

Exercise 2

A simple Gaussian location model

A). $p(w) \sim \text{Gamma}(\frac{d}{2}, \frac{\eta}{2})$, $p(\theta|w) \sim N(\mu, (wk)^{-1})$, prior $p(\theta)$?

$$p(\theta) = \int_w p(\theta, w) dw = \int_w p(\theta|w) p(w) dw$$

$$\propto \int_w w^{\frac{d+1}{2}-1} \exp\left\{-w \cdot \frac{k(\theta-\mu)^2}{2}\right\} \exp\left\{-w \cdot \frac{\eta}{2}\right\} dw$$

$$= \int_w w^{\frac{d+1}{2}-1} \exp\left\{-\frac{w}{2} [k(\theta-\mu)^2 + \eta]\right\} dw$$

Assume: $\alpha = \frac{d+1}{2}$, $\beta = k(\theta-\mu)^2 + \eta$

$$= \int_w w^{\alpha-1} \exp\left\{-\frac{w}{2} \beta\right\} dw$$

This is integration on a Gamma Function:

$$= \frac{\Gamma(\alpha)}{b^\alpha} \propto b^{-\alpha} = [k(\theta-\mu)^2 + \eta]^{-\frac{d+1}{2}}$$

$$= \left[1 + \frac{1}{d} \frac{(\theta-\mu)^2}{(\frac{\eta}{dk})}\right]^{-\frac{d+1}{2}}$$

\therefore center: $m = \mu$

degree of freedom: $v = d$

scale parameter: $s^2 = \frac{\eta}{dk}$

$$\propto \left[1 + \frac{1}{v} \frac{(\theta-\mu)^2}{s^2}\right]^{-\frac{v+1}{2}}$$

B). Bayesian theory:

$$p(\theta, w | y) \propto p(y | \theta, w) p(\theta, w)$$

$$p(\theta, w) \sim w^{\frac{d+1}{2}-1} \exp\left\{-w \cdot \frac{k(\theta-\mu)^2}{2}\right\} \exp\left\{-w \cdot \frac{\eta}{2}\right\}$$

$$p(y_i | \theta, w) \sim N(\theta, w^{-1}) \Rightarrow p(y | \theta, w) \propto \prod_{i=1}^n w^{\frac{1}{2}} \exp\left\{-\frac{w}{2}(y_i - \theta)^2\right\}$$

$$= w^{\frac{n}{2}} \exp\left\{-\frac{w}{2} \sum_{i=1}^n (y_i - \theta)^2\right\}$$

$$= w^{\frac{n}{2}} \exp\left\{-\frac{w}{2} \sum_{i=1}^n (y_i - \bar{y} + \bar{y} - \theta)^2\right\}$$

$$= w^{\frac{n}{2}} \exp\left\{-\frac{w}{2} [S_y + n(\bar{y} - \theta)^2]\right\}, \quad S_y = \sum_{i=1}^n (y_i - \bar{y})^2$$

$$p(\theta, w | y) \propto w^{\frac{n}{2}} \exp\left\{-\frac{w}{2} [S_y + n(\bar{y} - \theta)^2]\right\} w^{\frac{d+1}{2}-1} \exp\left\{-w \cdot \frac{k(\theta-\mu)^2}{2}\right\} \exp\left\{-w \cdot \frac{\eta}{2}\right\}$$

$$= W^{\frac{n+d+1}{2}-1} \exp\left\{-\frac{W}{2} [S_y + n(\bar{y} - \theta)^2 + k(\theta - \mu)^2]\right\} \exp\left\{-W \cdot \frac{\eta}{2}\right\}$$

$$= W^{\frac{n+d+1}{2}-1} \exp\left\{-\frac{W}{2} [(n+k)\theta^2 - 2(n\bar{y} + k\mu)\theta + (n\bar{y}^2 + k\mu^2)]\right\} \exp\left\{-\frac{W}{2} (S_y + \eta)\right\}$$

$$= W^{\frac{n+d+1}{2}-1} \exp\left\{-\frac{W(n+k)}{2} \left[\theta - \frac{n\bar{y} + k\mu}{n+k}\right]^2\right\} \exp\left\{-\frac{W}{2} \left[S_y + \eta - \frac{nk}{(n+k)} (\mu - \bar{y})^2\right]\right\} \dots (1)$$

$$\therefore d^* = n+d, \quad \mu^* = \frac{k\mu + n\bar{y}}{n+k}, \quad k^* = n+k, \quad \eta^* = \eta + S_y - \frac{nk}{n+k} (\mu - \bar{y})^2$$

$$= W^{\frac{d^*+1}{2}-1} \exp\left\{-\frac{Wk^*}{2} (\theta - \mu^*)^2\right\} \exp\left\{-\frac{W\eta^*}{2}\right\}$$

// C). According to (B), then:

$$p(\theta | y, w) \propto \exp\left\{-\frac{wk^*}{2} (\theta - \mu^*)^2\right\} \Rightarrow p(\theta | y, w) \sim N(\mu^*, (wk^*)^{-1})$$

D. Following (B), we know:

Exercise 2

③

$$p(\theta, w | y) \propto w^{\frac{d^*+1}{2}-1} \exp\left\{-\frac{wk^*}{2}(\theta - \mu^*)^2\right\} \exp\left\{-\frac{w\eta^*}{2}\right\}$$

$$p(w | y) = \int_{\theta} p(w, \theta | y) d\theta$$

$$\propto w^{\frac{d^*+1}{2}-1} \exp\left\{-\frac{w\eta^*}{2}\right\} \underbrace{\int_{\theta} \exp\left\{-\frac{wk^*}{2}(\theta - \mu^*)^2\right\} d\theta}_{\text{integration on normal function}}$$

$$\propto w^{\frac{d^*+1}{2}-1} \exp\left\{-\frac{w\eta^*}{2}\right\} \cdot \frac{\sqrt{2\pi}}{(wk^*)^{\frac{1}{2}}}$$

$$\propto w^{\frac{d^*}{2}-1} \exp\left\{-\frac{w\eta^*}{2}\right\} \quad (\text{Gamma kernel})$$

$$\Rightarrow p(w | y) \sim \text{Gamma}\left(\frac{d^*}{2}, \frac{\eta^*}{2}\right)$$

E. $p(w | y) \sim \text{Gamma}\left(\frac{d^*}{2}, \frac{\eta^*}{2}\right) \rightarrow \text{from (D)}$, what is $p(\theta | y)$?

$p(\theta | y, w) \sim N(\mu^*, (wk^*)^{-1}) \rightarrow \text{from (C)}$, parameters are: $\left[1 + \frac{1}{V} \cdot \frac{(\theta - \mu)^2}{S^2}\right]^{-\frac{V+1}{2}}$

According to (A), the update of posterior distribution parameters are:

$$\begin{cases} \text{center: } m = \mu^* \\ \text{degree of freedom: } V = d^* \\ \text{scale parameter: } S^2 = \frac{\eta^*}{(d^* k^*)} \end{cases}$$

μ^*, d^*, k^*, η^* are on page ② from (B)

F) $p(w) \sim \text{Gamma}\left(\frac{d}{2}, \frac{\eta}{2}\right) \propto w^{\frac{d}{2}-1} e^{-\frac{\eta}{2}w}$

$\lim_{d, \eta \rightarrow 0} p(w) \propto w^{-1}$

$\int_0^{+\infty} p(w) \rightarrow +\infty$

not pdf

② $p(\theta | w) \sim N(\mu, (wk)^{-1}) \propto (wk) \exp\left\{-\frac{wk}{2}(\theta - \mu)^2\right\}$

$\lim_{k \rightarrow 0} p(\theta | w) \propto 0$

not pdf

(G) According to (E)

$$p(\theta|y) \propto \left[1 + \frac{1}{d^*} \cdot \frac{(\theta - \mu^*)^2}{\left(\frac{\eta^*}{d^* k^*}\right)} \right]^{-\frac{d^*+1}{2}}$$

$$\begin{aligned} d^* &= n + d \\ k^* &= n + k \\ \mu^* &= (k\mu + n\bar{y}) / (n + k) \\ \eta^* &= \eta + S_y - \frac{nk}{n+k} (\mu - \bar{y})^2 \end{aligned}$$

$$\lim_{d, k, \eta \rightarrow 0} p(\theta|y) \propto \left[1 + \frac{n}{S_y} (\theta - \bar{y})^2 \right]^{-\frac{n+1}{2}} \quad (\text{Student dist.})$$

According to (D)

$$p(w|y) \sim \text{Gamma}\left(\frac{d^*}{2}, \frac{\eta^*}{2}\right) \propto w^{\frac{d^*}{2}-1} \exp\left\{-w \frac{\eta^*}{2}\right\}$$

$$\lim_{d, \eta \rightarrow 0} p(w|y) \propto w^{\frac{n}{2}-1} \exp\left\{-w \frac{S_y}{2}\right\}$$

used in hyperparameters

$$p(\theta|w) \propto 1 \quad \text{flat}$$

$$p(w) \propto \frac{1}{w}$$

(t dist. like B₀)
improper prior
thick limit
"objective" prior
Jeffrey's prior
reference prior
MDL
maximum likelihood

(Gamma dist.)

$$(H) \quad p(\theta|y) \propto \left[1 + \frac{1}{v} \cdot \frac{(\theta - m)^2}{S^2} \right]^{-\frac{v+1}{2}}, \quad m = \mu^*, \quad v = d^*, \quad S^2 = \frac{\eta^*}{d^* k^*}$$

$$\eta, d, k \rightarrow 0 \Rightarrow m = \mu^* = (k\mu + n\bar{y}) / (n + k) \rightarrow \bar{y}$$

$$v = d^* = n + k \rightarrow n$$

$$S^2 = \frac{\eta^*}{d^* k^*} \rightarrow \frac{S_y}{n^2} = \frac{\text{Var}(y)}{n}$$

Credible interval of θ :

$$\theta \in \bar{y} \pm t^* S = \bar{y} \pm t^* \sqrt{\frac{\text{Var}(y)}{n}}$$

Same as classical freq. CI.

The conjugate Gaussian Linear Model

Ex2 Page 5

A). Derive $p(\beta|y, w)$

$$w \sim \Gamma\left(\frac{d}{2}, \frac{\eta}{2}\right), \quad (\beta|w) \sim N(m, (wk)^{-1}), \quad (y|\beta, w) \sim N(x\beta, (w\Lambda)^{-1})$$

$$\begin{aligned} p(\beta, w|y) &\propto p(y|\beta, w) p(\beta, w) = p(y|\beta, w) p(\beta|w) p(w) \\ &\propto w^{\frac{n}{2}} \exp\left\{-\frac{1}{2}(y-x\beta)^T (w\Lambda)(y-x\beta)\right\} w^{\frac{p}{2}} \exp\left\{-\frac{1}{2}(\beta-m)^T (wk)(\beta-m)\right\} w^{\frac{d}{2}-1} \exp\left\{-\frac{\eta}{2}w\right\} \\ &= w^{\frac{n+p+d}{2}-1} \exp\left\{-\frac{w}{2}[(y-x\beta)^T \Lambda (y-x\beta) + (\beta-m)^T k(\beta-m) + \eta]\right\} \\ &= w^{\frac{n+p+d}{2}-1} \exp\left\{-\frac{1}{2}(\beta-m^*)^T (wk^*)(\beta-m^*)\right\} \cdot \exp\left\{-\frac{w}{2}[y^T \Lambda y + m^T k m + \eta - m^{*T} k^* m^*]\right\} \\ &\quad k^* = X^T \Lambda X + k, \quad m^* = (k^*)^{-1}(X^T \Lambda y + km), \end{aligned}$$

$$\underbrace{w^{\frac{p}{2}} \exp\left\{-\frac{1}{2}(\beta-m^*)^T (wk^*)(\beta-m^*)\right\}}_{\text{Normal } p(\beta|y, w)} \underbrace{w^{\frac{d}{2}-1} \exp\left\{-\frac{\eta^*}{2}w\right\}}_{\text{Gamma } p(w|y)}$$

$$d^* = n+d, \quad \eta^* = \eta + y^T \Lambda y + m^T k m - m^{*T} k^* m^*$$

Normal - Gamma distribution

$$\Rightarrow p(\beta|y, w) \sim N(m^*, (wk^*)^{-1}) \quad (A) \quad \checkmark$$

$$p(w|y) = \int_{\beta} p(w, \beta|y) \sim \Gamma\left(\frac{d^*}{2}, \frac{\eta^*}{2}\right) \quad (B) \quad \checkmark$$

(C). Derive $p(\beta|y)$?

$$\begin{aligned} p(\beta|y) &= \int_w p(w, \beta|y) \propto \int_w w^{\frac{p}{2}} \exp\left\{-\frac{1}{2}(\beta-m^*)^T (wk^*)(\beta-m^*)\right\} w^{\frac{d^*}{2}-1} \exp\left\{-\frac{\eta^*}{2}w\right\} \\ &= \int_w w^{\frac{p+d^*}{2}-1} \exp\left\{-w\left[\frac{1}{2}(\beta-m^*)^T k^*(\beta-m^*) + \frac{\eta^*}{2}\right]\right\} dw \quad (\text{integrate on } w \text{ distr}) \\ &\propto \left[\eta^* + (\beta-m^*)^T k^*(\beta-m^*)\right]^{-\frac{p+d^*}{2}} \\ &\propto \left[1 + \frac{1}{\eta^*} (\beta-m^*)^T \frac{k^*}{\left(\frac{\eta^*}{d^*}\right)} (\beta-m^*)\right]^{-\frac{d^*+p}{2}} \end{aligned}$$

$$p(x) \sim N(\mu, \Sigma)$$

$$\Leftrightarrow p(x) = \left(\frac{1}{2\pi|\Sigma|}\right)^{\frac{1}{2}} \exp\left\{-\frac{1}{2}(x-\mu)^T \Sigma^{-1}(x-\mu)\right\}$$