

## Sample solution

1.(a)  $E(X_i) = \frac{\theta}{2}$ , so,  $E(\bar{X}_n) = \frac{\theta}{2}$ , and  $E(2\bar{X}_n) = \theta$

(b)  $P(Y_n \leq t) = P(X_1 \leq t \cdots X_n \leq t) = \left(\frac{t}{\theta}\right)^n$ , for  $0 < t < \theta$

pdf is then  $f_{Y_n}(t) = \frac{nt^{n-1}}{\theta^n}$ ,  $0 < t < \theta$ .

(c)  $E(Y_n) = \int_0^\theta t \cdot \frac{nt^{n-1}}{\theta^n} dt = \frac{n}{n+1} \theta$

$$\text{Bias}(Y_n) = E(Y_n) - \theta = -\frac{\theta}{n+1}$$

$E\left(\frac{n+1}{n} Y_n\right) = \theta$ , so it is an unbiased estimator.

2.(a) Under  $H_0$ ,  $\bar{X}_n$  is approximately  $N\left(0.5, \frac{0.5 \times 0.5}{n}\right)$

Let  $P(\bar{X}_n > c) = \alpha$ , then  $P\left(\frac{\bar{X}_n - 0.5}{\sqrt{\frac{0.5 \times 0.5}{n}}} \geq \frac{c - 0.5}{0.5 \sqrt{\frac{1}{n}}}\right) = \alpha$ .

Therefore,  $\frac{c - 0.5}{0.5 \sqrt{\frac{1}{n}}} = z_\alpha$ . So,  $c = 0.5 + z_\alpha \frac{0.5}{\sqrt{n}}$ , where  $z_\alpha = 1.96$ .

(b) When  $\theta = 0.7$ ,  $\bar{X}_n$  is approximately  $N\left(0.7, \frac{0.7 \times 0.3}{n}\right)$

$$\text{power} = P(\bar{X}_n > c | H_A)$$

$$= P\left(\frac{\bar{X}_n - 0.7}{\sqrt{\frac{0.7 \times 0.3}{n}}} > \frac{0.5 + z_\alpha \frac{0.5}{\sqrt{n}} - 0.7}{\sqrt{\frac{0.7 \times 0.3}{n}}}\right)$$

$$= 1 - \Phi\left(\frac{-0.2 + z_\alpha \frac{0.5}{\sqrt{n}}}{\sqrt{\frac{0.21}{n}}}\right), \text{ where } \Phi \text{ is the cdf of } N(0, 1)$$

(c)  $p\text{-value} = P(\bar{X}_n > 0.7 | H_0) = 1 - \Phi\left(\frac{0.7 - 0.5}{0.5 \sqrt{\frac{1}{25}}}\right) = 1 - \Phi(2) < 0.05$ .

So, reject  $H_0$ .

$$3. \quad S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

From Student's theorem,  $\frac{(n-1)S^2}{\sigma^2} \sim \chi_{n-1}^2$ .

$$\text{So, } P\left(\chi_{n-1, 1-\frac{\alpha}{2}}^2 < \frac{(n-1)S^2}{\sigma^2} < \chi_{n-1, \frac{\alpha}{2}}^2\right) = 1-\alpha.$$

Then, a  $(1-\alpha)$  CI of  $\sigma^2$  is  $\left(\frac{(n-1)S^2}{\chi_{n-1, \frac{\alpha}{2}}^2}, \frac{(n-1)S^2}{\chi_{n-1, 1-\frac{\alpha}{2}}^2}\right)$

$$4. (a) \quad \cancel{E(\hat{\theta}_n)} = E(X_i^2) = \text{Var}(X_i) + [E(X_i)]^2 = \theta$$

$$\text{As } \frac{X_i^2}{\theta} \sim \chi_1^2, \quad \text{Var}(X_i^2) = \theta^2 \cdot \text{Var}\left(\frac{X_i^2}{\theta}\right) = \theta^2 \cdot 2 = 2\theta^2$$

Then the result follows from the CLT as  $X_i^2$ 's are i.i.d.

$$(b) \text{ By the } \Delta\text{-method, } C = [g'(\theta)]^2 \cdot 2\theta^2 = \frac{1}{\theta^2} \cdot 2\theta^2 = 2.$$

$$(c) \quad g(\hat{\theta}_n) \pm z_{\alpha/2} \sqrt{\frac{2}{n}} = (L, U), \quad \alpha = 0.05$$

$$(d) \quad (e^L, e^U).$$