

HW1

1.5-4

(a). Show that T_1 and T_2 are equivalent statistics if, and only if, we can write $T_2 = H(T_1)$ for some 1-1 transformation H of the range of T_1 into the range of T_2 . Which of the following statistics are equivalent? (Prove or disprove.)

If $T_2 = H(T_1)$ for some 1-1 transformation H of the range of T_1 into the range of T_2 , then

when $T_1(x) = T_1(y)$, $T_2(x) = H(T_1(x)) = H(T_1(y)) = T_2(y)$;

when $T_2(x) = T_2(y)$, $H(T_1(x)) = T_2(x) = T_2(y) = H(T_1(y))$; then T_1 and T_2 are equivalent.

If T_1 and T_2 are equivalent, then $\exists H$ make $T_2 = H(T_1)$ is a 1-1 transformation of the range of T_1 into the range of T_2 .

Therefore, T_1 and T_2 are equivalent statistics $\iff T_2 = H(T_1)$. ■

(b). $\prod_{i=1}^n x_i$ and $\sum_{i=1}^n \log x_i$, $x_i > 0$

$T_2(x) = \sum_{i=1}^n \ln x_i = \ln(\prod_{i=1}^n x_i) = \ln(T_1)$, $x_i > 0$. $H(x) = \ln x$ is a 1-1 transformation of $T_1 \in (0, \infty)$ into $T_2 \in (-\infty, \infty)$.

Thus, T_1 and T_2 are equivalent. ■

(c). $\sum_{i=1}^n x_i$ and $\sum_{i=1}^n \log x_i$, $x_i > 0$

$T_2(x) = \sum_{i=1}^n \ln x_i = T_1(\ln(x)) \neq T_1(x)$, $x_i > 0$.

There is not a H that can do a 1-1 transformation of the range of T_1 into the range of T_2 .

Thus, T_1 and T_2 are not equivalent. ■

(d). $(\sum_{i=1}^n x_i, \sum_{i=1}^n x_i^2)$ and $(\sum_{i=1}^n x_i, \sum_{i=1}^n (x_i - \bar{x})^2)$

Let $T_1 = (T_{11} = \sum_{i=1}^n x_i, T_{12} = \sum_{i=1}^n x_i^2)$, then

$$T_{21} = \sum_{i=1}^n x_i = T_{11}$$

$$T_{22} = \sum_{i=1}^n (x_i - \bar{x})^2 = \sum_{i=1}^n x_i^2 - 2\bar{x} \sum_{i=1}^n x_i + n(\bar{x})^2 = \sum_{i=1}^n x_i^2 - \frac{2}{n} (\sum_{i=1}^n x_i)^2 + \frac{1}{n} (\sum_{i=1}^n x_i)^2 = T_{12} - \frac{1}{n} T_{11}^2$$

H is a 1-1 transformation of the range of T_1 into the range of T_2 . Thus, T_1 and T_2 are equivalent. ■

(e). $(\sum_{i=1}^n x_i, \sum_{i=1}^n x_i^3)$ and $(\sum_{i=1}^n x_i, \sum_{i=1}^n (x_i - \bar{x})^3)$

Let $T_1 = (T_{11} = \sum_{i=1}^n x_i, T_{12} = \sum_{i=1}^n x_i^3)$, then

$$T_{21} = \sum_{i=1}^n x_i = T_{11}$$

$$T_{22} = \sum_{i=1}^n (x_i - \bar{x})^3 = \sum_{i=1}^n x_i^3 - 3\bar{x} \sum_{i=1}^n x_i^2 + 3\bar{x}^2 \sum_{i=1}^n x_i - n(\bar{x})^3 =$$

$$\sum_{i=1}^n x_i^3 - \frac{3}{n} \sum_{i=1}^n x_i \sum_{i=1}^n x_i^2 + \frac{3}{n^2} \left(\sum_{i=1}^n x_i \right)^3 - \frac{1}{n^2} \left(\sum_{i=1}^n x_i \right)^3 = T_{12} - \frac{3}{n} T_{11} \sum_{i=1}^n x_i^2 + \frac{2}{n^2} T_{11}^3$$

There is not statistics in T_1 can represent $\sum_{i=1}^n x_i^2$. There is not a H that can do a 1-1 transformation of the range of T_1 into the range of T_2 . Thus, T_1 and T_2 are not equivalent. ■

1.5-6 Let X take on the specified values v_1, \dots, v_k with probabilities $\theta_1, \dots, \theta_k$, respectively.

Suppose that X_1, \dots, X_n are independently and identically distributed as X . Suppose that $\theta = (\theta_1, \dots, \theta_k)$ is unknown and may range over the set $\Theta = \{(\theta_1, \dots, \theta_k) : \theta_i \geq 0, 1 \leq i \leq k, \sum_{i=1}^k \theta_i = 1\}$. Let N_j be the number of X_i which equal v_j .

(a). What is the distribution of (N_1, \dots, N_k) ?

$(N_1, \dots, N_k) \sim \text{Multinomial Distribution}$

$f_{\vec{\theta}}(\vec{n}) = n! \prod_{i=1}^k \frac{\theta_i^{n_i}}{n_i!} \mathbf{1}_{\{\sum N_i = n\}}$, where n_i = the number of times we get outcome $i = 1, \dots, k$ ■

(b). Show that $\mathbf{N} = (N_1, \dots, N_{k-1})$ is sufficient for θ .

$f_{\vec{\theta}}(\vec{N}) = n! \prod_{i=1}^k (N_i!)^{-1} \exp[\sum_{i=1}^k N_i \ln \theta_i] \mathbf{1}_{\{\sum N_i = n\}} = h(\vec{N}) \exp[\sum_{i=1}^k \eta_i(\vec{\theta}) T_i(\vec{N}) - B(\vec{\theta})]$, where $\chi = \{\vec{N} \in \{0, \dots, n\}^k \mid \sum N_i = n\}$

$h(\vec{N}) = n! \prod_{i=1}^k (N_i!)^{-1} \mathbf{1}_{\{\sum N_i = n\}}$, $B(\vec{\theta}) = 0$

$\eta_i(\vec{\theta}) = (\ln \theta_1, \dots, \ln \theta_k)$,

$T(\vec{N}) = (N_1, \dots, N_k)$ is a n.s.s of the family.

$T(\vec{N}) = (N_1, \dots, N_{k-1}, n - \sum_{i=1}^{k-1} N_i)$ is equivalent with (N_1, \dots, N_{k-1}) . Therefore \mathbf{N} is sufficient for θ . ■

1.5-7 Let X_1, \dots, X_n be a sample from a population with density $p(x, \theta)$ given by

$$p(x, \theta) = \begin{cases} \frac{1}{\sigma} \exp\{-\frac{x-\mu}{\sigma}\} & \text{if } x \geq \mu \\ 0 & \text{o.w.} \end{cases} \quad \text{Here } \theta = (\mu, \sigma) \text{ with } -\infty < \mu < \infty, \sigma > 0.$$

(a) Show that $\min(X_1, \dots, X_n)$ is sufficient for μ when σ is fixed.

When σ is fixed, $p(x_{1:n}, \mu) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x_i}{\sigma}] \exp[\frac{n\mu}{\sigma}] \prod_{i=1}^n \mathbf{1}_{\{x_i \geq \mu\}}$, where

$h(x) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x_i}{\sigma}]$, $g(T(x), \mu) = \exp[\frac{n\mu}{\sigma}] \prod_{i=1}^n \mathbf{1}_{\{x_i \geq \mu\}}$

$\mathbf{1}_{\{x_{(1)} \geq \mu\}}$ contains all the information about μ , then

$T(x) = \min(X_1, \dots, X_n)$ is sufficient for μ when σ is fixed. ■

- Another method is that $p(x_{1:n}|t)$ is free of μ

$X \sim \text{Expo}(\mu, 1/\sigma)$, $F_{\mu, \sigma}(x) = 1 - e^{-(x-\mu)/\sigma}$,

$\min(X_1, \dots, X_n) = X_{(1)} = n \frac{1}{\sigma} e^{-(x-\mu)/\sigma} [1 - (1 - e^{-(x-\mu)/\sigma})]^{n-1} = \frac{n}{\sigma} e^{-n(x-\mu)/\sigma}$

$p(x_{1:n}|t) = \frac{1}{n\sigma^{n-1}} e^{\frac{1}{\sigma}(\sum x_i - nx)}$ is free of μ

(b) Find a one-dimensional sufficient statistic for σ when μ is fixed.

When μ is fixed, $p(x_{1:n}, \sigma) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x}{\sigma} + \frac{n\mu}{\sigma}] \prod_{i=1}^n \mathbf{1}_{\{x_i \geq \mu\}}$, where

$h(x) = \prod_{i=1}^n \mathbf{1}_{\{x \geq \mu\}}$, $g(T(x), \sigma) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x}{\sigma} + \frac{n\mu}{\sigma}]$, then

$T(x) = \sum_{i=1}^n x_i$ is sufficient for σ when μ is fixed. ■

- Another method is that $p(x_{1:n}|t)$ is free of σ

$X \sim \text{Expo}(\mu, 1/\sigma)$, $F_{\mu, \sigma}(x) = 1 - e^{-(x-\mu)/\sigma}$,

$Y = X - \mu \sim \text{Exp}(1/\sigma)$, $T = \sum Y_i \sim \text{Gamma}(n, \sigma)$

$p(x_{1:n}|t) = \Gamma(n)t^{1-n}$ is free of σ

(c) Exhibit a two-dimensional sufficient statistic for θ .

$p(x_{1:n}, \mu, \sigma) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x}{\sigma} + \frac{n\mu}{\sigma}] \prod_{i=1}^n \mathbf{1}_{\{x_i \geq \mu\}}$, where

$h(x) = 1$, $g(T(x), \mu, \sigma) = \sigma^{-n} \exp[-\frac{\sum_{i=1}^n x}{\sigma} + \frac{n\mu}{\sigma}] \prod_{i=1}^n \mathbf{1}_{\{x_{(1)} \geq \mu\}}$, then

$T(x) = (x_{(1)}, \sum_{i=1}^n x_i)$ is a two-dimensional sufficient statistic for θ . ■

1.5-9 Let X_1, \dots, X_n be a sample from a population with density

$f_{\theta}(x) = \begin{cases} a(\theta)h(x) & \text{if } \theta_1 \leq x \leq \theta_2 \\ 0 & \text{o.w.} \end{cases}$ where $h(x) \geq 0$, $\theta = (\theta_1, \theta_2)$ with $-\infty < \theta_1 \leq \theta_2 < \infty$, and $a(\theta) =$

$[\int_{\theta_1}^{\theta_2} h(x)dx]^{-1}$ is assumed to exist. Find a two-dimensional sufficient statistic for this problem and apply your result to the $U[\theta_1, \theta_2]$ family of distributions.

Let $H'(x) = h(x)$, $a(\theta) = [\int_{\theta_1}^{\theta_2} h(x)dx]^{-1} = [H(\theta_2) - H(\theta_1)]^{-1}$

$f_{\theta_1, \theta_2}(x_{1:n}) = \prod_{i=1}^n a(\theta)h(x_i)\mathbf{1}_{\{x_i \in [\theta_1, \theta_2]\}} = \prod_{i=1}^n \mathbf{1}_{\{x_i \in [\theta_1, \theta_2]\}} [H(\theta_2) - H(\theta_1)]^{-n} \prod_{i=1}^n h(x_i)$, where

$g(T(x), \theta_1, \theta_2) = \prod_{i=1}^n \mathbf{1}_{\{x_i \in [\theta_1, \theta_2]\}} [H(\theta_2) - H(\theta_1)]^{-n}$, $h'(x) = \prod_{i=1}^n h(x_i)$

$\mathbf{1}_{\{x_{(n)} \leq \theta_2\}} \mathbf{1}_{\{x_{(1)} \geq \theta_1\}}$ contains all the information about θ , then

$T(x) = (x_{(1)}, x_{(n)})$ is a two-dimensional sufficient statistic for θ . ■

For $U[\theta_1, \theta_2]$, let $h(x) = 1$, $a(\theta) = (\theta_2 - \theta_1)^{-1}$

$f_{\theta_1, \theta_2}(x_{1:n}) = \prod_{i=1}^n a(\theta)h(x_i)\mathbf{1}_{\{x_i \in [\theta_1, \theta_2]\}} = \prod_{i=1}^n [\mathbf{1}_{\{x_i \in [\theta_1, \theta_2]\}}] [\theta_2 - \theta_1]^{-n} \prod_{i=1}^n 1$, where

$g(T(x), \theta_1, \theta_2) = \prod_{i=1}^n [\mathbf{1}_{\{x_{(n)} \leq \theta_2\}} \mathbf{1}_{\{x_{(1)} \geq \theta_1\}}] [\theta_2 - \theta_1]^{-n}$, $h'(x) = 1$

$T(x) = (x_{(1)}, x_{(n)})$ is a two-dimensional sufficient statistic for θ in the $U[\theta_1, \theta_2]$ family. ■