

1.
 (a) prior: $\mu \sim N(\mu_0, \tau_0^2)$
 likelihood: $y | \mu \sim N(\mu, \sigma^2)$
 posterior: $\mu | y \sim N(\mu_n, \tau_n^2)$

$$\mu_n = \frac{\frac{1}{\tau_0^2}}{\frac{n}{\sigma^2} + \frac{1}{\tau_0^2}} \mu + \frac{\frac{n}{\sigma^2}}{\frac{n}{\sigma^2} + \frac{1}{\tau_0^2}} \bar{y}$$

$$\tau_n^2 = \frac{1}{\frac{n}{\sigma^2} + \frac{1}{\tau_0^2}}$$

$$\theta \sim N(180, 40^2)$$

$$y | \theta \sim N(150, 20^2)$$

$$\mu_n = \frac{\frac{1}{1600}}{\frac{n}{400} + \frac{1}{1600}} 180 + \frac{\frac{n}{400}}{\frac{n}{400} + \frac{1}{1600}} 150 = \frac{180 + 600n}{1 + 4n}$$

$$\tau_n^2 = \frac{1}{\frac{n}{400} + \frac{1}{1600}} = \frac{1600}{1 + 4n}$$

$$\theta | y \sim N\left(\frac{180 + 600n}{1 + 4n}, \frac{1600}{1 + 4n}\right)$$

(b) $\tilde{y} | y \sim N(\mu_n, \sigma^2 + \tau_n^2)$

$$\hookrightarrow \tilde{y} | y \sim N\left(\frac{180 + 600n}{1 + 4n}, 20^2 + \frac{1600}{1 + 4n}\right)$$

(c) $n = 10$

$$\theta | y \sim N\left(\frac{6180}{41}, \frac{1600}{41}\right)$$

$$\rightarrow 95\% \text{ C.I.} : (138.5, 163)$$

$$\tilde{y} | y \sim N\left(\frac{6180}{41}, 400 + \frac{1600}{41}\right)$$

$$\rightarrow 95\% \text{ C.I.} : (109.7, 191.8)$$

(d) $n = 100$

$$\theta | y \sim N\left(\frac{60180}{401}, \frac{1600}{401}\right)$$

$$\rightarrow 95\% \text{ C.I.} : (146.2, 154)$$

$$\tilde{y} | y \sim N\left(\frac{60180}{401}, 400 + \frac{1600}{401}\right)$$

$$\rightarrow 95\% \text{ C.I.} : (110.7, 189.5)$$

2.

Prior: $P(\mu, \sigma^2) = P(\mu | \sigma^2) \cdot P(\sigma^2)$

$$\mu | \sigma^2 \sim \mathcal{N}(\mu_0, \sigma^2 / k_0)$$

$$\sigma^2 \sim \text{Inv-}\chi^2(v_0, \sigma_0^2) = \text{Inv} \sim \Gamma(v_0/2, v_0 \sigma_0^2 / 2)$$

Joint Prior:

$$P(\mu, \sigma^2) \propto \left(\frac{1}{\sigma^2}\right)^{1/2} \left(\frac{1}{\sigma^2}\right)^{\frac{v_0}{2}+1} \cdot \exp\left\{-\frac{1}{2\sigma^2}(k_0(\mu-\mu_0)^2 + v_0\sigma_0^2)\right\}$$

$$\sim \text{Normal-Inv-}\chi^2(\mu_0, \sigma_0^2/k_0; v_0, \sigma_0^2)$$

likelihood: $P(y | \mu, \sigma^2) \propto (\sigma^2)^{-n/2} \cdot \exp\left\{-\frac{1}{2\sigma^2} \sum (y_i - \mu)^2\right\}$

posterior: $p(\mu, \sigma^2 | y) \propto p(y | \mu, \sigma^2) \cdot p(\mu, \sigma^2)$

$$\propto (\sigma^2)^{-\frac{n+v_0}{2}-1} \cdot \exp\left\{-\frac{1}{2\sigma^2}(v_0\sigma_0^2 + (n-1)S^2 + \frac{nk_0(\bar{y}-\mu_0)^2}{n+k_0})\right\}$$

$$\sim \text{Normal-Inv-}\chi^2(\mu_n, \sigma_n^2/k_n; v_n, \sigma_n^2)$$

$$\mu_n = \frac{k_0}{k_0+n} \mu_0 + \frac{n}{k_0+n} \bar{y}$$

$$k_n = k_0 + n$$

$$v_n = v_0 + n$$

$$v_n \sigma_n^2 = v_0 \sigma_0^2 + (n-1)S^2 + \frac{k_0 n}{k_0+n} (\bar{y} - \mu_0)^2$$

Marginal Posterior for σ^2 : $p(\sigma^2|y)$

$$\begin{aligned}
 p(\sigma^2|y) &\propto p(\sigma^2) \cdot p(y|\sigma^2) = p(\sigma^2) \int p(y|\mu, \sigma^2) \cdot p(\mu, \sigma^2) d\mu \\
 &\propto (1/\sigma^2)^{1/2+1} \exp(-1/2 \sigma_0^2 / \sigma^2) \cdot \int \left(\frac{1}{\sigma^2}\right)^{n/2} \cdot \exp\left\{-\frac{1}{2\sigma^2}((n-1)S^2 + n(\bar{y}-\mu)^2)\right\} \\
 &\quad \cdot \left(\frac{k_0}{\sigma^2}\right)^{1/2} \exp\left\{-\frac{k_0}{2\sigma^2}(\mu-\mu_0)^2\right\} d\mu \\
 &\propto (1/\sigma^2)^{1/2+1} \cdot \exp\left\{-\frac{1/2 \sigma_0^2}{\sigma^2}\right\} \cdot \left(\frac{1}{\sigma^2}\right)^{n/2} \cdot \exp\left\{-\frac{1}{2\sigma^2}((n-1)S^2)\right\} \cdot \left(\frac{1}{\sigma^2}\right)^{1/2} \\
 &\quad \cdot \int \exp\left\{-\frac{1}{2\sigma^2}n(\bar{y}-\mu)^2 - \frac{k_0}{2\sigma^2}(\mu-\mu_0)^2\right\} d\mu
 \end{aligned}$$

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$$\begin{aligned}
 &* \int \exp\left\{-\frac{1}{2\sigma^2}n(\bar{y}-\mu)^2 - \frac{k_0}{2\sigma^2}(\mu-\mu_0)^2\right\} d\mu \\
 &= \int \exp\left\{-\frac{1}{2\sigma^2}(n(\bar{y}-\mu)^2 + k_0(\mu-\mu_0)^2)\right\} d\mu \\
 &= \int \exp\left\{-\frac{1}{2\sigma^2}\left(n\mu^2 - 2n\bar{y}\mu + n\bar{y}^2 + k_0\mu^2 + 2k_0\mu\mu_0 + k_0\mu_0^2\right)\right\} d\mu \\
 &= \int \exp\left\{-\frac{1}{2\sigma^2}\left((n+k_0)\left(\mu - \frac{n\bar{y}+k_0\mu_0}{n+k_0}\right)^2 + n\bar{y}^2 + k_0\mu_0^2 - \frac{(n\bar{y}+k_0\mu_0)^2}{n+k_0}\right)\right\} d\mu \\
 &= \exp\left\{-\frac{1}{2\sigma^2}\left(\frac{n k_0 (\bar{y}-\mu_0)^2}{n+k_0}\right)\right\} \int \exp\left\{-\frac{1}{2\sigma^2}(n+k_0)\left(\mu - \frac{n\bar{y}+k_0\mu_0}{n+k_0}\right)^2\right\} d\mu \\
 &= \exp\left\{-\frac{1}{2\sigma^2}\left(\frac{n k_0 (\bar{y}-\mu_0)^2}{n+k_0}\right)\right\} \frac{\sqrt{2\pi} \sigma}{\sqrt{n+k_0}}
 \end{aligned}$$

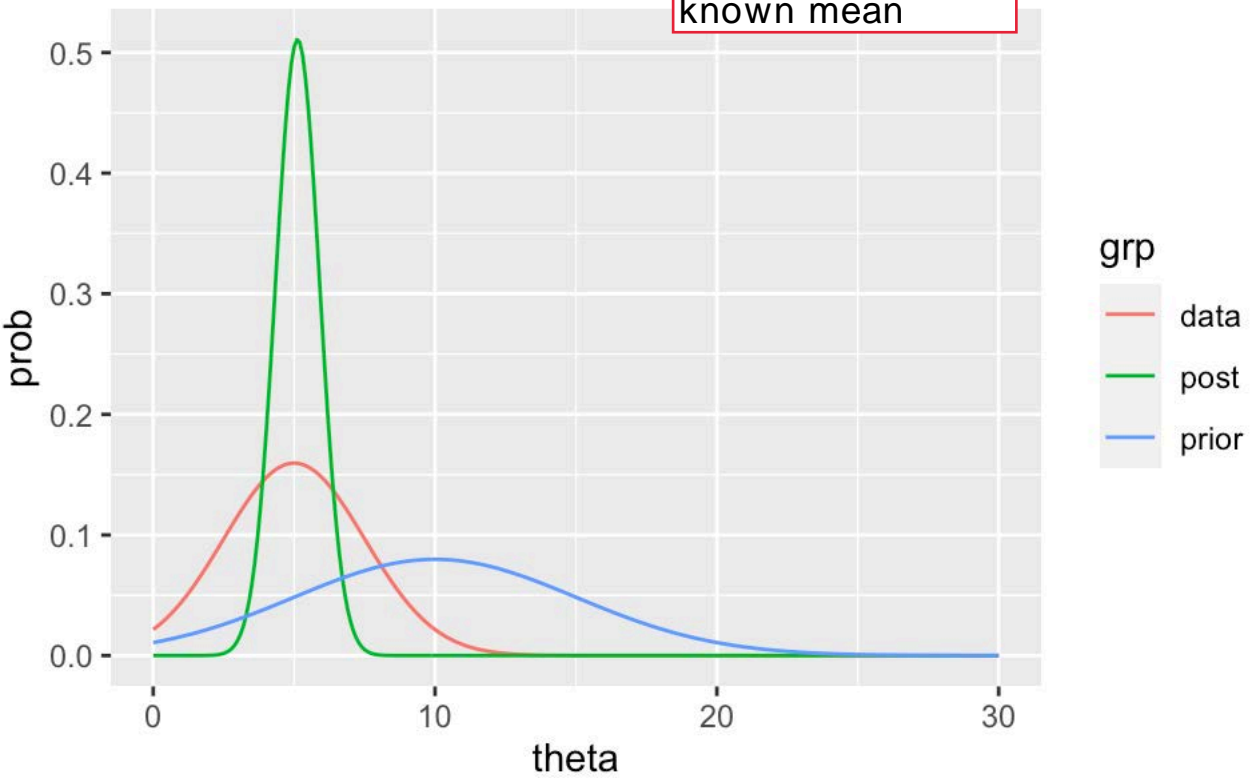
$$\therefore = (1/\sigma^2)^{V_0 + 1/2 + \frac{1}{2}} \cdot \frac{\sqrt{2\pi} \sigma}{\sqrt{n+k_0}} \exp \left\{ -\frac{1}{2\sigma^2} (V_0 \sigma_0^2 + (n-1)S^2 + \frac{nk_0}{n+k_0} (\bar{y} - \mu)^2) \right\}$$

$$\propto (1/\sigma^2)^{V_0 + 1/2} \cdot \exp \left\{ -\frac{1}{2\sigma^2} (V_n \sigma_n^2) \right\} \sim \text{Inv-}\chi^2(V_n, \sigma_n^2)$$

where , $V_n = V_0 + n$
 $V_n \sigma_n^2 = V_0 \sigma_0^2 + (n-1)S^2 + \frac{nk_0}{n+k_0} (\bar{y} - \mu)^2$

Prior & Data & Posterior

Normal dist with
known mean



Prior & Posterior

Normal dist with
known variance

