

Mathematical Exercises 2

Problem 1

$$\sigma^2 \sim \text{Inv-}\chi^2(v_0, \sigma_0^2), \quad \theta \text{ is known}$$

Priors $p(\sigma^2) \propto \frac{\exp\left[-\frac{v_0 \sigma_0^2}{2\sigma^2}\right]}{\sigma^2(1+v_0/2)}$

$$X = (x_1, \dots, x_n)$$

Likelihood:

$$p(x|\sigma^2, \theta) = \prod_{i=1}^n \frac{1}{(2\pi\sigma^2)^{1/2}} \cdot \exp\left[-\frac{1}{2\sigma^2}(x_i - \theta)^2\right]$$

$$\propto \frac{1}{(\sigma^2)^{n/2}} \cdot \exp\left[-\frac{ns^2}{2\sigma^2}\right], \quad \text{where } s^2 = \frac{\sum_{i=1}^n (x_i - \theta)^2}{n}$$

Posterior:

$$p(\sigma^2|x, \theta) \propto p(x|\sigma^2, \theta) p(\sigma^2) \propto \frac{\exp\left[-\frac{(v_0 \sigma_0^2 + ns^2)}{2\sigma^2}\right]}{\sigma^2 \left(1 + \underbrace{\frac{v_0}{2} + \frac{n}{2}}_{(v_0+n)/2}\right)},$$

i.e., $\sigma^2|x, \theta \sim \text{Inv-}\chi^2(v_n, \sigma_n^2)$

$$v_n = v_0 + n$$

$$v_n \sigma_n^2 = v_0 \sigma_0^2 + ns^2 \Leftrightarrow \sigma_n^2 = \frac{v_0 \sigma_0^2 + ns^2}{v_n} = \frac{v_0 \sigma_0^2 + ns^2}{v_0 + n}$$

b) Non-informative prior: let $v_0 \rightarrow 0 \Rightarrow p(\sigma^2) \propto \sigma^{-1/2}, n=3, \theta=1$

$$s^2 = \frac{\sum_{i=1}^3 (x_i - \theta)^2}{n} = 1.68$$

So, $\sigma^2|x, \theta \sim \text{Inv-}\chi^2(3, 1.68)$

Problem 2 , Posterior predictive distribution of x_{n+1} , given $x_{1:n}$:

$$a) \quad P(x_{n+1} | x_{1:n}) = \int P(x_{n+1} | \theta) \cdot P(\theta | x_{1:n}) d\theta$$

$$P(x_{n+1} | \theta) = \theta^{x_{n+1}} \cdot (1-\theta)^{1-x_{n+1}}$$

$$\theta | x_{1:n} \sim \text{Beta}(\alpha+s, \beta+f), \text{ so } P(\theta | x_{1:n}) = \frac{\Gamma(\alpha+s+\beta+f)}{\Gamma(\alpha+s)\Gamma(\beta+f)} \cdot \theta^{\alpha+s-1} \cdot (1-\theta)^{\beta+f-1}$$

$$\begin{aligned} P(x_{n+1} | x_{1:n}) &= \frac{\Gamma(\alpha+\beta+n)}{\Gamma(\alpha+s)\Gamma(\beta+f)} \cdot \int \theta^{x_{n+1}+\alpha+s-1} (1-\theta)^{1-x_{n+1}+\beta+f-1} d\theta \\ &= B(\alpha', \beta') = \frac{\Gamma(\alpha')\Gamma(\beta')}{\Gamma(\alpha'+\beta')} \\ &= \frac{\Gamma(\alpha')\Gamma(\beta')}{\Gamma(\alpha+s)\Gamma(\beta+f)(\alpha+\beta+n)} \end{aligned}$$

$$\text{So, } x_{n+1} = 1 \Rightarrow P(x_{n+1} = 1 | x_{1:n}) = \frac{\Gamma(\alpha+s+1)}{\Gamma(\alpha+s)(\alpha+\beta+n)} = \frac{\alpha+s}{\alpha+\beta+n}$$

$$\bullet \quad x_{n+1} = 0 \Rightarrow P(x_{n+1} = 0 | x_{1:n}) = \frac{\alpha+\beta+n-(\alpha+s)}{\alpha+\beta+n} = \frac{\beta+f}{\alpha+\beta+n}$$

$$b) \quad \theta \sim \text{Beta}(\alpha=1, \beta=1), \quad n=10, \quad s=8$$

$$P(x_{11} = 1 | x_{1:10}) = \frac{1+8}{1+1+10} = 0.75 \Rightarrow P(x_{11} = 0 | x_{1:10}) = 0.25$$

$$\begin{aligned} EU_{\text{bring}} &= 0.75 \cdot 20 + 0.25 \cdot 10 = 17.5 \\ EU_{\text{leave}} &= 0.75 \cdot 50 - 0.25 \cdot 50 = 25.0 \end{aligned} \quad \left. \vphantom{\begin{aligned} EU_{\text{bring}} \\ EU_{\text{leave}} \end{aligned}} \right\} \text{leave the umbrella}$$

c) Change the values for α and β and calculate new EU_{bring} and EU_{leave}