

1. Suppose  $X_1, X_2, \dots, X_n$  are iid  $\mathcal{N}(\mu, \sigma^2)$ , where  $-\infty < \mu < \infty$  and  $\sigma^2 > 0$  and both parameters are unknown. Set  $\mathbf{X} = (X_1, X_2, \dots, X_n)$  and  $\boldsymbol{\theta} = (\mu, \sigma^2)$ .

(a) Show that

$$\mathbf{T} = \mathbf{T}(\mathbf{X}) = \begin{pmatrix} \bar{X} \\ S^2 \end{pmatrix}$$

is a minimal sufficient statistic for  $\boldsymbol{\theta}$ .

(b) Define

$$W = \left( \frac{\bar{X} - \mu}{\sigma/\sqrt{n}} \right)^2 + \frac{(n-1)S^2}{\sigma^2}.$$

Find the finite-sample distribution of  $W$ .

(c) Is  $W$  an ancillary statistic? Explain.

2. Recall that if  $X$  has a beta( $\alpha, \beta$ ) distribution, then the probability density function (pdf) of  $X$  is

$$f_X(x|\alpha, \beta) = \frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1}(1-x)^{\beta-1} I(0 < x < 1),$$

where  $\alpha > 0$  and  $\beta > 0$ . In this problem, we are going to consider the beta subfamily where  $\alpha = \beta = \theta$ . Let  $X_1, X_2, \dots, X_n$  denote an iid sample from a beta( $\theta, \theta$ ) distribution.

(a) Show that

$$T(\mathbf{X}) = \prod_{i=1}^n X_i(1 - X_i)$$

is a sufficient statistic for  $\theta$ . Is  $T(\mathbf{X})$  complete?

(b) The two-dimensional statistic

$$\mathbf{T}^* = \mathbf{T}^*(\mathbf{X}) = \begin{pmatrix} \prod_{i=1}^n X_i \\ \prod_{i=1}^n (1 - X_i) \end{pmatrix}$$

is also a sufficient statistic for  $\theta$ . What must be true about the conditional distribution  $f_{\mathbf{X}|\mathbf{T}^*}(\mathbf{x}|\mathbf{t})$ ?

(c) Show that  $\mathbf{T}^*(\mathbf{X})$  is not a complete statistic.

3. Suppose  $W_n \sim \chi_n^2$ .

(a) Prove  $W_n \stackrel{d}{=} X_1 + X_2 + \dots + X_n$ , where  $X_1, X_2, \dots, X_n$  are iid  $\chi_1^2$ . Recall that “ $\stackrel{d}{=}$ ” means “equal in distribution.”

(b) Explain why  $W_n/n \xrightarrow{p} 1$ , as  $n \rightarrow \infty$ .

(c) Find sequences of constants  $a_n$ ,  $b_n$ , and  $c_n$  such that

$$a_n(W_n/n - b_n) + c_n \xrightarrow{d} \mathcal{N}(-3, 5),$$

as  $n \rightarrow \infty$ . Prove your answer. **Note:** Sequences of constants can be constant sequences of constants.

(d) Suppose  $V_n \sim F_{n,2n}$ . Does  $V_n$  converge in probability to  $1/2$ ,  $1$ , or  $2$ ? Or, does  $V_n$  not converge at all? Prove your claim.

4. Suppose  $X_1, X_2, \dots, X_n$  is an iid sample from

$$f_X(x|\theta) = \frac{2x}{\theta} e^{-x^2/\theta} I(x > 0),$$

where  $\theta > 0$  is unknown.

(a) Argue that  $e^{-\bar{X}_n}$  converges in probability to a constant and find the constant.

(b) Define the sequence

$$\hat{\theta}_n = \frac{1}{n} \sum_{i=1}^n X_i^2.$$

Prove that  $\sqrt{n}(\hat{\theta}_n - \theta) \xrightarrow{d} \mathcal{N}(0, \theta^2)$ , as  $n \rightarrow \infty$ .

(c) Find a function of the  $\hat{\theta}_n$  sequence, say  $g(\hat{\theta}_n)$ , whose large-sample variance is free of  $\theta$ . *Hint:* Use the (first-order) Delta Method.

5. Suppose  $X_1, X_2, \dots, X_n$  are iid exponential with mean  $\theta > 0$ .

(a) Prove that

$$T = \sum_{i=1}^n X_i \quad \text{and} \quad S = \frac{nX_{(1)}}{\sum_{i=1}^n X_i}$$

are independent statistics.

(b) What is the (finite-sample) sampling distribution of  $T$ ?

(c) Calculate  $E(S)$ .

6. Suppose that  $Y$  is a discrete random variable with probability mass function (pmf)

$$\begin{aligned} P(Y = 0) &= p + (1 - p)e^{-\lambda} \\ P(Y = y) &= (1 - p) \frac{\lambda^y e^{-\lambda}}{y!}, \quad y = 1, 2, 3, \dots \end{aligned}$$

(a) Reparameterize the model by defining  $\pi = P(Y = 0) = p + (1 - p)e^{-\lambda}$ . Solve this last equation for  $p$  in terms of  $\pi$  and  $\lambda$  and then substitute so that the pmf of  $Y$  depends only on  $\pi$  and  $\lambda$ . Show your work.

(b) For an iid sample  $Y_1, Y_2, \dots, Y_n$ , let  $n_0$  denote the number of zeroes in the sample. Show that the likelihood  $L(\pi, \lambda | \mathbf{y})$  factors into two pieces and that the maximum likelihood estimator of  $\pi$  is  $\hat{\pi} = n_0/n$ . Also, show that the maximum likelihood estimator for  $\lambda$  is the solution to a simple nonlinear equation involving  $\bar{Y}_+$ , the average of the nonzero  $Y$  values.

(c) Under the reparameterized model, set up equations that, when solved, will produce the method of moments estimators of  $\pi$  and  $\lambda$ . You do not have to solve the equations.