hurricane_posterior

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Model

$$Y_i(t+6) = \beta_{0,i} + \beta_{1,i}Y_i(t) + \beta_{2,i}\Delta_{i,1}(t) + \beta_{3,i}\Delta_{i,2}(t) + \beta_{4,i}\Delta_{i,3}(t) + \mathbf{X}_i\gamma + \epsilon_i(t)$$

$$\begin{bmatrix} \beta_{0,i} \\ \beta_{1,i} \\ \beta_{2,i} \\ \beta_{3,i} \\ \beta_{4,i} \end{bmatrix} \sim MVN(\begin{bmatrix} \mu_0 \\ \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \end{bmatrix}, \Sigma)$$

$$\epsilon_i \sim N(0, \sigma^2)$$

Priors

1.

$$\begin{bmatrix} \mu_{0,i} \\ \mu_{1,i} \\ \mu_{2,i} \\ \mu_{3,i} \\ \mu_{4,i} \end{bmatrix} \sim MVN(\begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}, V)$$

$$f_{\mu_i}(\mu_i) \propto \det(V)^{\frac{-1}{2}} e^{\frac{-1}{2} \mu_i^{\mathsf{T}} V^{-1} \mu_i} \propto e^{\frac{-1}{2} \mu_i^{\mathsf{T}} V^{-1} \mu_i}$$

2.

$$\Sigma \sim W^{-1}(S, \nu = 5)$$

$$f_{\Sigma^{-1}}(\Sigma) \propto |\Sigma^{-1}|^{n+1} e^{\frac{-1}{2}tr(\Sigma^{-1})}$$

where n is the dimension of dataset... this formula might be wrong meh determinant n dim of of sigma $\,$

Wishart

Due to property of Wishart distribution,

$$\Sigma^{-1} \sim W(S^{-1}, \nu = 5)$$

$$f_{\Sigma^{-1}}(\Sigma^{-1}) = |\Sigma^{-1}|^{\frac{\nu - d - 1}{2}} exp(-\frac{tr(S\Sigma^{-1})}{2}) \propto |\Sigma^{-1}|^{\frac{\nu - 5 - 1}{2}} exp(-\frac{tr(S\Sigma^{-1})}{2})$$

3.

$$\begin{split} \gamma \sim MVN(\begin{bmatrix} 0\\0\\0 \end{bmatrix}, 0.005^2I_3) \\ f_{\gamma}(\gamma) = (3*0.005^2)^{-1/2}exp(-\frac{\gamma^{\mathsf{T}}0.005^2I_3\gamma}{2}) \propto exp(-\frac{400\gamma^{\mathsf{T}}\gamma}{2}) \end{split}$$

4.

$$\sigma \sim Half - Cauchy(0, 10)$$
$$f_{\sigma}(\sigma) = \frac{2 * 10}{\pi(\sigma^2 + 10^2)}$$

By transformation theorem

$$f_{\sigma^2}(\sigma^2) = \frac{2*10}{\pi(\sigma^2 + 10^2)} \frac{1}{2\sigma} \propto \frac{1}{\pi(\sigma^2 + 10^2)\sigma}$$

Likelihood

Because random effects coefficients β_i is normal, $Y_i|\beta_i$ also follows a normal distribution by property of normal distribution.

For each hurricane Y_i

$$Y_{i}|\beta_{i}, \mu_{i}, \sigma^{2}, \Sigma, \gamma \sim MVN(\beta_{0,i} + \beta_{1,i}Y_{i}(t) + \beta_{2,i}\Delta_{i,1}(t) + \beta_{3,i}\Delta_{i,2}(t) + \beta_{4,i}\Delta_{i,3}(t) + \mathbf{X}_{i}\gamma, \sigma^{2}I_{i})$$

$$= MVN(D_{i}\beta_{i} + X_{i}\gamma, \sigma^{2}I_{n_{i}})$$

where

$$Y_{i} = \begin{bmatrix} Y_{i}(t=6) \\ Y_{i}(t=7) \\ \vdots \\ Y_{i}(t=t_{j}+6) \\ \vdots \\ Y_{i}(t=n_{i}+5) \end{bmatrix}_{n_{i}\times 1}$$

$$P_{i}(t) = \begin{bmatrix} 1 & Y_{i}(t) & \Delta_{i,1}(t) & \Delta_{i,2}(t) & \Delta_{i,3}(t) \end{bmatrix}$$

$$= \begin{bmatrix} 1 & Y_{i}(t=0) & \Delta_{i,1}(t=0) & \Delta_{i,2}(t=0) & \Delta_{i,3}(t=0) \\ 1 & Y_{i}(t=1) & \Delta_{i,1}(t=1) & \Delta_{i,2}(t=1) & \Delta_{i,3}(t=1) \end{bmatrix}_{n_{i}\times 5}$$

$$= \begin{bmatrix} 1 & Y_{i}(t=0) & \Delta_{i,1}(t=t_{j}) & \Delta_{i,2}(t=t_{j}) & \Delta_{i,3}(t=t_{j}) \\ \vdots \\ 1 & Y_{i}(t=n_{i}-1) & \Delta_{n,1}(t=n_{i}-1) & \Delta_{n,2}(t=n_{i}-1) & \Delta_{n,3}(t=n_{i}-1) \end{bmatrix}_{n_{i}\times 5}$$

$$\beta_i = \begin{bmatrix} \beta_{0,i} \\ \beta_{1,i} \\ \beta_{2,i} \\ \beta_{3,i} \\ \beta_{4,i} \end{bmatrix}$$

$$X_{i} = \begin{bmatrix} x_{i,1} & x_{i,2} & x_{i,3} \end{bmatrix}_{1 \times 3}$$

$$\gamma = \begin{bmatrix} \gamma_{1} \\ \gamma_{2} \\ \gamma_{3} \end{bmatrix}_{3 \times 1}$$

Likelihood for the ith hurricane is

$$f(Y_i|\beta_i, \mu_i, \sigma^2, \Sigma, \gamma) = det(\sigma^2 I_{n_i})^{-1/2} exp(-\frac{(Y_i - D_i\beta_i - X_i\gamma)^{\mathsf{T}}(Y_i - D_i\beta_i - X_i\gamma)}{2\sigma^2})$$

To calculate the joint likelihood for $Y = \begin{bmatrix} Y_1 & Y_2 \dots Y_i \dots Y_H \end{bmatrix}^{\mathsf{T}}$, we denote total number of observations for all hurricanes as $N = \sum_{i=0}^H n_i$ where n_i is the total number of observation for the ith hurricane and H is the total number of hurricanes.

All random effects coefficients β_i in

$$B = \begin{bmatrix} \beta_1 & \beta_2 \dots & \beta_i & \dots \beta_H \end{bmatrix}$$

$$= \begin{bmatrix} \beta_{0,1} & \beta_{0,2} & \dots & \beta_{0,i} & \dots & \beta_{0,H} \\ \beta_{1,1} & \beta_{1,2} & \dots & \beta_{1,i} & \dots & \beta_{1,H} \\ \beta_{2,1} & \beta_{2,2} & \dots & \beta_{2,i} & \dots & \beta_{2,H} \\ \beta_{3,1} & \beta_{3,2} & \dots & \beta_{3,i} & \dots & \beta_{3,H} \\ \beta_{4,1} & \beta_{4,2} & \dots & \beta_{4,i} & \dots & \beta_{4,H} \end{bmatrix}_{5 \times H}$$

Design matrix for random effects for all hurricanes are in D. $D = \begin{bmatrix} D_1(t) \\ D_2(t) \\ \vdots \\ D_i(t) \\ \vdots \\ D_H(t) \end{bmatrix}_{N \times 5}$

Due to independence of each hurricane, the joint likelihood is

$$\begin{split} L_Y(B,\mu,\sigma^2,\Sigma,\gamma) &= \prod_{i=1}^H L_{Y_i}(\beta_i,\mu,\sigma^2,\Sigma,\gamma) \\ &= \prod_{i=1}^H \det(\sigma^2 I_{n_i})^{-1/2} exp(-\frac{(Y_i - D_i\beta_i - X_i\gamma)^\intercal(Y_i - D_i\beta_i - X_i\gamma)}{2\sigma^2}) \\ &= \frac{1}{\sigma^N} exp(-\frac{(Y - DB - X\gamma)^\intercal(Y - DB - X\gamma)}{2\sigma^2}) \end{split}$$

Posterior

By Baye's Rule

$$f(B, \mu, \sigma^2, \Sigma, \gamma | Y) \propto f(Y | B, \mu, \sigma^2, \Sigma, \gamma) \times f(B | \mu, \Sigma) \times f(\mu) \times f(\Sigma) \times f(\sigma^2) \times f(\gamma)$$

where

$$\mu = \begin{bmatrix} \mu_0 \\ \mu_1 \\ \mu_2 \\ \mu_3 \\ \mu_4 \end{bmatrix}_{5 \times 1}$$

$$f(B|\mu, \Sigma) = \prod_{i=1}^{H} f(\beta_i|\mu, \Sigma)$$

$$= \prod_{i=1}^{H} det(\Sigma)^{-1/2} exp(-\frac{(\beta_i - \mu)^{\mathsf{T}} \Sigma^{-1} (\beta_i - \mu)}{2})$$

$$= det(A)^{H/2} \prod_{i=1}^{H} exp(-\frac{(\beta_i - \mu)^{\mathsf{T}} A (\beta_i - \mu)}{2})$$

where $A = \Sigma^{-1}$

$$f(\mu) = det(V)^{\frac{-1}{2}} exp(-\frac{\mu^{\mathsf{T}}V^{-1}\mu}{2})$$

We'll only use $f_{\Sigma^{-1}}$ because only Σ^{-1} shows up in the likelihood equation. We denote $A = \Sigma^{-1}$ in the posterior.

$$\begin{split} f_{\Sigma^{-1}}(\Sigma^{-1}) &\propto |\Sigma^{-1}|^{\frac{\nu-5-1}{2}} exp(-\frac{tr(S\Sigma^{-1})}{2}) \\ f_{\sigma^2}(\sigma^2) &= \frac{2*10}{\pi(\sigma^2+10^2)} \frac{1}{2\sigma} \propto \frac{1}{\pi(\sigma^2+10^2)\sigma} \\ f_{\gamma}(\gamma) &= exp(-\frac{400\gamma^{\mathsf{T}}\gamma}{2}) \end{split}$$

Final posterior

$$\begin{split} f(B,\mu,\sigma^2,\Sigma,\gamma|Y) &\propto f(Y|B,\mu,\sigma^2,\Sigma,\gamma) \times f(B|\mu,\Sigma) \times f(\mu) \times f(\Sigma^{-1}) \times f(\sigma^2) \times f(\gamma) \\ &= f(Y|B,\mu,\sigma^2,\Sigma,\gamma) \times f(B|\mu,A) \times f(\mu) \times f(A) \times f(\sigma^2) \times f(\gamma) \\ &= \prod_{i=1}^H \det(\sigma^2 I_{n_i})^{-1/2} exp(-\frac{(Y_i - D_i\beta_i - X_i\gamma)^\intercal(Y_i - D_i\beta_i - X_i\gamma)}{2\sigma^2}) \times \\ &\det(A)^{H/2} \prod_{i=1}^H exp(-\frac{(\beta_i - \mu)^\intercal A(\beta_i - \mu)}{2}) \times \\ &\det(V)^{\frac{-1}{2}} exp(-\frac{\mu^\intercal V^{-1}\mu}{2}) \times \\ &|A|^{\frac{\nu-5-1}{2}} exp(-\frac{tr(SA)}{2}) \times \\ &\frac{1}{\pi(\sigma^2 + 10^2)\sigma} \times \\ &exp(-\frac{400\gamma^\intercal \gamma}{2}) \end{split}$$

Conditional Posterior

For B,

$$f(B|\mu, \sigma^{2}, \Sigma, \gamma, Y) \propto \prod_{i=1}^{H} det(\sigma^{2}I_{n_{i}})^{-1/2} exp\left(-\frac{(Y_{i} - D_{i}\beta_{i} - X_{i}\gamma)^{\intercal}(Y_{i} - D_{i}\beta_{i} - X_{i}\gamma)}{2\sigma^{2}}\right) \times$$

$$det(A)^{H/2} \prod_{i=1}^{H} exp\left(-\frac{(\beta_{i} - \mu)^{\intercal}A(\beta_{i} - \mu)}{2}\right)$$

$$\propto \prod_{i=1}^{H} exp\left(-\frac{(Y_{i} - D_{i}\beta_{i} - X_{i}\gamma)^{\intercal}(Y_{i} - D_{i}\beta_{i} - X_{i}\gamma)}{2\sigma^{2}}\right) exp\left(-\frac{(\beta_{i} - \mu)^{\intercal}A(\beta_{i} - \mu)}{2}\right)$$

$$= exp\left(\sum_{i=1}^{H}\right)$$

For μ ,

$$\begin{split} f(\mu|B,\sigma^2,\Sigma,\gamma,Y) \propto & \prod_{i=1}^{H} exp(-\frac{(\beta_i-\mu)^\intercal A(\beta_i-\mu)}{2}) exp(-\frac{\mu^\intercal V^{-1}\mu}{2}) \\ = & exp(\sum_{i=1}^{H}) \end{split}$$

For σ^2 ,

$$f(\sigma^2|B,\mu,\Sigma,\gamma,Y) \propto \prod_{i=1}^{H} det(\sigma^2 I_{n_i})^{-1/2} exp(-\frac{(Y_i - D_i\beta_i - X_i\gamma)^{\intercal}(Y_i - D_i\beta_i - X_i\gamma)}{2\sigma^2}) \times \frac{1}{\pi(\sigma^2 + 10^2)\sigma}$$

$$= exp(\sum_{i=1}^{H})$$

For $A = \Sigma^{-1}$,

$$f(\Sigma^{-1}|B,\mu,\sigma^2,\gamma,Y) \propto det(A)^{H/2} \prod_{i=1}^{H} exp(-\frac{(\beta_i - \mu)^{\mathsf{T}} A(\beta_i - \mu)}{2}) \times |A|^{\frac{\nu - 5 - 1}{2}} exp(-\frac{tr(SA)}{2})$$

$$= exp(\sum_{i=1}^{H})$$

For γ ,

$$\begin{split} f(\gamma|B,\mu,\sigma^2,\Sigma,Y) \propto &= \prod_{i=1}^H exp(-\frac{(Y_i - D_i\beta_i - X_i\gamma)^\intercal(Y_i - D_i\beta_i - X_i\gamma)}{2\sigma^2}) \times \\ &exp(-\frac{400\gamma^\intercal\gamma}{2}) \\ &= &exp(\sum_{i=1}^H) \end{split}$$