

Solutions to exercises - Week 39

Sufficient statistic:

- Exercises 6.2, 6.3, 6.4, 6.5, 6.6 and 6.7

Minimal sufficient statistics:

- Exercises 6.8 and 6.9a-c

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Exercise 6.2

X_1, \dots, X_n are independent with pdfs

$$f_{X_i}(x_i | \theta) = \begin{cases} e^{i\theta - x_i} & \text{for } x_i \geq i\theta \\ 0 & \text{otherwise} \end{cases}$$

Note that for all x_i we may write

$$f_{X_i}(x_i | \theta) = e^{i\theta - x_i} I_{[i\theta, \infty)}(x_i) = e^{i\theta - x_i} I_{[\theta, \infty)}(x_i / i)$$

Hence the joint pdf may be written

$$f(\mathbf{x} | \theta) = \prod_{i=1}^n e^{i\theta - x_i} I_{[\theta, \infty)}(x_i / i) = e^{\theta \sum_{i=1}^n i} \underbrace{I_{[\theta, \infty)}(\min_i(x_i / i))}_{g(\min_i(x_i / i) | \theta)} \underbrace{e^{-\sum_{i=1}^n x_i}}_{h(\mathbf{x})}$$

Thus $T(\mathbf{X}) = \min_i(X_i / i)$ is sufficient for θ

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Exercise 6.3

X_1, \dots, X_n are iid with pdf

$$f(x | \mu, \sigma) = \begin{cases} (1/\sigma) e^{-(x-\mu)/\sigma} & \mu < x < \infty \\ 0 & \text{otherwise} \end{cases}$$

Note that we may write

$$f(x | \mu, \sigma) = (1/\sigma) e^{-(x-\mu)/\sigma} I_{(\mu, \infty)}(x)$$

Hence the joint pdf may be written

$$f(\mathbf{x} | \mu, \sigma) = \prod_{i=1}^n (1/\sigma) e^{-(x_i - \mu)/\sigma} I_{(\mu, \infty)}(x_i) = \frac{e^{n\mu/\sigma}}{\sigma^n} e^{-\sum_{i=1}^n x_i/\sigma} I_{(\mu, \infty)}(\min_i x_i)$$

Thus $T(\mathbf{X}) = (\min_i X_i, \sum_{i=1}^n X_i)$ is sufficient for (μ, σ)

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Exercise 6.4

X_1, \dots, X_n are iid with pmf or pdf of the form

$$f(x | \boldsymbol{\theta}) = h(x)c(\boldsymbol{\theta}) \exp\left\{\sum_{i=1}^k w_i(\boldsymbol{\theta})t_i(x)\right\}$$

where $\boldsymbol{\theta} = (\theta_1, \theta_2, \dots, \theta_d)$, $d \leq k$

The joint pmf or pdf may be written

$$f(\mathbf{x} | \boldsymbol{\theta}) = \prod_{j=1}^n \left\{ h(x_j) c(\boldsymbol{\theta}) \exp\left\{\sum_{i=1}^k w_i(\boldsymbol{\theta}) t_i(x_j)\right\} \right\} \\ = c(\boldsymbol{\theta})^n \exp\left\{\sum_{i=1}^k w_i(\boldsymbol{\theta}) \sum_{j=1}^n t_i(x_j)\right\} \prod_{j=1}^n h(x_j) \\ \underbrace{\hspace{10em}}_{g(T(\mathbf{x}) | \boldsymbol{\theta})} \underbrace{\hspace{10em}}_{h(\mathbf{x})}$$

Thus $T(\mathbf{X}) = \left(\sum_{j=1}^n t_1(X_j), \dots, \sum_{j=1}^n t_k(X_j)\right)$ is sufficient for $\boldsymbol{\theta}$

Exercise 6.5

X_1, \dots, X_n are independent with pdfs

$$f_{X_i}(x_i | \theta) = \begin{cases} \frac{1}{2i\theta} & -i(\theta-1) < x_i < i(\theta+1) \\ 0 & \text{otherwise} \end{cases}$$

Note that for all x_i we may write

$$\begin{aligned} f_{X_i}(x_i | \theta) &= \frac{1}{2i\theta} I_{(-i(\theta-1), i(\theta+1))}(x_i) \\ &= \frac{1}{2i\theta} I_{(-\infty, i(\theta+1))}(x_i) I_{(-i(\theta-1), \infty)}(x_i) \\ &= \frac{1}{2i\theta} I_{(-\infty, \theta+1)}(x_i/i) I_{(-(\theta-1), \infty)}(x_i/i) \end{aligned}$$

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Hence the joint pdf may be written

$$\begin{aligned} f(\mathbf{x} | \theta) &= \prod_{i=1}^n \left\{ \frac{1}{2i\theta} I_{(-\infty, \theta+1)}(x_i/i) I_{(-(\theta-1), \infty)}(x_i/i) \right\} \\ &= \left(\frac{1}{2\theta} \right)^n \left(\prod_{i=1}^n \frac{1}{i} \right) \left(\prod_{i=1}^n I_{(-\infty, \theta+1)}(x_i/i) \right) \left(\prod_{i=1}^n I_{(-(\theta-1), \infty)}(x_i/i) \right) \\ &= \left(\frac{1}{2\theta} \right)^n \frac{1}{n!} I_{(-\infty, \theta+1)}(\max_i(x_i/i)) I_{(-(\theta-1), \infty)}(\min_i(x_i/i)) \end{aligned}$$

Thus

$$T(\mathbf{X}) = (\min_i(X_i/i), \max_i(X_i/i))$$

is sufficient for θ

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Exercise 6.6

X_1, \dots, X_n are iid with pdf

$$f(x | \alpha, \beta) = \begin{cases} \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta} & \text{for } x > 0 \\ 0 & \text{otherwise} \end{cases}$$

We may write: $f(x | \alpha, \beta) = \frac{1}{\beta^\alpha \Gamma(\alpha)} x^{\alpha-1} e^{-x/\beta} I_{(0, \infty)}(x)$

Hence the joint pdf may be written

$$\begin{aligned} f(\mathbf{x} | \alpha, \beta) &= \prod_{i=1}^n \frac{1}{\beta^\alpha \Gamma(\alpha)} x_i^{\alpha-1} e^{-x_i/\beta} I_{(0, \infty)}(x_i) \\ &= \underbrace{\left(\frac{1}{\beta^\alpha \Gamma(\alpha)} \right)^n \left(\prod_{i=1}^n x_i \right)^{\alpha-1} e^{-\sum x_i/\beta}}_{g(T(\mathbf{x}) | \theta)} \underbrace{I_{(0, \infty)}(\min_i x_i)}_{h(\mathbf{x})} \end{aligned}$$

Thus $T(\mathbf{X}) = \left(\prod_{i=1}^n X_i, \sum_{i=1}^n X_i \right)$ is sufficient for (α, β)

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Exercise 6.7

$(X_1, Y_1), \dots, (X_n, Y_n)$ are independent with a pdf that is given by

$$f_{XY}(x, y | \theta_1, \theta_2, \theta_3, \theta_4) = \frac{1}{(\theta_2 - \theta_1)(\theta_4 - \theta_3)}$$

when $\theta_1 < x < \theta_2$ and $\theta_3 < y < \theta_4$ and by

$$f_{XY}(x, y | \theta_1, \theta_2, \theta_3, \theta_4) = 0$$

otherwise

Thus we may write

$$f_{XY}(x, y | \theta_1, \theta_2, \theta_3, \theta_4) = \frac{1}{(\theta_2 - \theta_1)(\theta_4 - \theta_3)} I_{(\theta_1, \theta_2)}(x) I_{(\theta_3, \theta_4)}(y)$$

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Hence the joint pdf may be written

$$\begin{aligned} f(\mathbf{x}, \mathbf{y} | \alpha, \beta) &= \prod_{i=1}^n f_{XY}(x_i, y_i | \theta_1, \theta_2, \theta_3, \theta_4) \\ &= \prod_{i=1}^n \frac{1}{(\theta_2 - \theta_1)(\theta_4 - \theta_3)} I_{(\theta_1, \theta_2)}(x_i) I_{(\theta_3, \theta_4)}(y_i) \\ &= \left(\frac{1}{(\theta_2 - \theta_1)(\theta_4 - \theta_3)} \right)^n I_{(\theta_1, \infty)}(\min x_i) I_{(-\infty, \theta_2)}(\max x_i) I_{(\theta_3, \infty)}(\min y_i) I_{(-\infty, \theta_4)}(\max y_i) \end{aligned}$$

Thus $T(\mathbf{X}, \mathbf{Y}) = (\min X_i, \max X_i, \min Y_i, \max Y_i)$ is sufficient for $(\theta_1, \theta_2, \theta_3, \theta_4)$

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Exercise 6.8

X_1, \dots, X_n are iid with pdf $f(x - \theta)$

The joint pdf takes the form $f(\mathbf{x} | \theta) = \prod_{i=1}^n f(x_i - \theta)$

Hence

$$\frac{f(\mathbf{x} | \theta)}{f(\mathbf{y} | \theta)} = \frac{\prod_{i=1}^n f(x_i - \theta)}{\prod_{i=1}^n f(y_i - \theta)} = \frac{\prod_{i=1}^n f(x_{(i)} - \theta)}{\prod_{i=1}^n f(y_{(i)} - \theta)}$$

For a general f this does not depend on θ if and only if \mathbf{y} and \mathbf{x} have the same order statistics

Thus the order statistics are minimal sufficient (cf. Theorem 6.2.13)

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Exercise 6.9.a

X_1, \dots, X_n are iid with pdf $f(x | \theta) = \frac{1}{\sqrt{2\pi}} e^{-(x-\theta)^2/2}$

The joint pdf takes the form

$$\begin{aligned} f(\mathbf{x} | \theta) &= \prod_{i=1}^n \frac{1}{\sqrt{2\pi}} e^{-(x_i - \theta)^2/2} \\ &= (2\pi)^{-n/2} \exp \left\{ -\frac{1}{2} \left[\sum_{i=1}^n x_i^2 - 2n\theta \bar{x} + n\theta^2 \right] \right\} \end{aligned}$$

Hence

$$\frac{f(\mathbf{x} | \theta)}{f(\mathbf{y} | \theta)} = \exp \left\{ -\frac{1}{2} \left[\left(\sum_{i=1}^n x_i^2 - \sum_{i=1}^n y_i^2 \right) - 2n\theta(\bar{x} - \bar{y}) \right] \right\}$$

This does not depend on θ if and only if $\bar{y} = \bar{x}$

Thus \bar{X} is a minimal sufficient statistic for θ (cf. Theorem 6.2.13)

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Exercise 6.9.b

X_1, \dots, X_n are iid with pdf $f(x | \theta) = \begin{cases} e^{-(x-\theta)} & \theta < x < \infty \\ 0 & \text{otherwise} \end{cases}$

The joint pdf takes the form

$$\begin{aligned} f(\mathbf{x} | \theta) &= \prod_{i=1}^n e^{-(x_i - \theta)} I_{(\theta, \infty)}(x_i) = e^{n\theta} e^{-\sum x_i} \prod_{i=1}^n I_{(\theta, \infty)}(x_i) \\ &= e^{n\theta} e^{-\sum x_i} I_{(\theta, \infty)}(\min_i x_i) \end{aligned}$$

Hence

$$\frac{f(\mathbf{x} | \theta)}{f(\mathbf{y} | \theta)} = \frac{e^{-\sum x_i} I_{(\theta, \infty)}(\min_i x_i)}{e^{-\sum y_i} I_{(\theta, \infty)}(\min_i y_i)}$$

This does not depend on θ if and only if $\min y_i = \min x_i$

Thus $\min_i X_i$ is a minimal sufficient statistic for θ

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Exercise 6.9.c

X_1, \dots, X_n are iid with pdf $f(x|\theta) = \frac{e^{-(x-\theta)}}{(1+e^{-(x-\theta)})^2}$

The joint pdf takes the form

$$f(\mathbf{x}|\theta) = \prod_{i=1}^n \frac{e^{-(x_i-\theta)}}{(1+e^{-(x_i-\theta)})^2} = \frac{e^{n\theta} e^{-\sum x_i}}{\prod_{i=1}^n (1+e^{-(x_i-\theta)})^2}$$

Hence

$$\frac{f(\mathbf{x}|\theta)}{f(\mathbf{y}|\theta)} = \frac{e^{\sum y_i - \sum x_i} \prod_{i=1}^n (1+e^{-(y_i-\theta)})^2}{\prod_{i=1}^n (1+e^{-(x_i-\theta)})^2} = \frac{e^{\sum y_{(i)} - \sum x_{(i)}} \prod_{i=1}^n (1+e^{-(y_{(i)}-\theta)})^2}{\prod_{i=1}^n (1+e^{-(x_{(i)}-\theta)})^2}$$

This does not depend on θ if and only if \mathbf{y} and \mathbf{x} and have the same order statistics

Thus the order statistics are minimal sufficient