

Untitled

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$$p(\mathbf{y}, \mathbf{s} | \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2, \xi) = p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2) p(\mathbf{s} | \xi)$$

where

$$\begin{aligned} p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2) &= p(y_1 | s_1 = s_1, \mu_{s_1}, \sigma_{s_1}^2) \times p(y_2 | s_2 = s_2, \mu_{s_2}, \sigma_{s_2}^2) \times \dots \times p(y_T | s_T = s_T, \mu_{s_T}, \sigma_{s_T}^2) \\ &= \prod_{k=1}^K \prod_{t:s_t=k} p(y_t | s_t = k, \mu_k, \sigma_k^2) \\ &= \prod_{k=1}^K \prod_{t:s_t=k} \frac{1}{\sqrt{2\pi\sigma_k^2}} \exp \left\{ -\frac{1}{2\sigma_k^2} (y_t - \mu_k)^2 \right\} \\ &\propto \prod_{k=1}^K \sigma_k^{2(-\frac{N_k}{2})} \exp \left\{ -\frac{1}{2\sigma_k^2} \sum_{t:s_t=k} (y_t - \mu_k)^2 \right\} \end{aligned}$$

where N_k is the number of observations in state k .

The full conditional distribution for μ_1, \dots, μ_k is therefore:

$$\begin{aligned} \pi(\mu_1, \dots, \mu_K | \mathbf{y}, \mathbf{s}, \sigma_1^2, \dots, \sigma_K^2, \xi) &\propto p(\mathbf{y}, \mathbf{s} | \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2, \xi) \pi(\mu_1, \dots, \mu_K) \\ &= p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2) p(\mathbf{s} | \xi) \pi(\mu_1, \dots, \mu_K) \\ &\propto p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2) \pi(\mu_1, \dots, \mu_K) \\ &= p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_k, \sigma_1^2, \dots, \sigma_K^2) \times \prod_{k=1}^K \pi(\mu_k) \end{aligned}$$

$$\begin{aligned}
& \propto \prod_{k=1}^K \sigma_k^{2(-\frac{N_k}{2})} \exp \left\{ -\frac{1}{2\sigma_k^2} \sum_{t:s_t=k} (y_t - \mu_k)^2 \right\} \times \prod_{k=1}^K \exp \left\{ -\frac{1}{2v} (\mu_k - m)^2 \right\} \\
& \propto \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2} \left(\sum_{t:s_t=k} (y_t^2 - 2y_t\mu_k + \mu_k^2) \right) - \frac{1}{2v} (\mu_k - m)^2 \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2} \left(\sum_{t:s_t=k} y_t^2 - 2\mu_k \sum_{t:s_t=k} y_t + \mu_k^2 \sum_{t:s_t=k} 1 \right) - \frac{1}{2v} (\mu_k^2 - 2\mu_k m + m^2) \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2} \left(\sum_{t:s_t=k} y_t^2 - 2\mu_k N_k \bar{y}_k + N_k \mu_k^2 \right) - \frac{1}{2v} (\mu_k^2 - 2\mu_k m + m^2) \right\} \\
& \propto \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2} (-2\mu_k N_k \bar{y}_k + N_k \mu_k^2) - \frac{1}{2v} (\mu_k^2 - 2\mu_k m) \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2 v} (v[-2\mu_k N_k \bar{y}_k + N_k \mu_k^2]) + \sigma_k^2 (\mu_k^2 - 2\mu_k m) \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{2\sigma_k^2 v} ([vN_k + \sigma_k^2] \mu_k^2 - 2\mu_k N_k \bar{y}_k v - 2\mu_k m \sigma_k^2) \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{\frac{2\sigma_k^2 v}{vN_k + \sigma_k^2}} \left[\mu_k^2 - 2 \left(\frac{vN_k \bar{y}_k + \sigma_k^2 m}{vN_k + \sigma_k^2} \right) \mu_k \right] \right\} \\
& = \prod_{k=1}^K \exp \left\{ -\frac{1}{\frac{2\sigma_k^2 v}{vN_k + \sigma_k^2}} \left[\left(\mu_k - \frac{vN_k \bar{y}_k + \sigma_k^2 m}{vN_k + \sigma_k^2} \right)^2 - \left(\frac{vN_k \bar{y}_k + \sigma_k^2 m}{vN_k + \sigma_k^2} \right)^2 \right] \right\} \\
& \propto \prod_{k=1}^K \exp \left\{ -\frac{vN_k + \sigma_k^2}{2\sigma_k^2 v} \left[\left(\mu_k - \frac{vN_k \bar{y}_k + \sigma_k^2 m}{vN_k + \sigma_k^2} \right)^2 \right] \right\}
\end{aligned}$$

So, because this factorises as a product of densities for μ_1, \dots, μ_K , the μ_k are independent in the full conditional distribution with

$$\mu_k | \mathbf{y}, \mathbf{s}, \sigma_k^2, \xi \sim N \left(\frac{vN_k \bar{y}_k + \sigma_k^2 m}{vN_k + \sigma_k^2}, \frac{\sigma_k^2 v}{vN_k + \sigma_k^2} \right)$$

where $\bar{y}_k = \frac{1}{N_k} \sum_{t:s_t=k} y_t$ for $k = 1, \dots, K$.

The full conditional distribution for $\sigma_1^2, \dots, \sigma_K^2$ is:

$$\begin{aligned}
\pi(\sigma_1^2, \dots, \sigma_K^2 | \mathbf{y}, \mathbf{s}, \mu_1, \dots, \mu_K, \xi) & \propto p(\mathbf{y}, \mathbf{s} | \mu_1, \dots, \mu_K, \sigma_1^2, \dots, \sigma_K^2, \xi) \pi(\sigma_1^2, \dots, \sigma_K^2) \\
& = p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_K, \sigma_1^2, \dots, \sigma_K^2) p(\mathbf{s} | \xi) \pi(\sigma_1^2, \dots, \sigma_K^2) \\
& \propto p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_K, \sigma_1^2, \dots, \sigma_K^2) \pi(\sigma_1^2, \dots, \sigma_K^2) \\
& = p(\mathbf{y} | \mathbf{s}, \mu_1, \dots, \mu_K, \sigma_1^2, \dots, \sigma_K^2) \times \prod_{k=1}^K \pi(\sigma_k^2)
\end{aligned}$$

$$\begin{aligned}
&= \prod_{k=1}^K \sigma_k^{2\left(-\frac{N_k}{2}\right)} \exp \left\{ -\frac{1}{2\sigma_k^2} \sum_{t:s_t=k} (y_t - \mu_k)^2 \right\} \times \prod_{k=1}^K \frac{\beta}{\Gamma(\alpha)} \sigma_k^{2(-\alpha-1)} \exp \left\{ -\frac{\beta}{\sigma_k^2} \right\} \\
&\propto \prod_{k=1}^K \sigma_k^{2\left(-\alpha-\frac{N_k}{2}-1\right)} \exp \left\{ -\frac{1}{2\sigma_k^2} \left(\sum_{t:s_t=k} (y_t - \mu_k)^2 \right) - \frac{\beta}{\sigma_k^2} \right\} \\
&= \prod_{k=1}^K \sigma_k^{2\left(-\alpha-\frac{N_k}{2}-1\right)} \exp \left\{ -\frac{1}{\sigma_k^2} \left(\beta + \sum_{t:s_t=k} \frac{(y_t - \mu_k)^2}{2} \right) \right\}
\end{aligned}$$

So, we can see that σ_k^2 is independent in its full conditional distributions and that

$$\sigma_k^2 | \mathbf{y}, \mathbf{s}, \mu_k, \xi \sim \text{invgamma} \left(\alpha + \frac{N_k}{2}, \beta + \sum_{t:s_t=k} \frac{(y_t - \mu_k)^2}{2} \right)$$