

# Statistical Inference Assignment 3

Junhao Yuan (20307130129)

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PROBLEM 1.

Let  $X_1, \dots, X_n$  be independent random variables with densities

$$f_{X_i}(x|\theta) = e^{i\theta-x} \mathbb{I}(x \geq i\theta). \quad (1)$$

Prove that  $T = \min_i (X_i/i)$  is a sufficient statistics for  $\theta$ .

SOLUTION.

The joint pdf of the samples  $X$  is

$$\begin{aligned} f_X(\mathbf{x}|\theta) &= \prod_{i=1}^n e^{i\theta-x_i} \mathbb{I}(x_i \geq i\theta) \\ &= \exp\left(\frac{n(n+1)}{2}\theta - \sum_{i=1}^n x_i\right) \mathbb{I}(T(\mathbf{x}) \geq 0) \\ &= g(T(\mathbf{x})|\theta)h(\mathbf{x}), \end{aligned}$$

where

$$\begin{aligned} g(T(\mathbf{x})|\theta) &= e^{\frac{n(n+1)}{2}\theta} \mathbb{I}(T(\mathbf{x}) \geq 0), \\ h(\mathbf{x}) &= \exp^{-\sum_{i=1}^n x_i}. \end{aligned}$$

Here, we use the fact that  $i > 0$  and  $x_i\theta \geq i\theta$  (for  $1 \leq i \leq n$ ) if and only if  $\min_i (x_i/i) \geq \theta$ . Thus, by the factorization theorem, we have  $T(X) = \min_i (X_i/i)$  is a sufficient statistics.

PROBLEM 2.

Suppose  $X_1, \dots, X_n \stackrel{iid}{\sim} N(\mu, \sigma^2)$ , show that  $\bar{X}$  and  $X_{(n)} - X_{(1)}$  are independent.

SOLUTION.

First consider  $\sigma^2$  fixed and let  $\mu$  vary,  $-\infty < \mu < \infty$ . It's obvious that  $\bar{X}$  is a sufficient statistics for  $\mu$ . Since the pdf of  $\bar{X}$  is ( $\sigma$  is known)

$$\begin{aligned} f_{\bar{X}}(\bar{x}) &= \sqrt{\frac{n}{2\pi}} \frac{1}{\sigma} e^{-\frac{n(\bar{x}-\mu)^2}{2\sigma^2}} \\ &= \sqrt{\frac{n}{2\pi}} e^{-\frac{n\bar{x}^2}{2\sigma^2}} \cdot \frac{1}{\sigma} e^{-\frac{n\mu^2}{2\sigma^2}} \cdot e^{\frac{n\mu}{\sigma^2}\bar{x}}, \end{aligned}$$

which has an open set in  $\mathbb{R}$ , we can know that  $\bar{x}$  is complete. Thus,  $\bar{X}$  is a complete and sufficient statistics for  $\mu$ . Since  $X_{(n)} - X_{(1)}$  is an ancillary statistics for  $\mu$  when  $\sigma$  is known, by the Basu's theorem, we know that  $\bar{X}$  and  $X_{(n)} - X_{(1)}$  are independent. But since  $\sigma$  is arbitrary, we have that  $\bar{X}$  and  $X_{(n)} - X_{(1)}$  are independent for any choice of  $\mu$  and  $\sigma$ .

**PROBLEM 3.**

Let  $X_1, \dots, X_n$  be a random sample from a population with location pdf  $f(x - \theta)$ . Show that the order statistics,  $T(X_1, \dots, X_n) = (X_{(1)}, \dots, X_{(n)})$ , are a sufficient Statistics for  $\theta$  and no further reduction is possible.

**SOLUTION.**

The pdf of  $X$  is

$$\begin{aligned} f(\mathbf{x}|\theta) &= \prod_{i=1}^n f(x_i|\theta) \\ &= \prod_{i=1}^n f(x_{(i)}|\theta), \end{aligned}$$

which shows that  $T(X) = (X_{(1)}, \dots, X_{(n)})$  is a sufficient statistics. Consider two samples  $X_1, \dots, X_n$  and  $Y_1, \dots, Y_n$ , then the ratio is

$$\begin{aligned} \frac{f(\mathbf{x}|\theta)}{f(\mathbf{y}|\theta)} &= \frac{\prod_{i=1}^n f(x_i|\theta)}{\prod_{j=1}^n f(y_j|\theta)} \\ &= \prod_{i=1}^n \frac{f(x_{(i)}|\theta)}{f(y_{(i)}|\theta)}. \end{aligned}$$

It's obvious that the ratio is a constant function of  $\theta$  is and only if  $T(X) = T(Y)$ . Thus,  $T(X)$  is a minimal sufficient for  $\theta$ , which means no further reduction is possible without further restrictions on  $f$ .

**PROBLEM 4.**

Suppose  $X$  follows a discrete distribution with the following pmf:

$$P(X = 0) = p, \quad P(X = 1) = 3p, \quad P(X = 2) = 1 - 4p, \quad 0 < p < 1/4.$$

Is the family of distribution of  $X$  complete? What about the following family:

$$P(X = 0) = p, \quad P(X = 1) = p^2, \quad P(X = 2) = 1 - p - p^2, \quad 0 < p < 1/2.$$

**SOLUTION.**

Let  $g$  be a function such that  $\mathbb{E}_p g(X) = 0$ . Then

$$\begin{aligned} 0 &= \mathbb{E}_p g(X) = pg(0) + 3pg(1) + (1 - 4p)g(2) \\ &= (g(0) + 3g(1) - 4g(2)) + g(2), \end{aligned}$$

for all  $0 < p < 1/4$ . Thus, we have

$$\begin{aligned} g(2) &= 0 \\ g(0) + 3g(1) - 4g(2) &= 0, \end{aligned}$$

which implies that  $g(X)$  need to statistics that  $g(2) = 0$  and  $g(0) + 3g(1) = 0$ . Thus,  $g(X)$  don't have to be identically zero, and this family of distribution of  $X$  is not complete.

For the second distribution, we have

$$\begin{aligned} 0 = \mathbb{E}_p g(X) &= pg(0) + p^2g(1) + (1 - p - p^2)g(2) \\ &= (g(1) - g(2))p^2 + (g(0) - g(2))p + g(2), \end{aligned}$$

for all  $0 < p < 1/2$ . This is a polynomial of degree 2 in  $p$ . To make it zero for all  $p$ , each coefficient must be zero. Thus,

$$g(0) = g(1) = g(2) = 0,$$

or in other words,

$$P_p(g(X) = 0) = 1.$$

Thus, we have this family of distribution is complete.