta,\Sigma,\sigma^{2}) \\ &= L(Y|B,\beta,\Sigma,\sigma^{2})f(B|\beta,\Sigma)f(\beta)f(\Sigma)f(\sigma^{2}) \\ &= \prod_{i=1}^{N} (2\pi)^{\frac{n_{i}}{2}} \left| \sigma^{2}I_{n_{i}} \right|^{-\frac{1}{2}} \exp\left(-\frac{1}{2} (Y_{i} - X_{i}\beta_{i})^{T} (\sigma^{2}I_{n_{i}})^{-1} (Y_{i} - X_{i}\beta_{i}^{T}) \right) |\Sigma|^{-\frac{n_{i}}{2}} \\ &\exp\left(-\frac{1}{2} (\beta_{i} - \mu)^{T} \Sigma^{-1} (\beta_{i} - \mu) \right) |\Sigma|^{-(d+1)} \exp(-\frac{1}{2} \Sigma^{-1}) \frac{1}{\sigma^{2}} \\ &= \prod_{i=1}^{N} \left| \sigma^{2}I_{n_{i}} \right|^{-\frac{1}{2}} |\Sigma|^{-(d+\frac{n}{2}+1)} \frac{1}{\sigma^{2}} \exp\left[-\frac{1}{2} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} (\sigma^{2}I_{n_{i}})^{-1} (Y_{i} - X_{i}\beta_{i}^{T}) + (\beta_{i} - \mu)^{T} \Sigma^{-1} (\beta_{i} - \mu) \right] \right] \exp(-\frac{1}{2} \operatorname{trace}(\Sigma^{-1})) \\ &= \prod_{i=1}^{N} \sigma^{2n_{i}(-\frac{1}{2})} |\Sigma^{-1}|^{(d+\frac{N}{2}+1)} \sigma^{-2} \exp\left[-\frac{1}{2} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} (\sigma^{2}I_{n_{i}})^{-1} (Y_{i} - X_{i}\beta_{i}^{T}) + (\beta_{i} - \mu)^{T} \Sigma^{-1} (\beta_{i} - \mu) \right] \right] \exp(-\frac{1}{2} \operatorname{trace}(\Sigma^{-1})) \\ &= \prod_{i=1}^{N} \sigma^{-n_{i}} |A|^{d+\frac{N}{2}+1} \sigma^{-2} \exp\left[-\frac{1}{2} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} (\sigma^{2}I_{n_{i}})^{-1} (Y_{i} - X_{i}\beta_{i}^{T}) + (\beta_{i} - \mu)^{T} A(\beta_{i} - \mu) \right] \right] \exp(-\frac{1}{2} \operatorname{trace}A) \\ &= \sigma^{-\sum_{i=1}^{N} n_{i} - 2} |A|^{d+\frac{N}{2}+1} \exp\left[-\frac{1}{2} \sum_{i=1}^{N} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} (\sigma^{2}I_{n_{i}})^{-1} (Y_{i} - X_{i}\beta_{i}^{T}) + (\beta_{i} - \mu)^{T} A(\beta_{i} - \mu) \right] \right] \exp\left[-\frac{1}{2} \operatorname{trace}A \right] \end{split}$$

Conditional distribution of sigma²

$$f(\sigma^{2}|B,\mu,\Sigma,Y) \propto f(B,\mu,\Sigma,\sigma^{2},Y)$$

$$\propto \sigma^{-\sum_{i=1}^{N} n_{i}-2} \exp\left[-\frac{1}{2\sigma^{2}} \sum_{i=1}^{N} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} I_{n_{i}} (Y_{i} - X_{i}\beta_{i}^{T}) \right] \right]$$

$$= (\sigma^{-2})^{\frac{\sum_{i=1}^{N} n_{i}}{2} + 1} \exp\left[-\frac{1}{2} \sum_{i=1}^{N} \left[(Y_{i} - X_{i}\beta_{i}^{T})^{T} (Y_{i} - X_{i}\beta_{i}^{T}) \right] \sigma^{-2} \right]$$

$$= \left(\frac{1}{\sigma^{2}}\right)^{\frac{\sum_{i=1}^{N} n_{i}}{2} + 1} \exp\left[-\frac{1}{2} \sum_{i=1}^{N} \sum_{t=1}^{n_{i}} \left[(y_{it} - x_{it}\beta_{i}^{T})^{2} \right] \frac{1}{\sigma^{2}} \right]$$

Let $W = \sigma^2$. It follows:

$$\begin{split} &\left(\frac{1}{\sigma^2}\right)^{\frac{\sum_{i=1}^{N}n_i}{2}+1} \exp\left[-\frac{1}{2}\sum_{i=1}^{N}\left[(Y_i-X_i\beta_i^T)^T(Y_i-X_i\beta_i^T)\right]\frac{1}{\sigma^2}\right] \\ &=\left(\frac{1}{W}\right)^{\frac{\sum_{i=1}^{N}n_i}{2}+1} \exp\left[-\frac{1}{2}\sum_{i=1}^{N}\left[(Y_i-X_i\beta_i^T)^T(Y_i-X_i\beta_i^T)\right]\frac{1}{W}\right] \\ &\Longrightarrow W \sim \text{inverse Gamma}(\frac{\sum_{i=1}^{N}n_i}{2},\frac{1}{2}\sum_{i=1}^{N}\sum_{t=1}^{n_i}(y_{it}-x_{it}\beta_i^T)^2) \end{split}$$

Conditional Distribution of Big Sigma

$$f(\Sigma^{-1}|B,\mu,\sigma^{2},Y) \propto |A|^{d+\frac{N}{2}+1} \exp\left[-\frac{1}{2} \sum_{i=1}^{N} (\beta_{i} - \mu)^{T} A(\beta_{i} - \mu)\right] \exp(-\frac{1}{2} \operatorname{trace} A)$$

$$= |A|^{d+\frac{N}{2}+1} \exp\left[-\frac{1}{2} \operatorname{trace} \left(A + \sum_{i=1}^{N} (\beta_{i} - \mu)^{T} A(\beta_{i} - \mu)\right)\right]$$

$$= |A|^{d+\frac{N}{2}+1} \exp\left[-\frac{1}{2} \operatorname{trace} \left(A \left(I + \sum_{i=1}^{N} (\beta_{i} - \mu)(\beta_{i} - \mu)^{T}\right)\right)\right]$$

$$V^{-1} = I + \sum_{i=1}^{N} (\beta_{i} - \mu)(\beta_{i} - \mu)^{T}$$

$$\frac{1}{2}(n^{*} - p - 1) = d + \frac{N}{2} + 1$$

$$\implies n^{*} = 3d + N + 2$$

Thus, $A = \Sigma^{-1} \sim \text{Wishart}\left(n^*, I + \sum_{i=1}^{N} (\beta_i - \mu)(\beta_i - \mu)^T\right)$