

高. 汪. 汪. 2019010448 吴叶斯 HW2

1.

$$P(Y|\theta) = \theta e^{-\theta Y}, \quad \log P(Y|\theta) = \log \theta - \theta Y$$

$$d^2 \log P(Y|\theta) / d\theta^2 = -\frac{1}{\theta^2}$$

$$\therefore J(\theta) = -E[d^2 \log P(Y|\theta) / d\theta^2 | \theta] \propto \frac{1}{\theta^2}$$

$$p(\theta) \propto J(\theta)^{1/2} \propto \frac{1}{\theta}$$

2.

$$a) P(\theta_1, \theta_2 | Y) \propto \theta_1^{\alpha_1 + Y_1 - 1} \cdot \theta_2^{\alpha_2 + Y_2 - 1} \cdot (1 - \theta_1 - \theta_2)^{\sum_{i=3}^J \alpha_i + Y_i - 1}$$

$$\alpha = \frac{\theta_1}{\theta_1 + \theta_2}, \quad \beta = \theta_1 + \theta_2$$

$$\left| \frac{\partial(\alpha, \beta)}{\partial(\theta_1, \theta_2)} \right| = \begin{vmatrix} \frac{\theta_2}{(\theta_1 + \theta_2)^2} & 1 \\ -\frac{\theta_1}{(\theta_1 + \theta_2)^2} & 1 \end{vmatrix} = \frac{1}{\theta_1 + \theta_2} = \frac{1}{\beta}$$

$$\therefore p(\alpha, \beta | Y) \propto \beta(\alpha, \beta)^{\alpha_1 + Y_1 - 1} \cdot [(1 - \alpha)\beta]^{\alpha_2 + Y_2 - 1} \cdot (1 - \beta)^{\sum_{i=3}^J (\alpha_i + Y_i) - 1}$$

$$= \alpha^{\alpha_1 + Y_1 - 1} \cdot (1 - \alpha)^{\alpha_2 + Y_2 - 1} \cdot \beta^{\alpha_1 + Y_1 + \alpha_2 + Y_2 - 1} \cdot (1 - \beta)^{\sum_{i=3}^J (\alpha_i + Y_i) - 1}$$

$$\propto \text{Beta}(\alpha | \alpha_1 + Y_1, \alpha_2 + Y_2) \cdot \text{Beta}(\beta | \alpha_1 + Y_1 + \alpha_2 + Y_2, \sum_{i=3}^J \alpha_i + Y_i)$$

$$\therefore p(\alpha | Y) \propto \text{Beta}(\alpha | \alpha_1 + Y_1, \alpha_2 + Y_2)$$

b) 对于先验 Beta(α_1, α_2)

$$p(\alpha | Y) \propto p(Y | \alpha) \cdot p(\alpha)$$

$$\propto \alpha^{Y_1} \cdot (1 - \alpha)^{Y_2} \cdot \alpha^{\alpha_1 - 1} \cdot (1 - \alpha)^{\alpha_2 - 1}$$

$$= \alpha^{Y_1 + \alpha_1 - 1} (1 - \alpha)^{Y_2 + \alpha_2 - 1} \propto \text{Beta}(\alpha | \alpha_1 + Y_1, \alpha_2 + Y_2)$$

3.

$$P(\mu, \sigma^2 | Y) \propto P(Y | \mu, \sigma^2) \cdot p(\mu, \sigma^2)$$

$$\propto (\sigma^2)^{-n/2} \exp \left\{ -\frac{(n-1)S^2 + n(\mu - \bar{y})^2}{2\sigma^2} \right\}$$

$$\cdot \sigma^{-1} (\sigma^2)^{-(U_0/2+1)} \exp \left\{ \frac{1}{2\sigma^2} [U_0 \sigma_0^2 + k_0 (\mu - \mu_0)^2] \right\}$$

$$\propto \text{Inv-}\chi^2 \left(\frac{\mu_0 k_0 + n \bar{y}}{h + k_0}, \frac{\sigma_0^2}{h + k_0}; h + U_0, \sigma_h^2 \right)$$

$$\text{其中 } \sigma_h^2 = \frac{U_0 \sigma_0^2 + (n-1)S^2 + \frac{n k_0 (\bar{y} - \mu_0)^2}{h + k_0}}{h + U_0}$$

对于 noninformative Prior

$$P(\mu, \sigma^2) \propto (\sigma^2)^{-1}$$

$$\Rightarrow U_0 = -1, \sigma_0 = 0, k_0 = 0$$

$$P(\mu, \sigma^2) \propto (\sigma^2)^{-1}$$

$$p(\mu, \sigma^2 | Y) \propto (\sigma^2)^{-(n/2+1)} \exp \left\{ -\frac{(n-1)S^2 + n(\mu - \bar{y})^2}{2\sigma^2} \right\}$$

$$\therefore \sigma^2 | \mu, Y \sim \text{Inv-}\chi^2 \left(n, \frac{(n-1)S^2 + n(\mu - \bar{y})^2}{n} \right)$$