

**Question 1**

(a)

$$\begin{aligned}
 f(x) &= \prod_{i=1}^n [\theta x_i^{\theta-1} I_{(0,1)}(x_i)] \\
 &= \theta^n \left( \prod_{i=1}^n x_i \right)^{\theta-1} \prod_{i=1}^n I_{(0,1)}(x_i) \\
 &= \left[ I_{(0,1)}(\min\{x_1, \dots, x_n\}) I_{(0,1)}(\max\{x_1, \dots, x_n\}) \right] \left[ \left( \prod_{i=1}^n x_i \right)^{\theta-1} \theta^n \right]
 \end{aligned}$$

By Neyman's factorization theorem,  $T = \prod_{i=1}^n x_i$  is the sufficient statistics for  $\theta$ .

(b)

$$\begin{aligned}
 f(x) &= \prod_{i=1}^n \theta a x_i^{a-1} \exp\{-\theta x_i^a\} I_{(0,1)}(x_i) \\
 &= (\theta a)^n \left( \prod_{i=1}^n x_i \right)^{a-1} \exp\{-\theta \sum_{i=1}^n x_i^a\} I_{(0,1)}(\min\{x_1, \dots, x_n\}) \\
 &= \left[ a^n \left( \prod_{i=1}^n x_i \right)^{a-1} I_{(0,1)}(\min\{x_1, \dots, x_n\}) \right] \left[ \theta^n \exp\{-\theta \sum_{i=1}^n x_i^a\} \right]
 \end{aligned}$$

By Neyman's factorization theorem,  $T = \sum_{i=1}^n x_i^a$  is the sufficient statistics for  $\theta$ .

(c)

$$\begin{aligned}
 f(x) &= \prod_{i=1}^n \frac{\theta a^\theta}{x_i^{\theta+1}} I_{(a,\infty)}(x_i) \\
 &= \theta^n a^{n\theta} \frac{1}{(\prod_{i=1}^n x_i)^{\theta+1}} I_{(a,\infty)}(\min\{x_1, \dots, x_n\}) \\
 &= \left[ I_{(a,\infty)}(\min\{x_1, \dots, x_n\}) \right] \left[ \frac{\theta^n a^{n\theta}}{(\prod_{i=1}^n x_i)^{\theta+1}} \right]
 \end{aligned}$$

By Neyman's factorization theorem,  $T = \prod_{i=1}^n x_i$  is the sufficient statistics for  $\theta$ .

### Question 2

(a) For any  $\theta \in \Theta$ , the posterior distribution of  $\theta|x$  is:

$$\frac{f(x|\theta)\pi(\theta)}{\sum_{\theta_i \in \Theta} f(x|\theta_i)\pi(\theta_i)} \quad \text{where } \pi(\theta) \text{ is any prior on } \theta$$

By the condition, we know it equals to some function  $g(\theta, T(x))$ , i.e.

$$\frac{f(x|\theta)\pi(\theta)}{\sum_{\theta_i \in \Theta} f(x|\theta_i)\pi(\theta_i)} = g(\theta, T(x))$$

where  $g(x, T(x))$  is a function of  $\theta$  and  $T(x)$  only. Thus

$$f(x|\theta) = \frac{g(x, T(x))}{\pi(\theta)} \sum_{\theta_i \in \Theta} f(x|\theta_i)\pi(\theta_i)$$

By factorization theorem,  $T(x)$  is sufficient for  $\theta$ .

(b) If  $T(x)$  is sufficient, then  $f(x|\theta)$  can be written as

$$f(x|\theta) = g(\theta, T(x))h(x)$$

Let  $\pi(\theta)$  be an arbitrary prior distribution, then the posterior of  $\theta$  is

$$\frac{f(x|\theta)\pi(\theta)}{\sum_{\theta_i \in \Theta} f(x|\theta_i)\pi(\theta_i)} = \frac{g(\theta, T(x))}{\sum_{\theta_i \in \Theta} g(\theta_i, T(x))\pi(\theta_i)} \pi(\theta)$$

The posterior depends on  $x$  only through  $T(x)$ . By factorization theorem,  $T(x)$  is sufficient for  $\theta$ .

### Question 3 (Problem 6.7)

(a)  $\theta = \{\xi, \eta, \sigma, \tau : -\infty < \xi, \eta < \infty, 0 < \sigma, \tau\}$  The parameter space

contains 4-dimensional rectangle space.

$$\begin{aligned}
f(x) &= \prod_{i=1}^m \frac{1}{\sqrt{2\pi\sigma^2}} e^{-\frac{1}{2\sigma^2}(x_i-\xi)^2} \\
&= \frac{1}{(2\pi)^{m/2}} \exp \left\{ -\frac{1}{2\sigma^2} \sum x_i^2 + \frac{\xi}{\sigma^2} \sum x_i - \frac{m\xi^2}{2\sigma^2} - m \log \sigma \right\} \\
f(y) &= \frac{1}{(2\pi)^{n/2}} \exp \left\{ -\frac{1}{2\tau^2} \sum y_i^2 + \frac{\eta}{\tau^2} \sum y_i - \frac{n\eta^2}{2\tau^2} - n \log \tau \right\}
\end{aligned}$$

Thus,  $(x,y)$  is distributed as  $f(x)f(y)$  with parameter space  $\theta = \{\xi, \eta, \sigma, \tau : -\infty < \xi, \eta < \infty, 0 < \sigma, \tau\}$ . The sufficient statistics  $T = \{\sum x_i, \sum x_i^2, \sum y_i, \sum y_i^2\}$  is also minimal since  $(x,y)$  belongs to exponential family.

(b) If  $\sigma = \tau$ , then the parameter space becomes

$$\Theta = \{\xi, \eta, \sigma : -\infty < \xi, \eta < \infty, 0 < \sigma < \infty\}$$

which contains a 3-dimensional triangle in it. The joint distribution of  $(x,y)$  is

$$\begin{aligned}
f(x, y) &= \frac{1}{(2\pi)^{(m+n)/2}} \exp \left\{ -\frac{1}{2\sigma^2} \left( \sum x_i^2 + \sum y_i^2 \right) \right. \\
&\quad \left. + \frac{\xi}{\sigma^2} \sum x_i + \frac{\eta}{\sigma^2} \sum y_i - \frac{m\xi^2 + n\eta^2}{2\sigma^2} - (m+n) \log \sigma \right\}
\end{aligned}$$

with natural parameter

$$\{\eta_1, \eta_2, \eta_3 : \eta_1 = \frac{1}{2\sigma^2}, \eta_2 = \frac{\xi}{\sigma^2}, \eta_3 = \frac{\eta}{\sigma^2}\}$$

This distribution belongs to exponential family with full rank. By theorem 6.22,  $\{\sum x_i^2 + \sum y_i^2, \sum x_i, \sum y_i\}$  is complete.

(c)  $\xi = \eta$  and  $\xi, \sigma, \eta$  are arbitrary. The joint distribution of  $(x,y)$  is

$$\begin{aligned}
f(x, y) &= \frac{1}{(2\pi)^{(m+n)/2}} \exp \left\{ -\frac{1}{2\sigma^2} \sum x_i^2 - \frac{1}{2\tau^2} \sum y_i^2 \right. \\
&\quad \left. + \frac{\xi}{\sigma^2} \sum x_i + \frac{\eta}{\tau^2} \sum y_i - \frac{m\xi^2}{2\sigma^2} - m \log \sigma - \frac{n\eta^2}{2\tau^2} - n \log \tau \right\}
\end{aligned}$$

This belongs to exponential family which is full rank. Therefore, the minimal sufficient statistics is  $(\sum x_i, \sum x_i^2, \sum y_i, \sum y_i^2)$ .

To show that the parameter space is full rank, let

$$\eta_1 = -\frac{1}{2\sigma^2}, \eta_2 = -\frac{1}{2\tau^2}, \eta_3 = \frac{\xi}{\sigma^2}, \eta_4 = \frac{\eta}{\tau^2}$$

The natural parameter space is

$$\Theta = \{(\eta_1, \eta_2, \eta_3, \eta_4) : \eta_1 < 0, \eta_2 < 0, \frac{\eta_3}{\eta_4} = \frac{\eta_1}{\eta_2}\}$$

If we pick

$$\begin{aligned} \eta^{(1)} &= (-1/2, -1/2, 1, 1) \\ \eta^{(2)} &= (-1/4, -1/4, 2, 2) \\ \eta^{(3)} &= (-1/9, -1/3, 3, 9) \\ \eta^{(4)} &= (-1/16, -1/16, 4, 16) \\ \eta^{(5)} &= (-1/25, -1/25, 5, 5) \end{aligned}$$

then the  $4 \times 4$  matrix

$$\det \begin{pmatrix} \eta^{(2)} - \eta^{(1)} \\ \eta^{(3)} - \eta^{(1)} \\ \eta^{(4)} - \eta^{(1)} \\ \eta^{(5)} - \eta^{(1)} \end{pmatrix} \neq 0$$

Thus it's full rank.

#### Question 4 (Problem 6.20)

(a)

$$\begin{aligned} p_\theta(x) &= \frac{1}{\sqrt{2\pi\theta^2}} \exp \left\{ -\frac{1}{2\theta} \left( \sum x_i^2 - 2\theta \sum x_i + n\theta^2 \right) \right\} \\ &= \frac{1}{\sqrt{2\pi\theta^2}} \exp \left\{ -\frac{1}{\theta} \left( \frac{1}{2} \sum x_i^2 \right) - \sum x_i + \frac{n}{2}\theta \right\} \end{aligned}$$

$T_1(x) = (\sum x_i, \sum x_i^2)$  is sufficient. Consider  $T_2(x) = \sum x_i^2$  which is sufficient statistic also, but for  $x \neq y \exists T_2(x) = T_2(y)$  s.t.  $T_1(x) \neq T_1(y)$ . Thus  $T_1(x)$  is not minimal.

(b) To show  $(\sum x_i, \sum x_i^2, \sum x_i^3)$  is minimal sufficient, we need to show that

$$\exists \eta^{(j)} = (\eta_1^{(j)}, \eta_2^{(j)}, \eta_3^{(j)}), \quad j = 0, 1, 2, 3$$

such that

$$\left( \frac{p(x, \eta^{(1)})}{p(x, \eta^{(0)})}, \frac{p(x, \eta^{(2)})}{p(x, \eta^{(0)})}, \frac{p(x, \eta^{(3)})}{p(x, \eta^{(0)})} \right)$$

is a one-to-one transformation of  $T = (\sum x_i, \sum x_i^2, \sum x_i^3)$ .

From the pdf:

$$C \exp(-n\theta^4) \exp\left(4\theta^3 \sum x_i - 6\theta^2 \sum x_i^2 + 4\theta \sum x_i^3 - \sum x_i^4\right)$$

we know the natural parameters are

$$4\theta^3 = \eta_1, -6\theta^2 = \eta_2, 4\theta = \eta_3$$

and

$$\left( \sum_{i=1}^3 (\eta_i^{(1)} - \eta_i^{(0)}) T_i(x), \sum_{i=1}^3 (\eta_i^{(2)} - \eta_i^{(0)}) T_i(x), \sum_{i=1}^3 (\eta_i^{(3)} - \eta_i^{(0)}) T_i(x) \right)$$

is minimal sufficient for any  $\eta^{(0)}, \eta^{(1)}, \eta^{(2)}, \eta^{(3)} \in \Theta$ .

It can also be written as

$$\begin{pmatrix} \eta_1^{(1)} - \eta_1^{(0)} & \eta_2^{(1)} - \eta_2^{(0)} & \eta_3^{(1)} - \eta_3^{(0)} \\ \eta_1^{(2)} - \eta_1^{(0)} & \eta_2^{(2)} - \eta_2^{(0)} & \eta_3^{(2)} - \eta_3^{(0)} \\ \eta_1^{(3)} - \eta_1^{(0)} & \eta_2^{(3)} - \eta_2^{(0)} & \eta_3^{(3)} - \eta_3^{(0)} \end{pmatrix} \begin{pmatrix} T_1 \\ T_2 \\ T_3 \end{pmatrix}$$

Let

$$\eta^{(0)} = (0, 0, 0), \eta^{(1)} = (1/2, -3/2, 2), \eta^{(2)} = (4, -6, 4), \eta^{(3)} = (32, -24, 8)$$

then it's easy to see that  $(\eta^{(1)} - \eta^{(0)}, \eta^{(2)} - \eta^{(0)}, \eta^{(3)} - \eta^{(0)})^t$  span  $E_3$  and the determinant is  $72 \neq 0$ . Thus,  $(T_1, T_2, T_3)$  is minimal sufficient.

### Question 5 (Problem 6.21)

Proof:  $T = (\sum x_i, \sum x_i^2)$

$$E\left(\frac{\sum x_i}{n}\right) = E(\bar{x}) = \text{Var}(\bar{x}) + [E(\bar{x})]^2 = \frac{\xi^2}{n} + \xi^2 = \frac{n+1}{n}\xi^2$$

$$\Rightarrow E\left[\frac{n}{n+1}\left(\frac{\sum x_i}{n}\right)^2\right] = \xi^2$$

$$E\left(\sum x_i^2\right) = nE(x_i^2) = 2n\xi^2 \quad \Rightarrow \quad E\left(\frac{\sum x_i^2}{2n}\right) = \xi^2$$

$$\Rightarrow E f(T(x)) = 0 \quad \text{where} \quad f(T(x)) = \frac{n}{n+1}\left(\frac{\sum x_i}{n}\right)^2 - \frac{\sum x_i^2}{2n}$$

But  $f(T) \neq 0$  a.e., thus  $T = (\sum x_i, \sum x_i^2)$  is not complete.