

$$1. \quad x | \sigma^2 \sim N(\mu, \sigma^2), \quad y | x, \sigma^2 \sim N(\rho x, \sigma^2) \quad \pi(\sigma^2) \propto \frac{1}{\sigma^2}, \sigma^2 > 0$$

$$\begin{aligned} (a) \quad m(x) &= \int_0^\infty \frac{1}{\sqrt{2\pi}\sigma^2} \cdot \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) \cdot \frac{1}{\sigma^2} d\sigma^2 \\ &= \int_0^\infty \frac{1}{\sqrt{2\pi}} \underbrace{\sigma^{-2-\frac{1}{2}-1}}_{\text{a kernel for } \text{IG}\left(\frac{1}{2}, \frac{(x-\mu)^2}{2}\right)} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) d\sigma^2 \end{aligned}$$

$$= \frac{1}{\sqrt{2\pi}} \frac{\Gamma(\frac{1}{2})}{\left(\frac{(x-\mu)^2}{2}\right)^{\frac{1}{2}}} < \infty$$

$$\begin{aligned} \Rightarrow \quad \pi(\sigma^2 | x) &\propto f(x | \sigma^2) \pi(\sigma^2) \\ &\propto \underbrace{(\sigma^2)^{-\frac{1}{2}} \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right)}_{\text{kernel for } \text{IG}\left(\frac{1}{2}, \frac{(x-\mu)^2}{2}\right)} \cdot (\sigma^2)^{-1} \end{aligned}$$

$$\Rightarrow \quad \sigma^2 | x \sim \text{IG}\left(\frac{1}{2}, \frac{(x-\mu)^2}{2}\right)$$

$$\begin{aligned} (b) \quad f(y | x) &= \int_0^\infty \frac{1}{\sqrt{2\pi}\sigma^2} \exp\left(-\frac{(y-\rho x)^2}{2\sigma^2}\right) \cdot \frac{\left(\frac{(x-\mu)^2}{2}\right)^{\frac{1}{2}}}{\Gamma(\frac{1}{2})} (\sigma^2)^{-\frac{1}{2}-1} \\ &\quad \times \exp\left(-\frac{(x-\mu)^2}{2\sigma^2}\right) d\sigma^2 \\ &= \frac{1}{\sqrt{2\pi}} \cdot \frac{\left(\frac{(x-\mu)^2}{2}\right)^{\frac{1}{2}}}{\Gamma(\frac{1}{2})} \cdot \int_0^\infty \underbrace{(\sigma^2)^{-1-1} \exp\left(-\frac{(y-\rho x)^2 + (x-\mu)^2}{2\sigma^2}\right)}_{\text{a kernel for } \text{IG}\left(1, \frac{(y-\rho x)^2 + (x-\mu)^2}{2}\right)} d\sigma^2 \end{aligned}$$

$$= \frac{\left(\frac{(x-\mu)^2}{2}\right)^{1/2}}{\sqrt{2\pi} \Gamma(\frac{1}{2})} \cdot \frac{\Gamma(1)}{\left(\frac{(y-\mu x)^2 + (x-\mu)^2}{2}\right)^1}, \quad -\infty < y < \infty$$

$$\begin{aligned}
 2. \quad m(x) &= \int_{-\infty}^{\infty} \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n \cdot \exp\left(-\frac{S^2}{2\sigma^2} - \frac{n(\bar{x}-\theta)^2}{2\sigma^2}\right) \cdot 1 \, d\theta \\
 &= \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n \cdot \exp\left(-\frac{S^2}{2\sigma^2}\right) \cdot \underbrace{\int_{-\infty}^{\infty} \exp\left(-\frac{n(\bar{x}-\theta)^2}{2\sigma^2}\right) d\theta}_{\text{a kernel for } N(\bar{x}, \sigma^2/n)} \\
 &= \left(\frac{1}{\sqrt{2\pi\sigma^2}}\right)^n \exp\left(-\frac{S^2}{2\sigma^2}\right) \cdot \sqrt{2\pi \sigma^2/n} < \infty
 \end{aligned}$$

$S^2 = \sum_{i=1}^n (x_i - \bar{x})^2$

$$\Rightarrow \pi(\theta|x) \propto \underbrace{\exp\left(-\frac{n(\bar{x}-\theta)^2}{2\sigma^2}\right)}_{\text{a kernel for } N(\bar{x}, \sigma^2/n)}$$

$\delta^\pi(\bar{x})$ minimizes posterior expected loss

$$\begin{aligned}
 \rho(\pi, \delta | \bar{x}) &= E\left(L(\theta, \delta) | \bar{x}\right) = E\left(e^{c(\delta-\theta)} - c(\delta-\theta) - 1 | \bar{x}\right) \\
 &= E\left(e^{c(\delta-\theta)} | \bar{x}\right) - c \cdot (\delta - \bar{x}) - 1
 \end{aligned}$$

$$\begin{aligned}
 E(e^{c(\delta-\theta)} | \bar{x}) &= \int_{-\infty}^{\infty} e^{c(\delta-\theta)} \cdot \frac{1}{\sqrt{2\pi\sigma^2/n}} \exp\left(-\frac{(\theta-\bar{x})^2}{2\sigma^2/n}\right) d\theta \\
 &= \frac{e^{c\delta}}{\sqrt{2\pi\sigma^2/n}} \cdot \int_{-\infty}^{\infty} \exp\left(-\frac{\theta^2 - 2\theta\bar{x} + \bar{x}^2 + c2\sigma^2/n\theta}{2\sigma^2/n}\right) d\theta \\
 &= \frac{e^{c\delta}}{\sqrt{2\pi\sigma^2/n}} \cdot \int_{-\infty}^{\infty} \exp\left(-\frac{\theta^2 - 2(\bar{x} + \frac{c\sigma^2}{n})\theta + \bar{x}^2 + (\bar{x} + \frac{c\sigma^2}{n})^2}{2\sigma^2/n}\right) d\theta \\
 &\quad \underbrace{\hspace{10em}}_{\substack{\uparrow \\ \text{a kernel for } N(\bar{x} + \frac{c\sigma^2}{n}, \sigma^2/n)}} \\
 &= e^{c\delta} \cdot \exp\left(-\frac{\bar{x}^2 - (\bar{x} + \frac{c\sigma^2}{n})^2}{2\sigma^2/n}\right) \\
 &= e^{c\delta} \cdot \exp\left(-\frac{-\frac{2c\sigma^2}{n}\bar{x} + \frac{c^2\sigma^4}{n^2}}{2\sigma^2/n}\right) \\
 &\quad \underbrace{\hspace{10em}}_{\equiv A > 0}
 \end{aligned}$$

Then

$$p(\pi, \delta | \bar{x}) = A e^{c\delta} - c(\delta - \bar{x}) - 1$$

$$\frac{d p(\pi, \delta | \bar{x})}{d \delta} = c A e^{c\delta} - c$$

$$\frac{d^2 p(\pi, \delta | \bar{x})}{d \delta^2} = c^2 A e^{c\delta} > 0$$

$$\Rightarrow \text{let } c A e^{c\delta} - c = 0$$

$$\exp\left(c\delta - \frac{-2c\sigma^2\bar{x} + \frac{c^2\sigma^4}{n}}{2\sigma^2}\right) = 1$$

$$c\delta = \frac{2\sigma^2\bar{x}}{2\sigma^2} - \frac{c\sigma^4}{2\sigma^2 n} = \boxed{\bar{x} - \frac{c\sigma^2}{2n}}$$

$$\Rightarrow \delta^\pi(\bar{x}) = \bar{x} - \frac{c\sigma^2}{2n}$$

3. $x_i \stackrel{\text{iid}}{\sim} \text{Ber}(\theta), i=1, \dots, n$ & $\theta \sim \text{Be}(\alpha, \beta)$

Since the prior is proper,

$$\begin{aligned} \text{a. } \pi(\theta | x_1, \dots, x_n) &\propto \theta^{\sum x_i} (1-\theta)^{n-\sum x_i} \cdot \theta^{\alpha-1} (1-\theta)^{\beta-1} \\ &= \theta^{\alpha+\sum x_i-1} (1-\theta)^{n-\sum x_i+\beta-1} \end{aligned}$$

$$\Rightarrow \pi(\theta | \sum x_i) \text{ is } \text{Be}(\alpha + \sum x_i, n - \sum x_i + \beta)$$

Under squared error loss function,

$$\delta^\pi(\sum x_i) = E(\theta | \sum x_i) = \frac{\alpha + \sum x_i}{\alpha + \beta + n}$$

$$\begin{aligned} \text{b. } R(\theta, \delta^\pi) &= E \left(\left(\frac{\alpha + \sum x_i}{\alpha + \beta + n} - \theta \right)^2 \mid \theta \right) \\ &= E \left(\frac{(\alpha + \sum x_i) - (\alpha + \beta + n)\theta}{(\alpha + \beta + n)^2} \mid \theta \right) \\ &= E \left(\frac{(y - n\theta + (\alpha - (\alpha + \beta)\theta))^2}{((\alpha + \beta + n)^2)} \mid \theta \right) \end{aligned}$$

$$\begin{aligned} y &\sim \text{Bin}(n, \theta) & E(y) &= n\theta \\ & & \text{Var}(y) &= n\theta(1-\theta) \end{aligned}$$

$$= b^2 \text{Var}(\bar{X}) + (a + b\theta - \theta)^2$$

$$= \frac{b^2}{n} \theta(1-\theta) + (a + (b-1)\theta)^2$$

$$= \frac{n\theta(1-\theta) + (\alpha + (\alpha+\beta)\theta)^2}{(\alpha+\beta+n)^2}$$

$$= \frac{\alpha^2 + (n - \alpha(\alpha+\beta))\theta + ((\alpha+\beta)^2 - n)\theta^2}{(\alpha+\beta+n)^2}$$

expectation with $\pi(\theta) = \text{Be}(\alpha, \beta)$. $E(\theta^k) = \frac{B(\alpha+k, \beta)}{B(\alpha, \beta)}$ is given.

$$c. \quad r(\pi, \delta) = E^\pi \left(R(\theta, \delta^\pi) \right)$$

$$= E^\pi \left(\frac{\alpha^2 + (n - \alpha(\alpha+\beta))\theta + ((\alpha+\beta)^2 - n)\theta^2}{(\alpha+\beta+n)^2} \right)$$

$$= \frac{1}{(\alpha+\beta+n)^2} \left\{ \alpha^2 + (n - \alpha(\alpha+\beta)) \cdot \frac{\alpha}{\alpha+\beta} + ((\alpha+\beta)^2 - n) \cdot \frac{\alpha(\alpha+1)}{(\alpha+\beta+1)(\alpha+\beta)} \right\}$$

Enough for
the full credit



$$= \frac{\alpha (n\beta + (\alpha+1)(\alpha+\beta)^2)}{(\alpha+\beta+n)^2 (\alpha+\beta)(\alpha+\beta+1)}$$

d. $\text{Be}(\alpha, \beta)$ is strictly positive for $\theta \in (0, 1)$

$$r(\pi, \delta) < \infty$$

$R(\theta, \delta)$ is continuous of θ

$\xrightarrow{\text{so}}$ δ is admissible by prop 2.4, 22

(the only proposition we covered in class)