

BDA Ch.2 Exercise 8.

$$\begin{array}{ll}
 Y | \theta \sim N(\theta, \underbrace{\sigma^2}_{\text{known}}) & | \quad \sigma^2 = 400 \\
 \theta \sim N(\mu_0, \tau_0^2) & | \quad \mu_0 = 180, \tau_0^2 = 1600 \\
 \rightarrow \theta | y \sim N(\mu_n, \tau_n^2) & | \quad \bar{y} = 150
 \end{array}
 \quad \langle \text{numeric values} \rangle$$

$$\begin{aligned}
 (a) \quad \theta | y &\sim N(\mu_n, \tau_n^2) \\
 \mu_n &= \frac{\frac{1}{\tau_0^2}}{\frac{1}{\tau_0^2} + \frac{n}{\sigma^2}} \mu_0 + \frac{\frac{n}{\sigma^2}}{\frac{1}{\tau_0^2} + \frac{n}{\sigma^2}} \bar{y} \quad \tau_n^2 = \frac{1}{\frac{1}{\tau_0^2} + \frac{n}{\sigma^2}}
 \end{aligned}$$

Now, plug numeric values.

$$\begin{aligned}
 \theta | y &\sim N \left(\frac{\frac{1}{1600}}{\frac{1}{1600} + \frac{n}{400}} 180 + \frac{\frac{n}{400}}{\frac{1}{1600} + \frac{n}{400}} \cdot 150, \frac{1}{\frac{1}{1600} + \frac{n}{400}} \right) \\
 &= N \left(\frac{180 + 600n}{1 + 4n}, \frac{1600}{1 + 4n} \right)
 \end{aligned}$$

$$\begin{aligned}
 (b) \quad \tilde{Y} | \theta &\sim N(\theta, \sigma^2) \\
 \tilde{Y} | y &\sim N(\mu_n, \sigma^2 + \tau_n^2) \\
 &= N \left(\frac{180 + 600n}{1 + 4n}, 400 + \frac{1600}{1 + 4n} \right)
 \end{aligned}$$

$$\begin{aligned}
 (c) \quad n=10 &\rightarrow \theta | y \sim N \left(\frac{6180}{41}, \frac{1600}{41} \right) \rightarrow 95\% \text{ CI} : (138.4, 162.9) \\
 \tilde{Y} | y &\sim N \left(\frac{6180}{41}, 400 + \frac{1600}{41} \right) \rightarrow 95\% \text{ CI} : (109.6, 191.2)
 \end{aligned}$$

$$\begin{aligned}
 (d) \quad n=100 &\rightarrow \theta | y \sim N \left(\frac{60180}{401}, \frac{1600}{401} \right) \rightarrow 95\% \text{ CI} : (146.1, 153.9) \\
 \tilde{Y} | y &\sim N \left(\frac{60180}{401}, 400 + \frac{1600}{401} \right) \rightarrow 95\% \text{ CI} : (110.6, 189.4)
 \end{aligned}$$

Normal data with a conjugate prior distribution.

Likelihood $\left(\begin{array}{l} Y_i | \mu, \sigma^2 \sim N(\mu, \sigma^2) \\ \text{data} | \mu, \sigma^2 : p(\text{data} | \mu, \sigma^2) = \prod p(y_i | \mu, \sigma^2) = \left(\frac{1}{2\pi\sigma^2} \right)^{\frac{n}{2}} \cdot e^{-\frac{\sum (y_i - \mu)^2}{2\sigma^2}} \end{array} \right.$

prior $\left(\begin{array}{l} \mu, \sigma^2 : p(\mu, \sigma^2) = p(\mu | \sigma^2) \cdot p(\sigma^2) \\ \mu | \sigma^2 \sim N\left(\mu_0, \frac{\sigma_0^2}{k_0}\right) \xrightarrow{p(\mu | \sigma^2)} \text{pdf.} \left(\frac{k_0}{2\pi\sigma^2} \right)^{\frac{1}{2}} \cdot e^{-\frac{k_0(\mu - \mu_0)^2}{2\sigma^2}} \\ \sigma^2 \sim \chi^{-2}(V_0, \sigma_0^2) = \Gamma^{-1}\left(\frac{V_0}{2}, \frac{V_0 \sigma_0^2}{2}\right) \xrightarrow{p(\sigma^2)} \text{pdf.} \frac{\left(\frac{V_0 \sigma_0^2}{2}\right)^{\frac{V_0}{2}}}{\Gamma(\frac{V_0}{2})} \cdot \left(\frac{1}{\sigma^2}\right)^{\frac{V_0}{2}+1} \cdot e^{-\frac{V_0 \sigma_0^2}{2\sigma^2}} \end{array} \right.$

posterior $p(\mu, \sigma^2 | \text{data}) \propto p(\text{data} | \mu, \sigma^2) \cdot p(\mu | \sigma^2) \cdot p(\sigma^2)$

[결과] prior $\mu, \sigma^2 \sim N\text{-Inv-}\chi^2\left(\mu_0, \frac{\sigma_0^2}{k_0}; V_0, \sigma_0^2\right)$

posterior $\mu, \sigma^2 | y \sim N\text{-Inv-}\chi^2\left(\mu_n, \frac{\sigma_n^2}{k_n}; V_n, \sigma_n^2\right)$

$$\begin{cases} \mu_n = \frac{k_0}{k_0+n} \mu_0 + \frac{n}{k_0+n} \bar{y} \\ k_n = k_0 + n \\ V_n = V_0 + n \\ V_n \sigma_n^2 = V_0 \sigma_0^2 + (n-1)S^2 + \frac{k_0 n}{k_0+n} (\bar{y} - \mu_0)^2 \end{cases}$$

Marginal $\sigma^2 | y$
$$\begin{aligned} p(\sigma^2 | y) &\propto p(\sigma^2) \cdot p(y | \sigma^2) = p(\sigma^2) \cdot \int p(y, \mu | \sigma^2) d\mu = p(\sigma^2) \cdot \int p(y | \mu, \sigma^2) p(\mu | \sigma^2) d\mu \\ &\propto \int \left(\frac{1}{\sigma^2} \right)^{\frac{n}{2}} e^{-\frac{\sum (y_i - \mu)^2}{2\sigma^2}} \cdot \left(\frac{k_0}{2\pi\sigma^2} \right)^{\frac{1}{2}} e^{-\frac{k_0(\mu - \mu_0)^2}{2\sigma^2}} d\mu \cdot \left(\frac{1}{\sigma^2} \right)^{\frac{V_0}{2}+1} \cdot e^{-\frac{V_0 \sigma_0^2}{2\sigma^2}} \\ &\propto \left(\frac{1}{\sigma^2} \right)^{\frac{n+1}{2}} \cdot e^{-\frac{(n+1)\sigma^2}{2\sigma^2}} \cdot e^{-\frac{n k_0 (\bar{y} - \mu_0)^2}{n k_0 + 2\sigma^2}} \cdot \left(\frac{1}{\sigma^2} \right)^{-\frac{1}{2}} \cdot \left(\frac{1}{\sigma^2} \right)^{\frac{V_0}{2}+1} \cdot e^{-\frac{V_0 \sigma_0^2}{2\sigma^2}} \\ &= \left(\frac{1}{\sigma^2} \right)^{\frac{V_0+n+1}{2}} \cdot \exp \left[-\frac{1}{2\sigma^2} \left\{ V_0 \sigma_0^2 + (n+1)S^2 + \frac{n k_0 (\bar{y} - \mu_0)^2}{n+k_0} \right\} \right] \end{aligned} \quad \Bigg) \text{Equal.}$$

Compare: $\chi^{-2}(V_n, \sigma_n^2) \rightarrow \text{pdf} \propto \left(\frac{1}{\sigma^2} \right)^{\frac{V_n}{2}+1} \cdot e^{-\frac{V_n \sigma_n^2}{2\sigma^2}}$
 $= \Gamma^{-1}\left(\frac{V_n}{2}, \frac{V_n \sigma_n^2}{2}\right)$

$\therefore \sigma^2 | y \sim \chi^2(V_n, \sigma_n^2)$, $V_n = V_0 + n$
 $V_n \sigma_n^2 = V_0 \sigma_0^2 + (n+1)S^2 + \frac{n k_0 (\bar{y} - \mu_0)^2}{n+k_0}$

$$\otimes \int \left(\frac{1}{\sigma^2}\right)^{\frac{n}{2}} \cdot e^{-\frac{\Sigma(y-m)^2}{2\sigma^2}} \cdot \left(\frac{1}{\sigma^2}\right)^{\frac{1}{2}} \cdot e^{-\frac{k_0(m-m_0)^2}{2\sigma^2}} dm$$

$$= \int \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} \cdot \exp\left[-\frac{1}{2\sigma^2} \{\Sigma(y-m)^2 + k_0(m-m_0)^2\}\right] dm$$

$$= \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} \int \exp\left[-\frac{1}{2\sigma^2} \{\Sigma(y-\bar{y})^2 + \Sigma(\bar{y}-m)^2 + 2\Sigma(y-\bar{y})(\bar{y}-m) + k_0(m-m_0)^2\}\right] dm$$

$$= \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} e^{-\frac{(n+1)\bar{y}^2}{2\sigma^2}} \int \exp\left[-\frac{1}{2\sigma^2} \{\Sigma(m-\bar{y})^2 + k_0(m-m_0)^2\}\right] dm$$

$$\boxed{} = \Sigma(m-\bar{y})^2 + k_0(m-m_0)^2 = nm^2 - 2n\bar{y}m + n\bar{y}^2 + k_0m^2 - 2k_0m_0m + k_0m_0^2$$

$$= (n+k_0) \left(m - \frac{n\bar{y} + k_0m_0}{n+k_0}\right)^2 + n\bar{y}^2 + k_0m_0^2 - \frac{(n\bar{y} + k_0m_0)^2}{n+k_0}$$

$$= (n+k_0)(m-A)^2 + \frac{nk_0(\bar{y}-m_0)^2}{n+k_0}$$

$$\otimes = \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} e^{-\frac{(n+1)\bar{y}^2}{2\sigma^2}} \int \exp\left[-\frac{1}{2\sigma^2} \left\{ (n+k_0)(m-A)^2 + \frac{nk_0(\bar{y}-m_0)^2}{n+k_0} \right\}\right] dm$$

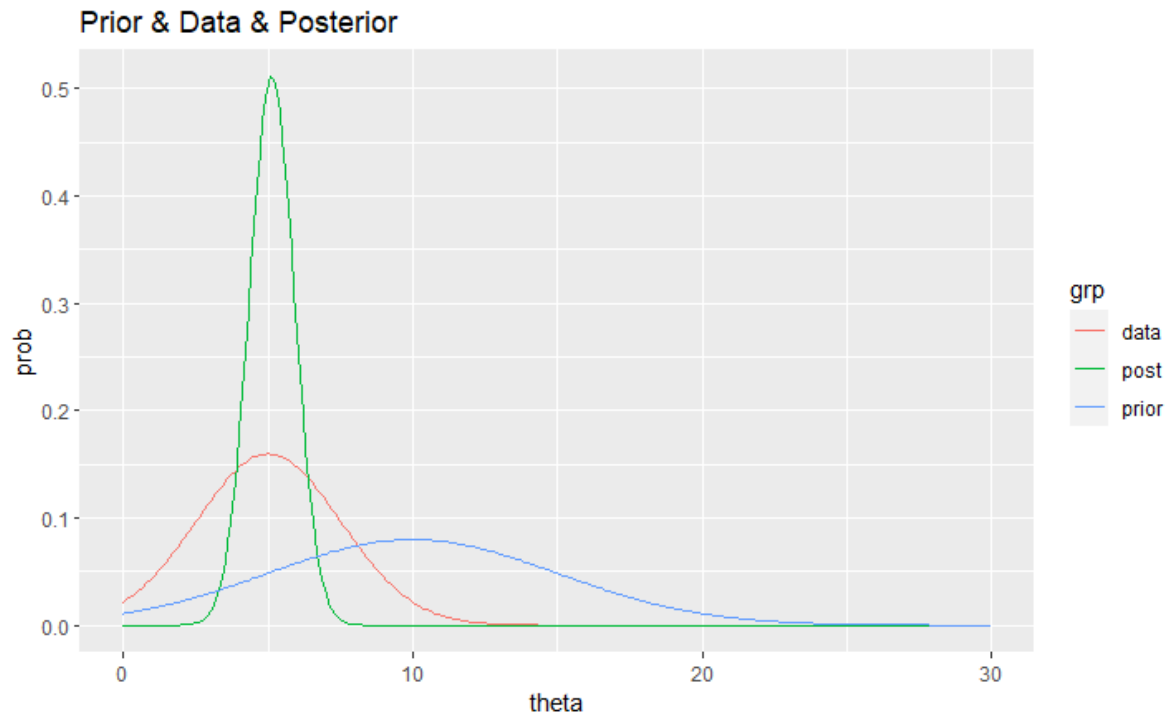
$$= \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} \cdot e^{-\frac{(n+1)\bar{y}^2}{2\sigma^2}} \cdot e^{-\frac{\frac{nk_0(\bar{y}-m_0)^2}{n+k_0}}{2\sigma^2}} \cdot \int e^{-\frac{(n+k_0)(m-A)^2}{2\sigma^2}} dm$$

$$= \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} \cdot e^{-\frac{(n+1)\bar{y}^2}{2\sigma^2}} \cdot e^{-\frac{\frac{nk_0(\bar{y}-m_0)^2}{n+k_0}}{2\sigma^2}} \cdot \frac{\sqrt{2\pi} \sigma}{\sqrt{n+k_0}}$$

$$\propto \left(\frac{1}{\sigma^2}\right)^{\frac{n+1}{2}} e^{-\frac{(n+1)\bar{y}^2}{2\sigma^2}} \cdot e^{-\frac{\frac{nk_0(\bar{y}-m_0)^2}{n+k_0}}{2\sigma^2}} \cdot \left(\frac{1}{\sigma^2}\right)^{-\frac{1}{2}}$$

3. R코드 시각화 따라해보기

Normal model with known variance (결과만)



Normal model with known mean (결과만)

