

AR(1) plus noise model

data: $\{y_1, \dots, y_m\}$

model: $y_t = \theta_t + \epsilon_t \quad \epsilon_t \sim N(0, \sigma^2)$
 $\theta_t = \alpha + \beta \theta_{t-1} + w_t \quad w_t \sim N(0, \tau^2)$
 $\theta_1 \sim N(a_1, R_1) \quad \gamma = (\alpha, \beta, \tau^2, \sigma^2)$

- Derive
- i) $P(\theta_t | \theta_{-t}, y_{1:m}, \gamma) \quad t=2, \dots, m-1$
 - ii) $P(\theta_1 | \theta_{2:m}, y_{1:m}, \gamma)$
 - iii) $P(\theta_m | \theta_{1:(m-1)}, y_{1:m}, \gamma)$

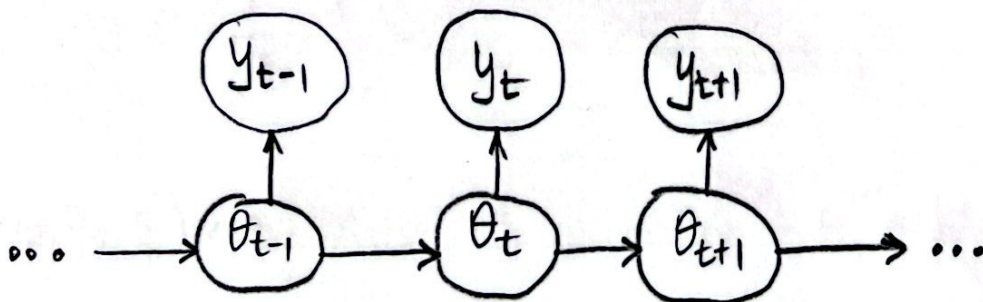
Note: The dynamic/Markovian structure of the evolution of θ_t allows us to write

$t=2, \dots, m-1$

$$P(\theta_t | \theta_{-t}, y_{1:m}, \gamma) = P(\theta_t | \theta_{t-1}, \theta_{t+1}, y_t, \gamma)$$

$$P(\theta_1 | \theta_{2:n}, y_{1:n}, \gamma) = P(\theta_1 | \theta_2, y_1, \gamma)$$

$$P(\theta_m | \theta_{1:(m-1)}, y_{1:n}, \gamma) = P(\theta_m | \theta_{m-1}, y_m, \gamma)$$



Simulate data

$\theta_0 = 0 \quad \sigma = 1 \quad \tau = 0.25 \quad \alpha = 0 \quad \beta = 1$

$m = 100 \Rightarrow \{y_1, \dots, y_m\}$

Run Gibbs sampler for $\theta_{1:m}$

with $a_1 = 0 \quad R_1 = 1$

burn-in 1000
draws 1000
step 1

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For $t=2,3,\dots,m-1$

$$\begin{aligned}
P(\theta_t | \theta_{t-1}, \theta_{t+1}, y_t, \gamma) &\propto P(y_t | \theta_t, \gamma) P(\theta_t | \theta_{t-1}, \gamma) P(\theta_{t+1} | \theta_t, \gamma) \\
&\propto \exp\left\{-\frac{1}{2\sigma^2} (y_t^2 - 2y_t\theta_t + \theta_t^2)\right\} \exp\left\{-\frac{1}{2\tau^2} (\theta_t^2 - 2\theta_t(\alpha + \beta\theta_{t-1}) + (\alpha + \beta\theta_{t-1})^2)\right\} \\
&\times \exp\left\{-\frac{1}{2\tau^2} ((\theta_{t+1} - \alpha)^2 - 2\beta\theta_t(\theta_{t+1} - \alpha) + \beta^2\theta_t^2)\right\} \\
&\propto \exp\left\{-\frac{1}{2} \left[\theta_t^2 \left(\frac{1}{\sigma^2} + \frac{1}{\tau^2} + \frac{\beta^2}{\tau^2} \right) - 2\theta_t \left(\frac{y_t}{\sigma^2} + \frac{\alpha + \beta\theta_{t-1}}{\tau^2} + \frac{\beta(\theta_{t+1} - \alpha)}{\tau^2} \right) \right]\right\}
\end{aligned}$$

$$\Rightarrow (\theta_t | \theta_{t-1}, \theta_{t+1}, y_t, \gamma) \sim N(m_t, C_t) \quad C_t^{-1} = \sigma^{-2} + (1 + \beta^2)\tau^{-2}$$

$$m_t = C_t \left(\frac{y_t}{\sigma^2} + \frac{\alpha(1-\beta) + \beta(\theta_{t+1} + \theta_{t-1})}{\tau^2} \right)$$

$t=1$

$$\begin{aligned}
P(\theta_1 | \theta_2, y_1, \gamma) &\propto P(y_1 | \theta_1, \gamma) P(\theta_2 | \theta_1, \gamma) P(\theta_1) \\
&\propto \exp\left\{-\frac{1}{2\sigma^2} (y_1^2 - 2\theta_1 y_1 + \theta_1^2)\right\} \exp\left\{-\frac{1}{2R_1} (\theta_2^2 - 2\theta_1 a_1 + a_1^2)\right\} \\
&\times \exp\left\{-\frac{1}{2\tau^2} ((\theta_2 - \alpha)^2 - 2\beta\theta_1(\theta_2 - \alpha) + \beta^2\theta_1^2)\right\} \\
&\propto \exp\left\{-\frac{1}{2} \left[\theta_1^2 \left(\frac{1}{\sigma^2} + \frac{1}{R_1} + \frac{\beta^2}{\tau^2} \right) - 2\theta_1 \left(\frac{y_1}{\sigma^2} + \frac{a_1}{R_1} + \frac{\beta(\theta_2 - \alpha)}{\tau^2} \right) \right]\right\}
\end{aligned}$$

$$\Rightarrow (\theta_1 | \theta_2, y_1, \gamma) \sim N(m_1, C_1) \quad C_1^{-1} = \sigma^{-2} + R_1^{-1} + \beta^2\tau^{-2}$$

$$m_1 = C_1 \left(\frac{y_1}{\sigma^2} + \frac{a_1}{R_1} + \frac{\beta(\theta_2 - \alpha)}{\tau^2} \right)$$

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t=m

$$P(\theta_m | \theta_{m-1}, y_m, \delta) \propto P(y_m | \theta_m, \delta) P(\theta_m | \theta_{m-1}, \delta)$$

$$\propto \exp\left\{-\frac{1}{2\sigma^2}(y_m^2 - 2\theta_m y_m + \theta_m^2)\right\} \exp\left\{-\frac{1}{2\tau^2}\left(\theta_m^2 - 2\theta_m(\alpha + \beta\theta_{m-1}) + (\alpha + \beta\theta_{m-1})^2\right)\right\}$$

$$\propto \exp\left\{-\frac{1}{2}\left[\theta_m^2\left(\frac{1}{\sigma^2} + \frac{1}{\tau^2}\right) - 2\theta_m\left(\frac{y_m}{\sigma^2} + \frac{\alpha + \beta\theta_{m-1}}{\tau^2}\right)\right]\right\}$$

$$\Rightarrow (\theta_m | \theta_{m-1}, y_m, \delta) \sim N(m_m, C_m)$$

$$C_m^{-1} = \sigma^{-2} + \tau^{-2}$$

$$m_m = C_m \left(\frac{y_m}{\sigma^2} + \frac{\alpha + \beta\theta_{m-1}}{\tau^2} \right)$$

Assuming $P(\alpha, \beta, \tau^2, \sigma^2) = P(\alpha)P(\beta)P(\tau^2)P(\sigma^2)$

where

$$\delta = \begin{pmatrix} \alpha \\ \beta \end{pmatrix} \sim N(d_0, D_0)$$

$$\tau^2 \sim IG(a_\tau, b_\tau) \quad \sigma^2 \sim IG(a_\sigma, b_\sigma)$$

we now derive the fixed parameters' full conditionals

$$P(\sigma^2 | y_{1:m}, \theta_{1:m}, \alpha, \beta, \tau^2) \propto (\sigma^2)^{-(a_\sigma+1)} e^{-\frac{b_\sigma}{\sigma^2}} \times (\sigma^2)^{-\left(\frac{m}{2}+1\right)} e^{-\frac{\sum (y_t - \theta_t)^2 / 2}{\sigma^2}}$$

$$\sim IG\left(a_\sigma + \frac{m}{2}; b_\sigma + \sum_{t=1}^m (y_t - \theta_t)^2 / 2\right)$$

$$P(\tau^2 | y_{1:m}, \theta_{1:m}, \alpha, \beta, \sigma^2) \propto (\tau^2)^{-(a_\tau+1)} e^{-b_\tau / \tau^2} \times (\tau^2)^{-\left(\frac{m-1}{2}+1\right)} e^{-\frac{\sum_{t=2}^m (\theta_t - \alpha - \beta \theta_{t-1})^2 / 2}{\tau^2}}$$

$$\sim IG\left(a_\tau + \frac{m-1}{2}; b_\tau + \frac{\sum_{t=2}^m (\theta_t - \alpha - \beta \theta_{t-1})^2}{2}\right)$$

$$P(\delta | y_{1:m}, \theta_{1:m}, \tau^2, \sigma^2) \propto \exp\left\{-\frac{1}{2} (\delta - d_0)^T D_0^{-1} (\delta - d_0)\right\}$$

$$\times \exp\left\{-\frac{1}{2\tau^2} (\tilde{y} - X\delta)^T (\tilde{y} - X\delta)\right\}$$

$$\tilde{y} = (\theta_2, \dots, \theta_m)^T$$
$$X = \begin{bmatrix} 1 & \theta_1 \\ 1 & \theta_2 \\ \vdots & \vdots \\ 1 & \theta_{m-1} \end{bmatrix}$$

$$\propto \exp\left\{-\frac{1}{2} \left[\delta^T (D_0^{-1} + X^T X / \tau^2) - 2\delta^T (D_0^{-1} d_0 + \frac{X^T \tilde{y}}{\tau^2}) \right]\right\}$$

$$\Rightarrow N\left[\left(D_0^{-1} + X^T X / \tau^2\right)^{-1} \left(D_0^{-1} d_0 + \frac{X^T \tilde{y}}{\tau^2}\right), \left(D_0^{-1} + \frac{X^T X}{\tau^2}\right)\right]$$

Local Level NDLM

$$y_t = \theta_t + \epsilon_t \quad \epsilon_t \sim N(0, \sigma^2)$$

$$\theta_t = \theta_{t-1} + w_t \quad w_t \sim N(0, \tau^2)$$

$$\theta_1 \sim N(a_1, R_1)$$

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For $t=2, 3, \dots, m-1$

$$P(\theta_t | \theta_{t-1}, \theta_{t+1}, y_t, \gamma) \propto P(y_t | \theta_t, \gamma) P(\theta_t | \theta_{t-1}, \gamma) P(\theta_{t+1} | \theta_t, \gamma)$$

$$\propto \exp\left\{-\frac{1}{2\sigma^2}(y_t^2 - 2y_t\theta_t + \theta_t^2)\right\} \exp\left\{-\frac{1}{2\tau^2}(\theta_t^2 - 2\theta_t\theta_{t-1} + \theta_{t-1}^2)\right\}$$

$$\times \exp\left\{-\frac{1}{2\tau^2}(\theta_{t+1}^2 - 2\theta_{t+1}\theta_t + \theta_t^2)\right\}$$

$$\propto \exp\left\{-\frac{1}{2}\left[\theta_t^2\left(\frac{1}{\sigma^2} + \frac{2}{\tau^2}\right) - 2\theta_t\left(\frac{y_t}{\sigma^2} + \frac{\theta_{t-1} + \theta_{t+1}}{2}\right)\right]\right\}$$

$$\Rightarrow (\theta_t | \theta_{t-1}, \theta_{t+1}, y_t, \gamma) \sim N(m_t, C_t)$$

$$C_t^{-1} = \sigma^{-2} + \tau^{-2}$$

$$m_t = C_t \left(\frac{y_t}{\sigma^2} + \frac{\theta_{t-1} + \theta_{t+1}}{\tau^2} \right)$$

t=1

$$P(\theta_1 | \theta_2, y_1, \gamma) \propto P(y_1 | \theta_1, \gamma) P(\theta_2 | \theta_1, \gamma) P(\theta_1)$$

$$\propto \exp\left\{-\frac{1}{2\sigma^2}(y_1^2 - 2\theta_1 y_1 + \theta_1^2)\right\} \exp\left\{-\frac{1}{2\tau^2}(\theta_2^2 - 2\theta_1\theta_2 + \theta_1^2)\right\}$$

$$\times \exp\left\{-\frac{1}{2R_1}(\theta_1^2 - 2\theta_1 a_1 + a_1^2)\right\}$$

$$\propto \exp\left\{-\frac{1}{2}\left[\theta_1^2\left(\frac{1}{\sigma^2} + \frac{1}{\tau^2} + \frac{1}{R_1}\right) - 2\theta_1\left(\frac{y_1}{\sigma^2} + \frac{\theta_2}{\tau^2} + \frac{a_1}{R_1}\right)\right]\right\}$$

$$\Rightarrow (\theta_1 | \theta_2, y_1, \gamma) \sim N(m_1, C_1)$$

$$C_1^{-1} = \sigma^{-2} + \tau^{-2} + R_1^{-1}$$

$$m_1 = C_1 \left(\frac{y_1}{\sigma^2} + \frac{\theta_2}{\tau^2} + \frac{a_1}{R_1} \right)$$

t=2 Similarly,

$$(\theta_n | \theta_{n-1}, y_n, \gamma) \sim N(m_n, C_n)$$

$$C_n^{-1} = \sigma^{-2} + \tau^{-2}$$

$$m_n = C_n \left(\frac{y_n}{\sigma^2} + \frac{\theta_{n-1}}{\tau^2} \right)$$