

MATH 557 - ASSIGNMENT 1 SOLUTIONS

1 We construct the joint pmf/pdf in each case, and inspect the required conditional pdf. Note that 1-1 transformations of the statistics are also sufficient.

(a) For the

$$f_{\underline{X}|\alpha,\beta}(\underline{x}|\alpha,\beta) = \left\{ \frac{\Gamma(\alpha+\beta)}{\Gamma(\alpha)\Gamma(\beta)} \right\}^n \left\{ \prod_{i=1}^n x_i \right\}^{\alpha-1} \left\{ \prod_{i=1}^n (1-x_i) \right\}^{\beta-1}$$

suggesting the sufficient statistic

$$\underline{T}(\underline{X}) = \left(\prod_{i=1}^n x_i, \prod_{i=1}^n (1-x_i) \right)^T$$

and the result follows using the Neyman Factorization Theorem.

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(b) Writing $\lambda = \log \theta$, we realize that this is the *Poisson*($\log \theta$) model. Hence by properties of the Exponential Family

$$T(\underline{X}) = \sum_{i=1}^n X_i$$

is a sufficient statistic for $\log \theta$.

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(c) We have

$$\begin{aligned} f_{\underline{X}|\theta}(\underline{x}|\theta) &= \frac{1}{\theta} && \theta < x_1, \dots, x_n < 2\theta \\ &= \frac{1}{\theta} I_{(x_{(n)}/2, x_{(1)})}(\theta) \end{aligned}$$

suggesting the sufficient statistic

$$\underline{T}(\underline{X}) = (X_{(1)}, X_{(n)})^T$$

and the result follows using the Neyman Factorization Theorem.

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2 (a) We have

$$f_{\underline{X}|\lambda}(\underline{x}|\lambda) = \lambda^n \exp \left\{ -\lambda \sum_{i=1}^n x_i \right\} = \lambda^n \exp \{ -\lambda T(\underline{x}) \}$$

say, so that for two points \underline{x} and \underline{y} the ratio

$$\frac{f_{\underline{X}|\lambda}(\underline{x}|\lambda)}{f_{\underline{X}|\lambda}(\underline{y}|\lambda)} = \exp \left\{ -\lambda (T(\underline{x}) - T(\underline{y})) \right\}$$

which is a constant if and only if $T(\underline{x}) = T(\underline{y})$. Therefore $T(\underline{x})$ is minimal sufficient.

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- (b) Given $x_{(1)}, \dots, x_{(m)}$, we can construct the joint pdf for the order statistic data by noting that if $X_{(m)} = x_{(m)}$, then we have $X_{(r)} > x_{(m)}$ for the $n - m$ order statistics $X_{(r)}$, $r = m + 1, \dots, n$. Thus, as the “survivor” function takes the form $1 - F_{X|\lambda}(x|\lambda) = e^{-\lambda x}$, we have

$$f_{\underline{X}|\lambda}(\underline{x}|\lambda) = m! \binom{n}{m} \times \lambda^m \exp \left\{ -\lambda \sum_{i=1}^m x_{(i)} \right\} \times \exp \{ -(n - m)\lambda x_{(m)} \}$$

where the combinatorial term counts the number of possible arrangements of the random sample points. Thus a sufficient statistic is

$$T(\underline{X}) = \sum_{i=1}^m X_{(i)} + (n - m)X_{(m)}$$

by the Neyman Factorization Theorem.

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3 We have for $t = 0, 1, \dots$,

$$f_{\underline{X}|\theta}(\underline{x}|\theta) = \frac{\theta^{\sum_{i=1}^n x_i} e^{-n\theta}}{\prod_{i=1}^n x_i!}$$

and $T(\underline{X}) \sim \text{Poisson}(n\theta)$ from distributional results, so that

$$f_{\underline{X}|T(\underline{X})}(\underline{x}|t) = \frac{\theta^{\sum_{i=1}^n x_i} e^{-n\theta} / \prod_{i=1}^n x_i!}{(n\theta)^t e^{-n\theta} / t!} = \frac{t!}{x_1! \dots x_n!} \left(\frac{1}{n} \right)^t \quad \underline{x} \in A_t$$

and zero otherwise, where

$$A_t \equiv \{ \underline{x} : x_1 + \dots + x_n = t \}.$$

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