

#1.

4.2.2. Let  $X_1, \dots, X_n$  be a sample of iid  $\text{Gamma}(\theta, 1)$  random variables with  $\theta \in (0, \infty)$ .

a) Determine the likelihood function  $L(\theta)$ .

$$L(\theta) = \frac{1}{\Gamma(\theta)^n} (x_1 x_2 \cdots x_n)^{\theta-1} e^{-(x_1 + \cdots + x_n)}, \quad x_i > 0.$$

b) Use the Fisher-Neyman factorization theorem to determine a sufficient statistic  $S$  for  $\theta$ .

In Fisher-Neyman notation,  $g(S, \theta) = \frac{1}{\Gamma(\theta)^n} (x_1 x_2 \cdots x_n)^{\theta-1}$  and  $h(x_1, \dots, x_n) = e^{-(x_1 + \cdots + x_n)}$ .

A sufficient statistic for  $\theta$  is  $S = X_1 X_2 \cdots X_n$ .

#2. 4.2.4 Let  $X_1, \dots, X_n$  be a sample of iid  $\text{Beta}(4, \theta)$  random variables with  $\theta \in (0, \infty)$ .

a) Determine the likelihood function  $L(\theta)$ .

$$L(\theta) = \left( \frac{\Gamma(4+\theta)}{\Gamma(4)\Gamma(\theta)} \right)^n (x_1 x_2 \cdots x_n)^3 \cdot ((1-x_1)(1-x_2) \cdots (1-x_n))^{\theta-1}$$

b) In Fisher-Neyman notation,  $g(S, \theta) = \left( \frac{\Gamma(4+\theta)}{\Gamma(4)\Gamma(\theta)} \right)^n ((1-x_1)(1-x_2) \cdots (1-x_n))^{\theta-1}$  and  $h(x_1, \dots, x_n) = (x_1 x_2 \cdots x_n)^3$ . A sufficient statistic is  $S = (1-X_1)(1-X_2) \cdots (1-X_n)$ .

#3. 4.2.8 Let  $X_1, \dots, X_n$  be a sample of iid  $\text{Beta}(\theta_1, \theta_2)$  random variables with  $\theta \in \mathbf{R}^+ \times \mathbf{R}^+$ . Use the Fisher-Neyman factorization theorem to determine a sufficient statistic  $S$  for  $\vec{\theta}$ .

$$L(\theta) = \left( \frac{\Gamma(\theta_1+\theta_2)}{\Gamma(\theta_1)\Gamma(\theta_2)} \right)^n (x_1 x_2 \cdots x_n)^{\theta_1-1} \cdot ((1-x_1)(1-x_2) \cdots (1-x_n))^{\theta_2-1}.$$

Every term in  $L(\theta)$  contains  $\theta_1$  or  $\theta_2$ , so we need a pair of statistics for sufficiency, namely  $(S_1, S_2) = (x_1 x_2 \cdots x_n, (1-x_1)(1-x_2) \cdots (1-x_n))$ .

#4. 4.2.10. Let  $X_1, \dots, X_n$  be a sample of iid random variables with pdf  $f(x) = \theta x^{\theta-1}$  for  $x \in (0, 1)$  and  $\theta \in (0, \infty)$ . Use the Fisher-Neyman factorization theorem to determine a sufficient statistic for  $\theta$ .

$L(\theta) = \theta^n (x_1 x_2 \cdots x_n)^{\theta-1}$ . Here  $g(S, \theta) = L(\theta)$  and  $h(x_1, \dots, x_n) = 1$ . A sufficient statistic is  $S = X_1 X_2 \cdots X_n$ .

#5. Consider the family of distributions with pmf  $p_X(x) = \begin{cases} p & \text{if } x = -1 \\ 2p & \text{if } x = 0 \\ 1 - 3p & \text{if } x = 1 \end{cases}$ .

Here  $p$  is an unknown parameter, and  $0 \leq p \leq 1/3$ .

Let  $X_1, X_2, \dots, X_n$  be iid with common pmf a member of this family. Consider the statistics  $A$  = the number of  $i$  with  $X_i = -1$ ,  $B$  = the number of  $i$  with  $X_i = 0$ ,  $C$  = the number of  $i$  with  $X_i = 1$ .

(i) Write down the joint pmf of  $X_1, X_2, \dots, X_n$ . This is most easily done using the statistics  $A, B, C$ .  $p_{X_1, \dots, X_n}(x_1, \dots, x_n) = p^A (2p)^B (1 - 3p)^C$ .

(ii) Use the Fischer-Neyman Lemma to show that  $C$  is a sufficient statistic for  $p$ .

Since  $A + B + C = n$ , we can rewrite the joint pdf in the form

$$p_{X_1, \dots, X_n}(x_1, \dots, x_n) = p^{A+B} 2^B (1 - 3p)^C = p^{n-C} 2^B (1 - 3p)^C = p^{n-C} (1 - 3p)^C \cdot 2^B.$$

In Fisher-Neyman notation,  $g(C; p) = p^{n-C} (1 - 3p)^C$  and  $h(x_1, \dots, x_n) = 2^B$ ;  $C$  is a sufficient statistic for  $p$ .

(iii) Does it appear that  $A$  is also sufficient for  $p$ ? Explain why or why not.

$A$  is not sufficient for  $p$ . The factorization trick of (iii) that works for  $C$  does not work for  $A$ .

#6. Let  $X_1, X_2, X_3$  be iid, binomial(1,  $p$ ). Let  $T = X_1 + X_2 + 2X_3$ . The purpose of this problem is to determine whether  $T$  is a sufficient statistic for  $p$ . Recall that the definition says that  $T$  is sufficient for  $p$  if for all  $p \in [0, 1]$

$$p_{X_1, X_2, X_3 | T}(x_1, x_2, x_3 \mid T = x_1 + x_2 + 2x_3) \text{ does not depend on } p.$$

Let's examine one particular instance of this definition.

Compute  $p_{X_1, X_2, X_3 | T}(0, 0, 1 \mid T = 2)$ . Does it depend on  $p$ ? Is  $T$  a sufficient statistic for  $p$ ?

Solution: There are only two triples for which  $T = 2$ , namely  $(X_1, X_2, X_3) = (1, 1, 0)$  and  $(0, 0, 1)$ . Therefore  $P(T = 2) = p_{X_1, X_2, X_3}(1, 1, 0) + p_{X_1, X_2, X_3}(0, 0, 1) = p^2(1 - p) + (1 - p)^2 p = p(1 - p) \cdot (p + 1 - p) = p(1 - p)$ . Therefore

$$p_{X_1, X_2, X_3 | T}(0, 0, 1 \mid T = 2) = \frac{p_{X_1, X_2, X_3 | T}(0, 0, 1)}{P(T = 2)} = \frac{(1 - p)^2 p}{p(1 - p)} = 1 - p,$$

which certainly depends on  $p$ . We conclude that  $T$  is not a sufficient statistic for  $p$ .