## Mathematical Statistics Homework 3

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## A Sufficient Statistic for a Parameter 7.2

Exercise 2.1  $X \sim N(0, \theta)$ 

$$f(x;\theta) = \frac{1}{\sqrt{2\pi\theta}} e^{\frac{-x^2}{2\theta}}$$
,  $-\infty < x < \infty$ ,  $0 < \theta < \infty$ 

$$L(\theta) = \prod_{i=1}^{n} f(x_i; \theta)$$

$$= \prod_{i=1}^{n} \frac{1}{\sqrt{2\pi\theta}} e^{\frac{-x_i^2}{2\theta}}$$

$$= (2\pi\theta)^{-\frac{n}{2}} e^{-\frac{1}{2\theta} \sum_{i=1}^{n} x_i^2}$$

$$= k_1 (\sum_{i=1}^{n} x_i^2; \theta) k_2(x_1, x_2 \dots x_n; \theta)$$

$$= k_1 (\sum_{i=1}^{n} x_i^2; \theta) \cdot 1$$

thus, according to the factorization theorem of Neyman,  $\sum_{i=1}^{n} x_i^2$  is a sufficient statistics for  $\theta \sharp$ Exercise 2.2  $X \sim Poisson(\theta)$ 

$$f(x;\theta) = \frac{\theta^x e^{-\theta}}{x!}, \qquad 0 < \theta < \infty, \quad x > 0$$

$$L(\theta) = \prod_{i=1}^{n} \frac{\theta^{x_i} e^{-\theta}}{x_i!}$$

$$= \left(\frac{1}{\prod_{i=1}^{n} x_i!}\right) \left(\theta^{\sum_{i=1}^{n} x_i} e^{-n\theta}\right)$$

$$= k_2(x_1, x_2 \dots x_n; \theta) k_1(\sum_{i=1}^{n} x_i; \theta)$$

thus, according to the factorization theorem of Neyman,  $\sum_{i=1}^{n} x_i$  is a sufficient statistics for  $\theta \sharp$ Exercise 2.7  $X \sim \Gamma(\theta, 6)$ 

$$f(x;\theta) = \frac{1}{\Gamma(\theta)6^{\theta}} x^{\theta - 1} e^{\frac{-x}{6}} \qquad 0 < x < \infty, \quad \theta > 0$$

$$L(\theta) = \prod_{i=1}^{n} \frac{1}{\Gamma(\theta)6^{\theta}} x_i^{\theta-1} e^{\frac{-x_i}{6}}$$

$$= \left(\frac{1}{\Gamma(\theta)6^{\theta}} (\prod_{i=1}^{n} x_i)^{\theta-1}\right) e^{\frac{-1}{6} \sum_{i=1}^{n} x_i}$$

$$= k_1 (\prod_{i=1}^{n} x_i; \theta) k_2(x_1, x_2 \dots x_n; \theta)$$

thus, according to the factorization theorem of Neyman,  $\prod_{i=1}^n x_i$  is a sufficient statistics for  $\theta \sharp$ 

## Properties of a Sufficient Statistic 7.3

Exercise 3.3

Given a random sample  $X \sim \Gamma(1, \theta)$  of sample size 2

$$f(x; \theta) = \frac{1}{\theta} e^{-x/\theta}$$
  $0 < x < \infty, \quad 0 < \theta < \infty$ 

The joint PDF is given by

$$f(x_1, x_2; \theta) = \frac{1}{\theta^2} e^{\frac{-1}{\theta}(x_1 + x_2)}$$
  $0 < x_1, x_2 < \infty, \quad 0 < \theta < \infty$ 

$$\begin{cases} Y_1 = X_1 + X_2 \\ Y_2 = X_2 \end{cases} \rightarrow \begin{cases} X_1 = Y_1 - Y_2 \\ X_2 = Y_2, \end{cases} 0 < Y_2 < Y_1 < \infty$$

$$|J| = |\begin{vmatrix} 1 & 0 \\ -1 & 1 \end{vmatrix}| = 1$$

$$f(y_1, y_2; \theta) = \frac{1}{\theta^2} e^{\frac{-1}{\theta}(y_1)}$$
  $0 < y_2 < y_1 < \infty, \quad 0 < \theta < \infty$ 

marginal of  $f_{Y_1}$  is

$$f(y_1) = \int_0^{y_1} \frac{1}{\theta^2} e^{\frac{-1}{\theta}(y_1)} dy_2$$
  
=  $\frac{1}{\theta^2} e^{\frac{-1}{\theta}(y_1)} y_1, \quad 0 < y_1 < \infty, \quad 0 < \theta < \infty$ 

the conditional pdf of  $Y_2$  on  $Y_1$  is

$$f(Y_2|y_1;\theta) = \frac{f(y_1, y_2; \theta)}{f(y_1; \theta)}$$

$$= \frac{\frac{1}{\theta^2} e^{\frac{-1}{\theta}(y_1)}}{\frac{1}{\theta^2} e^{\frac{-1}{\theta}(y_1)} y_1}$$

$$= \frac{1}{y_1} , 0 < y_2 < y_1 < \infty$$

Since  $Y_2 = X_2$ , and  $X_2 \sim \Gamma(1, \theta)$ , thus  $E(Y_2) = \theta \cdot 1 = \theta \sharp$  $Var(Y_2) = 1 \cdot \theta^2 = \theta^2 \sharp$ 

$$E(Y_2|y_1) = \int_0^{y_1} y_2 \frac{1}{y_1} dy_2$$
  
=  $\frac{1}{2} y_1 = \varphi(y_1) \sharp$ 

$$Var(\varphi(y_1)) = Var(\frac{1}{2}Y_1),$$
 since  $Y_1 \sim \Gamma(2, \theta)$   
=  $\frac{1}{4}2\theta^2$   
=  $\frac{\theta^2}{2} \sharp$