

SOLUTION FOR HOMEWORK 3, STAT 6331

Welcome to your third homework. Reminder: if you find a mistake/misprint, do not e-mail or call me. Write it down on the first page of your solutions and you may give yourself a partial credit — but keep in mind that the total for your homeworks cannot exceed 20 points.

Now let us look at your problems.

1. Let us show that $P((X, Y) = (i, j)|X - Y = 0, p)$ depends on p . This will yield that $X - Y$ is not SS. Write (remember that we may write $A \cap B := (A, B)$ to shorten formulas)

$$\begin{aligned} P((X, Y) = (i, j)|X - Y = 0, p) &= \frac{P((X, Y) = (i, j), X - Y = 0|p)}{P(X - Y = 0|p)} \\ &= \frac{P((X, Y) = (i, j), X = Y|p)}{P(X - Y = 0|p)} = \frac{P(X = i, Y = i|p)I(i = j)}{P(X - Y = 0|p)}. \end{aligned}$$

We have a specific distribution for the random variables, so let us use it to calculate the probabilities. We have $P(X = i, Y = i|p) = p^2(1 - p)^{2i-2}$ because X and Y are iid $\text{Geometric}(p)$. Our next step is to calculate $P(X - Y = 0|p)$. Write

$$\begin{aligned} P(X - Y = 0|p) &= P(X = Y|p) = \sum_{(i,j): i=j, i \geq 1} P((X, Y) = (i, j)|p) \\ &= \sum_{i=1}^{\infty} P(X = i|p)P(Y = i|p) = \sum_{i=1}^{\infty} p(1 - p)^{i-1}p(1 - p)^{i-1} = p^2 \sum_{i=1}^{\infty} [(1 - p)^2]^{i-1}. \end{aligned}$$

Now remember that for a partial geometric sum we have a nice formula $\sum_{k=0}^{n-1} q^k = (1 - q^n)/(1 - q)$ for $q \in (0, 1)$ (to prove it, multiply both sides by $(1 - q)$ and then cancel the same terms). We continue,

$$P(X - Y = 0|p) = p^2 \sum_{i=0}^{\infty} [(1 - p)^2]^i = \frac{p^2}{1 - (1 - p)^2} = \frac{p^2}{p(1 - p)} = \frac{p}{2 - p}.$$

Thus, the probability does depend on p and thus $X - Y$ is not SS.

2. First of all, let us stress that the likelihood function

$$T(\underline{X}) = \frac{f_{\theta_1}(\underline{X})}{f_{\theta_0}(\underline{X})}$$

is indeed a statistic because it is a function of observations. The fact that it also depends on θ_0 and θ_1 is irrelevant because these are two known values of an underlying parameter θ which we do not know. In other words, a statistic may depend on Ω and any other given information but not on the underlying parameter θ which we infer about.

Now I present several possible solutions. You can think about more...

A. Set $h(\underline{X}) := [f_{\theta_0}(\underline{X})f_{\theta_1}(\underline{X})]^{1/2}$ and $g(T(\underline{X}), \theta) := [T(\underline{X})]^{1/2}$ if $\theta = \theta_1$ and $g(T(\underline{X}), \theta) := [T(\underline{X})]^{-1/2}$ if $\theta = \theta_0$. Then we can write

$$f_\theta(\underline{X}) = f_\theta(g(T(\underline{X}), \theta))h(\underline{X}).$$

By the Factorization Theorem T is SS.

B. Set $h(\underline{X}) := f_{\theta_0}(\underline{X})$, and also $g(T(\underline{X}), \theta) := 1$ if $\theta = \theta_0$ and $g(T(\underline{X}), \theta) := T(\underline{X})$ if $\theta = \theta_1$. Then

$$f_\theta(\underline{X}) = g(T(\underline{X}), \theta)h(\underline{X}).$$

By the Factorization Theorem T is SS.

3. It is given that $\underline{X} := \{X_1, \dots, X_n\}$ is a sample from $U(\theta - 1/2, \theta + 1/2)$. Let us write the joint pdf so, may be, the Factorization Theorem will point on a SS. Write

$$\begin{aligned} & f(\underline{x}|\theta) \prod_{l=1}^n I(\theta - 1/2 \leq x_l \leq \theta + 1/2) \\ &= I(\theta - 1/2 \leq \min_{l \in \{1, \dots, n\}} x_l) I(\max_{l \in \{1, \dots, n\}} x_l \leq \theta + 1/2) = I(x_{(1)} \geq \theta - 1/2) I(x_{(n)} \leq \theta + 1/2) \\ & \quad =: g((x_{(1)}, x_{(n)}), \theta). \end{aligned}$$

Thus, by the Factorization Theorem, the pair $T(\underline{X}) := (X_{(1)}, X_{(n)})$ of the smallest and largest observations is SS.

4. a) Let T be SS and $T := \psi(U)$. By the Factorization Theorem (and remember that it is a criteria with iff), there exist some functions g and h such that

$$f(\underline{X}|\theta) = g(T, \theta)h(\underline{X}) = g(\psi(U), \theta)h(\underline{X}) =: g_1(U, \theta)h(\underline{X}).$$

Here g_1 is a new function such that $g_1(z, \theta) := g(\psi(z), \theta)$. Again using the Factorization Theorem, we establish that U is SS.

Remark: Please note that I used both the sufficient and necessary parts of the Factorization Theorem. Do you remember which part is the necessary one and which is sufficient one?

b) Now, let U be not SS and $T = \psi(U)$. Assume that T is sufficient. Then part a) yields that U must be SS, and the contradictory proves that T is not sufficient.

5. Problem 6.1. Let $X \sim N(0, \sigma^2)$. Is $|X|$ SS? Below I present 3 possible solutions.

a) Yes, and it does not matter that X is normal, the important part is that X is symmetrically distributed about 0 because then

$$f(|X| |X|, \sigma) = \begin{cases} 1/2 & \text{if } X = |X| \\ 1/2 & \text{if } X = -|X|. \end{cases}$$

Indeed, if $|X| =: a$ is given then X can be either a or $-a$, and this occurs with the same probability if the distribution is symmetric about 0.

b) Write

$$f(x| |x|, \sigma) = \frac{f(x| \sigma)}{q(|x| | \sigma)} = \frac{f(x| \sigma)}{f(-|x| | \sigma) + f(|x| : | \sigma)}.$$

Using the assumed symmetry we conclude that

$$f(x| |x|, \sigma) = 1/2I(x \in \{-|x|, |x|\}).$$

c) Write using the specific of the normal distribution,

$$f(x|\sigma) = (2\pi\sigma^2)^{1/2}e^{-x^2/(2\sigma^2)} = (2\pi\sigma^2)^{1/2}e^{-|x|^2/(2\sigma^2)}.$$

As a result, by the Factorization Theorem, X is SS.

6. Problem 6.2. Write using the assumed independence,

$$\begin{aligned} f_{\underline{X}}(\underline{x}|\theta) &= \prod_{i=1}^n f_{X_i}(x_i|\theta) \\ &= \prod_{i=1}^n e^{i\theta-x_i} I(x_i \geq i\theta) = e^{\theta \sum_{i=1}^n i} e^{-\sum_{i=1}^n x_i} \prod_{i=1}^n I(x_i/i \geq \theta) \\ &= e^{\theta \sum_{i=1}^n i} e^{-\sum_{i=1}^n x_i} \prod_{i=1}^n I(\min_i(x_i/i) \geq \theta) = [e^{-\sum_{i=1}^n x_i}] [e^{\theta \sum_{i=1}^n x_i} I(\min_i(x_i/i) \geq \theta)]. \end{aligned}$$

By the Factorization Theorem the statistic $T := \min_i(x_i/i)$ is SS.

7. Problem 6.3. By the iid property (note that below the indicator function is an extremely important part of the density — you must not skip it!),

$$f(\underline{x}|\mu, \sigma) = \prod_{l=1}^n [\sigma^{-1} e^{-(x_l - \mu)/\sigma} I(x_l > \mu)] = \sigma^{-n} e^{-[\sum_{l=1}^n x_l - n\mu]/\sigma} I(x_{(1)} > \mu).$$

We conclude that by the FT:

- a) if σ is given then $T_1 := X_{(1)}$ is SS for μ .
- b) if μ is given then $T_2 := \bar{X}$ is SS for σ .
- c) (T_1, T_2) is the two-dimensional SS for (μ, σ) .

8. Problem 6.5. By the independence

$$\begin{aligned} f(\underline{x}|\theta) &= \prod_{l=1}^n [2l\theta]^{-1} I(-l(\theta - 1) < x_l < l(\theta + 1)) = (2\theta)^{-n} \prod_{l=1}^n l^{-1} I(-(\theta - 1) < x_l/l < \theta + 1) \\ &= (2\theta)^{-n} \prod_{l=1}^n l^{-1} I(-(\theta - 1) < \min_l x_l/l) I(\max_l x_l/l < \theta + 1) =: g(\min_l(x_l/l), \max_l x_l/l, \theta). \end{aligned}$$

By the FT, the two-dimensional statistic $T := (\min_l(x_l/l), \max_l x_l/l)$ is SS.

9. Problem 6.6. It is given that X_1, \dots, X_n is a sample from $\text{Gamma}(\alpha, \beta)$. Then

$$\begin{aligned} f(x|\alpha, \beta) &= \prod_{l=1}^n \frac{1}{\Gamma(\alpha)\beta^\alpha} x_l^{\alpha-1} e^{-x_l/\beta} I(x_l \geq 0) \\ &= [\Gamma(\alpha)\beta^\alpha]^{-n} I(\min_l x_l \geq 0) \prod_{l=1}^n x_l^{\alpha-1} e^{-\sum_{l=1}^n x_l/\beta} \\ &= \{[\Gamma(\alpha)\beta^\alpha]^{-n} I(\min_l x_l \geq 0)\} \left\{ \prod_{l=1}^n x_l^{\alpha-1} e^{-\sum_{l=1}^n x_l/\beta} \right\}. \end{aligned}$$

The first factor in the curly brackets is a function in observations and the second is a function $g(\prod_{l=1}^n x_l, \sum_{l=1}^n x_l, \alpha, \beta)$.

Now we can conclude using the FT that:

- a) if α is given/known then $T_1 := \sum_{l=1}^n x_l$ is SS for β .
- b) if β is given/known then $T_2 := \prod_{l=1}^n x_l$ is SS for α .
- c) $T := (T_1, T_2)$ is SS for the pair (α, β) of parameters.

10. Problem 6.7. Begin with drawing a diagram showing the support of the bivariate density in R^2 (this is just a square with $[\theta_1, \theta_3]$ being the horizontal side in x-axis, and $[\theta_2, \theta_4]$ being the vertical side in y-axis).

Then we can write for a sample from (X, Y) ,

$$\begin{aligned} f(\underline{(x, y)} | \theta_1, \theta_2, \theta_3, \theta_4) &= [(\theta_3 - \theta_1)(\theta_4 - \theta_2)]^n \prod_{l=1}^n I(\theta_1 \leq x_l \leq \theta_3) I(\theta_2 \leq y_l \leq \theta_4) \\ &= [(\theta_3 - \theta_1)(\theta_4 - \theta_2)]^n \prod_{l=1}^n I(\theta_1 \leq \min_l x_l) I(\max_l x_l \leq \theta_3) I(\theta_2 \leq \min_l y_l) I(\max_l y_l \leq \theta_4). \end{aligned}$$

By the FT, the statistic $T := (x_{(1)}, x_{(n)}, y_{(1)}, y_{(n)})$ is SS for $(\theta_1, \theta_2, \theta_3, \theta_4)$.