

## Exercise 7.104

Let  $Y_n$  be a binomial random variable with  $n$  trials and with success probability  $p$ . Suppose that  $n$  tends to infinity and  $p$  tends to zero in such a way that  $np$  remains fixed at  $np = \lambda$ . Prove that the distribution of  $Y_n$  converges to a Poisson distribution with mean  $\lambda$ .

## Exercise 8.8

Suppose that  $Y_1, Y_2, Y_3$  denote a random sample from an exponential distribution with density function

$$f(y) = \begin{cases} \left(\frac{1}{\theta}\right) e^{-y/\theta}, & y > 0 \\ 0, & \text{elsewhere} \end{cases}$$

Consider the following five estimators of  $\theta$ :

$$\hat{\theta}_1 = Y_1, \quad \hat{\theta}_2 = \frac{Y_1 + Y_2}{2}, \quad \hat{\theta}_3 = \frac{Y_1 + 2Y_2}{3}, \quad \hat{\theta}_4 = \min(Y_1, Y_2, Y_3), \quad \hat{\theta}_5 = \bar{Y}$$

- a Which of these estimators are unbiased?
- b Among the unbiased estimators, which has the smallest variance?

## Exercise 8.15

Let  $Y_1, Y_2, \dots, Y_n$  denote a random sample of size  $n$  from a population whose density is given by

$$f(y) = \begin{cases} 3\beta^3 y^{-4}, & \beta \leq y \\ 0, & \text{elsewhere} \end{cases}$$

where  $\beta > 0$  is unknown. (This is one of the Pareto distributions introduced in Exercise 6.18.) Consider the estimator  $\hat{\beta} = \min(Y_1, Y_2, \dots, Y_n)$ .

- a** Derive the bias of the estimator  $\hat{\beta}$ .
- b** Derive  $\text{MSE}(\hat{\beta})$ .

## Exercise 8.16

Suppose that  $Y_1, Y_2, \dots, Y_n$  constitute a random sample from a normal distribution with parameters  $\mu$  and  $\sigma^2$ .

- a** Show that  $S = \sqrt{S^2}$  is a biased estimator of  $\sigma$ . [Hint: Recall the distribution of  $(n-1)S^2/\sigma^2$ ]
- b** Adjust  $S$  to form an unbiased estimator of  $\sigma$ .
- c** Find an unbiased estimator of  $\mu - z_\alpha \sigma$ , the point that cuts off a lower-tail area of  $\alpha$  under this normal curve.

## Exercise 8.17

If  $Y$  has a binomial distribution with parameters  $n$  and  $p$ , then  $\hat{p}_1 = Y/n$  is an unbiased estimator of  $p$ . Another estimator of  $p$  is  $\hat{p}_2 = (Y + 1)/(n + 2)$ .

- a** Derive the bias of  $\hat{p}_2$ .
- b** Derive  $\text{MSE}(\hat{p}_1)$  and  $\text{MSE}(\hat{p}_2)$ .
- c** For what values of  $p$  is  $\text{MSE}(\hat{p}_1) > \text{MSE}(\hat{p}_2)$ ?

## Exercise 7.104

The mgf for  $Y_n$  is given by

$$m_{Y_n}(t) = [1 - p + pe^t]^n$$

Let  $p = \lambda/n$  and this becomes

$$m_{Y_n}(t) = \left[1 - \frac{\lambda}{n} + \frac{\lambda}{n}e^t\right]^n = \left[1 + \frac{1}{n}(\lambda e^t - 1)\right]^n.$$

As  $n \rightarrow \infty$ , this is  $\exp(\lambda e^t - 1)$ , the mgf for the Poisson with mean  $\lambda$ .

## Exercise 8.8

- a** Note that  $\hat{\theta}_1, \hat{\theta}_2, \hat{\theta}_3$  and  $\hat{\theta}_5$  are simple linear combinations of  $Y_1, Y_2$ , and  $Y_3$ . So, it is easily shown that all four of these estimators are unbiased. From Ex. 6.81 it was shown that  $\hat{\theta}_4$  has an exponential distribution with mean  $\theta/3$ , so this estimator is biased.
- b** It is easily shown that  $V(\hat{\theta}_1) = \theta^2$ ,  $V(\hat{\theta}_2) = \theta^2/2$ ,  $V(\hat{\theta}_3) = 5\theta^2/9$ , and  $V(\hat{\theta}_5) = \theta^2/9$ , so the estimator  $\hat{\theta}_5$  is unbiased and has the smallest variance.

## Exercise 8.15

Using standard techniques, it can be shown that  $E(Y) = (3/2)\beta$ ,  $E(Y^2) = 3\beta^2$ . Also it is easily shown that  $Y_{(1)}$  follows the Pareto family with density function

$$g_{(1)}(y) = 3n\beta^{3n}y^{-(3n+1)}, y \geq \beta.$$

Thus,  $E(Y_{(1)}) = \left(\frac{3n}{3n-1}\right)\beta$  and  $E(Y_{(1)}^2) = \frac{3n}{3n-2}\beta^2$ .

**a** With  $\hat{\beta} = Y_{(1)}$ ,  $B(\hat{\beta}) = \left(\frac{3n}{3n-1}\right)\beta - \beta = \left(\frac{1}{3n-1}\right)\beta$ .

**b** Using the above,  $\text{MSE}(\hat{\beta}) = \text{MSE}(Y_{(1)}) = E(Y_{(1)}^2) - 2\beta E(Y_{(1)}) + \beta^2 = \frac{2}{(3n-1)(3n-2)}\beta^2$ .

## Exercise 8.16

It is known that  $(n-1)S^2/\sigma^2$  is chi-square with  $n-1$  degrees of freedom.

**a**  $E(S) = E\left\{\frac{\sigma}{\sqrt{n-1}}\left[\frac{(n-1)S^2}{\sigma^2}\right]^{1/2}\right\} = \frac{\sigma}{\sqrt{n-1}} \int_0^\infty v^{1/2} \frac{1}{\Gamma[(n-1)/2]2^{(n-1)/2}} v^{(n-3)/2} e^{-v/2} dv = \frac{\sigma}{\sqrt{n-1}} \frac{\sqrt{2}\Gamma(n/2)}{\Gamma[(n-1)/2]}.$

**b** The estimator  $\hat{\sigma} = \frac{\sqrt{n-1}\Gamma(n/2)}{\sqrt{2}\Gamma[(n-1)/2]}S$  is unbiased for  $\sigma$ .

**c** Since  $E(\bar{Y}) = \mu$ , the unbiased estimator of the quantity is  $\bar{Y} - z_\alpha \hat{\sigma}$ .

## Exercise 8.17

It is given that  $\hat{p}_1$  is unbiased, and since  $E(Y) = np$ ,  $E(\hat{p}_2) = (np+1)/(n+2)$ .

**a**  $B(\hat{p}_2) = (np+1)/(n+2) - p = (1-2p)/(n+2)$ .

**b** Since  $\hat{p}_1$  is unbiased,  $\text{MSE}(\hat{p}_1) = V(\hat{p}_1) = p(1-p)/n$ ,  $\text{MSE}(\hat{p}_2) = V(\hat{p}_2) + B^2(\hat{p}_2) = \frac{np(1-p) + (1-2p)^2}{(n+2)^2}$ .

**c** Considering the inequality

$$\frac{np(1-p) + (1-2p)^2}{(n+2)^2} < \frac{p(1-p)}{n}$$

this can be written as

$$(8n+4)p^2 - (8n+4)p + n < 0$$

Solving for  $p$  using the quadratic formula, we have

$$p = \frac{8n+4 \pm \sqrt{(8n+4)^2 - 4(8n+4)n}}{2(8n+4)} = \frac{1}{2} \pm \sqrt{\frac{n+1}{8n+4}}.$$

So  $\text{MSE}(\hat{p}_1) > \text{MSE}(\hat{p}_2)$  for  $\frac{1}{2} - \sqrt{\frac{n+1}{8n+4}} < p < \frac{1}{2} + \sqrt{\frac{n+1}{8n+4}}$ .