derive posterior distribution

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The suggested Bayesian model

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

where $Y_i(t)$ the wind speed at time t (i.e. 6 hours earlier), $\Delta_{i,1}(t)$, $\Delta_{i,2}(t)$ and $\Delta_{i,3}(t)$ are the changes of latitude, longitude and wind speed between t and t+6, and $\epsilon_{i,t}$ follows a normal distributions with mean zero and variance σ^2 , independent across t.

In the model, $\beta_i = (\beta_{0,i}, \beta_{1,i}, ..., \beta_{7,i})$ are the random coefficients associated the *i*th hurricane, we assume that

$$\boldsymbol{\beta}_i \sim N(\boldsymbol{\beta}, \boldsymbol{\Sigma})$$

follows a multivariate normal distributions with mean β and covariance matrix Σ .

We assume the following non-informative or weak prior distributions for σ^2 , β and Σ .

$$P(\sigma^2) \propto \frac{1}{\sigma^2}; \quad P(\boldsymbol{\beta}) \propto 1; \quad P(\Sigma^{-1}) \propto |\Sigma|^{-(d+1)} \exp(-\frac{1}{2}\Sigma^{-1})$$

d is dimension of β .

Let $\mathbf{B} = (\boldsymbol{\beta}_1^\top, ..., \boldsymbol{\beta}_n^\top)^\top$, derive the posterior distribution of the parameters $\Theta = (\mathbf{B}^\top, \boldsymbol{\beta}^\top, \sigma^2, \Sigma)$. Note from given Bayesian model, let

$$\epsilon_{i}(t) = Y_{i}(t+6) - \left(\beta_{0,i} + \beta_{1,i}Y_{i}(t) + \beta_{2,i}\Delta_{i,1}(t) + \beta_{3,i}\Delta_{i,2}(t) + \beta_{4,i}\Delta_{i,3}(t)\right) \stackrel{i.i.d}{\sim} N(0,\sigma^{2})$$
or
$$Y_{i}(t+6) \sim N(\boldsymbol{X}_{i}(t)\boldsymbol{\beta}_{i}^{\top},\sigma^{2})$$

where $\boldsymbol{X}_{i}(t)=(1,Y_{i}(t),\Delta_{i,1}(t),\Delta_{i,2}(t),\Delta_{i,3}(t))$, and $\boldsymbol{\beta}_{i}=(\beta_{0,i},\beta_{1,i},\beta_{2,i},\beta_{3,i},\beta_{4,i})$. Therefore, the wind speed of i^{th} hurricane at time t follows the normal distribution with the pdf below

$$f_{Y_i(t+6)}(y_i(t+6) \mid \boldsymbol{X}_i(t), \boldsymbol{\beta}_i, \sigma^2) = \frac{1}{\sqrt{2\pi}\sigma} \exp\left\{-\frac{1}{2\sigma^2} \left(y_i(t+6) - \boldsymbol{X}_i(t)\boldsymbol{\beta}_i^{\top}\right)^2\right\}$$

Therefore, the conditional distribution of Y_i , the wind speed of i^{th} hurricane follows the multivariate normal distribution below, (since $Y_i(t)$'s are independent across t)

$$(\boldsymbol{Y}_i \mid \boldsymbol{X}_i, \boldsymbol{\beta}_i, \sigma^2) \sim \mathcal{N}(\boldsymbol{X}_i \boldsymbol{\beta}_i^{\top}, \sigma^2 I)$$

where Y_i is an m_i -dimensional vector and \boldsymbol{X}_i is a $m_i \times d$ matrix.

Hence, the joint likelihood function of all i's hurricanes can be expresses as

$$L_Y(\mathbf{B}, \sigma^2) = \prod_{i=1}^n \left\{ \det(2\pi\sigma^2 I)^{-\frac{1}{2}} \exp\left(-\frac{1}{2} (\boldsymbol{Y}_i - \boldsymbol{X}_i \boldsymbol{\beta}_i^\top)^\top (\sigma^2 I)^{-1} (\boldsymbol{Y}_i - \boldsymbol{X}_i \boldsymbol{\beta}_i^\top)\right) \right\}$$

where I is an identical matrix with dimension consistent with Y_i .

From Bayesian theorem, the posterior distribution for Θ is

$$\pi(\mathbf{\Theta}|\mathbf{Y}) = \pi(\mathbf{B}, \boldsymbol{\beta}, \sigma^2, \boldsymbol{\Sigma} \mid \mathbf{Y}) \propto L_Y(\mathbf{B}, \sigma^2) \times \pi(\mathbf{B} \mid \boldsymbol{\beta}, \boldsymbol{\Sigma}) \times \pi(\boldsymbol{\beta}) \times \pi(\sigma^2) \times \pi(\boldsymbol{\Sigma}),$$

where $\pi(\mathbf{B} \mid \boldsymbol{\beta}, \Sigma)$ is the joint multivariate normal density of β , since

$$\beta_i \sim N(\beta, \Sigma)$$

Therefore

$$\pi(\mathbf{B}\mid\boldsymbol{\beta},\boldsymbol{\Sigma}) = \prod_{i=1}^n \Big\{ \det(2\pi\boldsymbol{\Sigma})^{-\frac{1}{2}} \exp(-\frac{1}{2}(\boldsymbol{\beta}_i - \boldsymbol{\beta})\boldsymbol{\Sigma}^{-1}(\boldsymbol{\beta}_i - \boldsymbol{\beta})^\top) \Big\}.$$

So we have the following posterior distribution:

$$\begin{split} \pi(\mathbf{B}, \boldsymbol{\beta}, \sigma^2, \boldsymbol{\Sigma} \mid \boldsymbol{Y}) &\propto \prod_{i=1}^n \left\{ (2\pi\sigma^2)^{-m_i/2} \exp\left\{ -\frac{1}{2} (\boldsymbol{Y}_i - \boldsymbol{X}_i \boldsymbol{\beta}_i^\top)^\top (\sigma^2 I)^{-1} (\boldsymbol{Y}_i - \boldsymbol{X}_i \boldsymbol{\beta}_i^\top) \right\} \right\} \\ &\times \prod_{i=1}^n \left\{ \det(2\pi\boldsymbol{\Sigma})^{-\frac{1}{2}} \exp\left\{ -\frac{1}{2} (\boldsymbol{\beta}_i - \boldsymbol{\beta}) \boldsymbol{\Sigma}^{-1} (\boldsymbol{\beta}_i - \boldsymbol{\beta})^\top \right\} \right\} \times \frac{1}{\sigma^2} \times |\boldsymbol{\Sigma}|^{-(d+1)} \exp\left\{ -\frac{1}{2} \boldsymbol{\Sigma}^{-1} \right\} \end{split}$$