

### Solutions of Exercises 5.5.1 to 5.5.3.

5.5.1. (a) If  $X \in \mathcal{C}(0, 1)$ , and  $U = 2X/(1+X^2)$ , then  $EU = 0$  since  $|U| \leq 1$  and  $U$  has a symmetric distribution about 0. To evaluate  $EU^2$ , we make the change of variable,  $\theta = \arctan(x)$  with  $dx = (1/\cos^2\theta)d\theta$ .

$$\begin{aligned} EU^2 &= \int_{-\infty}^{\infty} \left( \frac{2x}{1+x^2} \right)^2 \frac{1}{\pi(1+x^2)} dx = \frac{8}{\pi} \int_0^{\infty} \frac{x^2}{(1+x^2)^3} dx \\ &= \frac{8}{\pi} \int_0^{\pi/2} \sin^2\theta \cos^4\theta \frac{1}{\cos^2\theta} d\theta = \frac{8}{\pi} \int_0^{\pi/2} \sin^2\theta \cos^2\theta d\theta \end{aligned}$$

Standard recursive methods for integrating  $\cos^m\theta \sin^n\theta$  give

$$\int \sin^2\theta \cos^2\theta d\theta = \frac{1}{4}[\sin^3\theta \cos\theta] + \frac{1}{8}[\theta - \sin\theta \cos\theta]$$

(as may be checked by differentiating). Hence,  $\text{Var}(U) = EU^2 = (8/\pi)(\pi/16) = 1/2$ .

(b) We reject  $H_0$  if  $\sum_1^n U_i > k$ , where  $n = 100$ . The central limit theorem gives  $\sum_1^n U_i$  as approximately normal with mean zero and variance  $n/2$ . Hence,  $P(\sum_1^n U_i > k) = P(\sqrt{2/n} \sum_1^n U_i > \sqrt{2/n}k) = .05$ , gives  $\sqrt{2/n}k$  as approximately 1.645. Hence,  $k = 1.645\sqrt{50} = 11.63$ .

5.5.2. Part (a) is false unless  $\alpha$  is restricted to be less than or equal to  $1 - (1/2)^n$ . *Proof for  $n = 2$ :* Look at the contours of the likelihood ratio,

$$\varphi(x_1, x_2) = \log(f(x_1, x_2|\theta)/f(x_1, x_2|0)) = |x_1| - |x_1 - \theta| + |x_2| - |x_2 - \theta|.$$

We are to show that the sets  $\{\log(f(x_1, x_2|\theta)/f(x_1, x_2|0)) > k\}$  depend on  $\theta$ . If  $\alpha \geq 3/4$ , there is a UMP test of size  $\alpha$  whose acceptance region is any subset of the set  $\{(x_1, x_2)|x_1 < 0, x_2 < 0\}$  of probability  $1 - \alpha$  under  $H_0$ . However, if  $\alpha < .75$ , the region required to achieve size  $\alpha$  depends on  $\theta$ .

(b) The locally best test of (5.78) becomes: Reject  $H_0$  if

$$\frac{\partial}{\partial\theta} \log f(\mathbf{x}|\theta)|_{\theta=0} = \frac{\partial}{\partial\theta} - \sum_{i=1}^n |x_i - \theta|_{\theta=0} = + \sum_{i=1}^n \text{sgn}(x_i)$$

is too large. This is equivalent to rejecting  $H_0$  if  $T$  is too large, where  $T$  is the number of positive  $X_i$ . This is the sign test. Under  $H_0: \theta = 0$ ,  $T$  has a binomial distribution,  $\mathcal{B}(n, 1/2)$ . For general  $\theta$ , the distribution of  $T$  is  $\mathcal{B}(n, p(\theta))$  where  $p(\theta) = P_\theta(X > 0) = 1 - e^{-\theta}/2$  for  $\theta \geq 0$  and  $p(\theta) = e^\theta/2$  for  $\theta < 0$ , so the power function can easily be computed.

(c) The test  $\phi(\mathbf{X}) = I(X_1 > 0)$  has power function  $\beta_\phi(\theta) = P_\theta(X_1 > 0) = p(\theta)$ . Hence,  $\beta'_\phi(\theta) = e^{-|\theta|}/2$ , and  $\beta''_\phi(\theta)$  does not exist at  $\theta = 0$ . Thus the method for finding locally best unbiased tests does not work.

5.5.3. (a) If  $X_1, \dots, X_n$  is a sample from the logistic distribution with location parameter  $\theta$ , then  $\log f(x_1, \dots, x_n|\theta) = -\sum_{i=1}^n \log(2(1 + \cosh(x_i - \theta)))$ , and

$$\frac{\partial}{\partial\theta} \log f(x_1, \dots, x_n) \Big|_{\theta=0} = \sum_{i=1}^n \frac{\sinh(x_i)}{1 + \cosh(x_i)}.$$

If we let  $U = \sum_{i=1}^n \sinh(X_i)/(1 + \cosh(X_i))$ , then since the distribution of  $U$  is continuous, we may omit the  $\gamma(x)$  term in (5.78) and conclude that the test of  $H_0$  vs.  $H_1$  that rejects  $H_0$  if and only if  $U > k$ , is a locally best test of its size for any  $k > 0$ . As an aid in choosing  $k$  to achieve a preassigned size  $\alpha$ , we note that under  $H_0$  the distribution of  $\sinh(X_i)/(1 + \cosh(X_i))$  is uniform on the interval  $(-1, 1)$ . This follows upon noticing that the distribution function of the logistic distribution,  $\mathcal{L}(0, 1)$ , may be written as  $F(x) = (1/2)(1 + \sinh(x)/(1 + \cosh(x)))$  and using the fact that  $F(X) \in \mathcal{U}(0, 1)$ . Thus,  $U$  is the sum of  $n$  independent  $\mathcal{U}(-1, 1)$ 's.

(b) To find the locally best unbiased test, (5.88), we must compute

$$\begin{aligned} \frac{\partial^2}{\partial \theta^2} \log f(x_1, \dots, x_n) \Big|_{\theta=0} &= \sum_{i=1}^n \frac{-(1 + \cosh(x_i)) \cosh(x_i) + (\sinh(x_i))^2}{(1 + \cosh(x_i))^2} \\ &= - \sum_{i=1}^n \frac{\cosh(x_i) + 1}{(1 + \cosh(x_i))^2} = - \sum_{i=1}^n \frac{1}{1 + \cosh(x_i)}. \end{aligned}$$

The locally best unbiased test then rejects  $H_0$  if  $V > k_1 + k_2 U$ , where

$$V = - \sum_{i=1}^n \frac{1}{1 + \cosh(x_i)} + \left( \sum_{i=1}^n \frac{\sinh(x_i)}{1 + \cosh(x_i)} \right)^2$$

and where  $k_1$  and  $k_2$  are chosen so that the test is unbiased and has a preassigned size. However, the test will be unbiased if  $k_2$  is chosen equal to zero by the argument given at the bottom of page 239, because the distribution of  $V$  is symmetric about zero when  $H_0$  is true. Thus, the test that rejects  $H_0$  when  $V > k_1$  is a locally best unbiased test of its size.