

**M3S3/M4S3**  
**ASSESSED COURSEWORK 1 : SOLUTIONS**

1. We use the result from the handout

$$\sqrt{n} \left( \begin{pmatrix} X_{(k_1)} \\ X_{(k_2)} \end{pmatrix} - \begin{pmatrix} x_{p_1} \\ x_{p_2} \end{pmatrix} \right) \xrightarrow{\mathcal{L}} N \left( 0, \begin{bmatrix} \frac{p_1(1-p_1)}{\{f_X(x_{p_1})\}^2} & \frac{p_1(1-p_2)}{f_X(x_{p_1})f_X(x_{p_2})} \\ \frac{p_1(1-p_2)}{f_X(x_{p_1})f_X(x_{p_2})} & \frac{p_2(1-p_2)}{\{f_X(x_{p_2})\}^2} \end{bmatrix} \right)$$

which implies, marginally, that, for arbitrary  $p$ , and  $k = \lceil np \rceil$ ,

$$\sqrt{n}(X_{(k)} - x_p) \xrightarrow{\mathcal{L}} N \left( 0, \frac{p(1-p)}{\{f_X(x_p)\}^2} \right)$$

(i) For the **sample median**,  $k = \lceil np \rceil$  with  $p = 0.5$ , we have

$$\sqrt{n}(X_{(k)} - x_{0.5}) \xrightarrow{\mathcal{L}} N \left( 0, \frac{0.5(1-0.5)}{\{f_X(x_{0.5})\}^2} \right)$$

Now, in the  $N(\mu, \sigma^2)$  case,  $x_p = \mu$ , and  $\{f_X(\mu)\}^2 = 1/(2\pi\sigma^2)$ , so we have

$$\sqrt{n}(X_{(k)} - \mu) \xrightarrow{\mathcal{L}} N \left( 0, \frac{0.5(1-0.5)}{1/(2\pi\sigma^2)} \right) \equiv N \left( 0, \frac{\pi\sigma^2}{2} \right)$$

and hence

$$X_{(k)} \sim AN \left( \mu, \frac{\pi\sigma^2}{2n} \right).$$

Note that  $\pi/2 \approx 1.57 > 1$ , so the asymptotic variance is greater for the sample **median** than for the sample **mean**.

[5 MARKS]

(ii) For the **sample interquartile range**,  $R_{IQ}$ ;  $k_1 = \lceil np_1 \rceil$  and  $k_2 = \lceil np_2 \rceil$  with  $p_1 = 0.25$  and  $p_2 = 0.75$ . Then from above

$$\sqrt{n} \left( \begin{pmatrix} X_{(k_1)} \\ X_{(k_2)} \end{pmatrix} - \begin{pmatrix} x_{0.25} \\ x_{0.75} \end{pmatrix} \right) \xrightarrow{\mathcal{L}} N \left( 0, \begin{bmatrix} \frac{0.25(1-0.25)}{\{f_X(x_{0.25})\}^2} & \frac{0.25(1-0.75)}{f_X(x_{0.25})f_X(x_{0.75})} \\ \frac{0.25(1-0.75)}{f_X(x_{0.25})f_X(x_{0.75})} & \frac{0.75(1-0.75)}{\{f_X(x_{0.25})\}^2} \end{bmatrix} \right)$$

Now, by the results given for the standard normal

$$x_{0.25} = \mu - 0.674\sigma \quad x_{0.75} = \mu + 0.674\sigma$$

and so, by elementary transformation theory

$$\begin{aligned} f_X(x_{0.25}) &= f_X(\mu - 0.674\sigma) = \phi(-0.674)/\sigma = 0.318/\sigma \\ f_X(x_{0.75}) &= f_X(\mu + 0.674\sigma) = \phi(0.674)/\sigma = 0.318/\sigma \end{aligned}$$

Thus, the asymptotic variance-covariance matrix is

$$\Sigma = \frac{\sigma^2}{0.318^2} \begin{bmatrix} 0.25 \times 0.75 & 0.25 \times 0.25 \\ 0.25 \times 0.25 & 0.75 \times 0.25 \end{bmatrix} = \sigma^2 \begin{bmatrix} 1.854 & 0.618 \\ 0.618 & 1.854 \end{bmatrix}$$

Now, setting the vector  $\mathbf{a} = (-1, 1)^\top$  yields that

$$R_{IQ} = X_{(k_2)} - X_{(k_1)} = \mathbf{a}^\top (X_{(k_1)}, X_{(k_2)})^\top$$

and hence (by continuous mapping/Slutsky)

$$\sqrt{n}(R_{IQ} - r_{IQ}) \xrightarrow{\mathcal{L}} N\left(0, \mathbf{a}^\top \Sigma \mathbf{a}\right)$$

where  $r_{IQ} = x_{0.75} - x_{0.25}$ . Here

$$r_{IQ} = x_{0.75} - x_{0.25} = 2 \times 0.674\sigma = 1.348\sigma$$

and

$$\mathbf{a}^\top \Sigma \mathbf{a} = 2.472\sigma^2$$

and hence

$$R_{IQ} \sim AN\left(1.348\sigma, \frac{2.472\sigma^2}{n}\right).$$

[10 MARKS]

2. We have

$$\begin{aligned} d_H(f_1, f_2) &= \int_{-\infty}^{\infty} \left(\sqrt{f_1(x)} - \sqrt{f_2(x)}\right)^2 dx = \int_{-\infty}^{\infty} \left(f_1(x) + f_2(x) - 2\sqrt{f_1(x)f_2(x)}\right) dx \\ &= 2 - 2 \int_{-\infty}^{\infty} \sqrt{f_1(x)f_2(x)} dx \\ &\leq 2 \end{aligned}$$

as

$$\int_{-\infty}^{\infty} \sqrt{f_1(x)f_2(x)} dx \geq 0.$$

[5 MARKS]