## M2S1 : ASSESSED COURSEWORK 3 : SOLUTIONS

## 1. (a) Using the results given in lectures: If

$$Y = AX$$
 where  $A = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & 4 \end{pmatrix}$ .

and  $X \sim N(\mathbf{0}, I_3)$ , where **0** is the zero vector, and  $I_3$  is the  $3 \times 3$  identity matrix, then

$$Y \sim N\left(A\mathbf{0}, AI_3 A^T\right) \equiv N\left(\mathbf{0}, \Sigma\right)$$

where

$$\Sigma = AA^{T} = \begin{pmatrix} 1 & 0 & 1 \\ 0 & 2 & -1 \\ 0 & 0 & 4 \end{pmatrix} \begin{pmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 1 & -1 & 4 \end{pmatrix} = \begin{pmatrix} 2 & -1 & 4 \\ -1 & 5 & -4 \\ 4 & -4 & 16 \end{pmatrix}$$

[4 MARKS]

(b) The covariance between  $Y_1$  and  $Y_3$  is given by  $\Sigma_{13}$ , that is,  $Cov_{f_{Y_1,Y_3}}[Y_1,Y_3] = 4$ .

[4 MARKS]

(c) By the transformation result from (a), the marginal distribution of  $Y_1$  is obtained by the result

$$Y_1 = BY \sim N \left( B\mathbf{0}, B\Sigma B^T \right)$$

where  $B\Sigma B^T = \Sigma_{11} = 2$  and hence

 $Y_1 \sim N\left(0,2\right)$ 

(can deduce this directly, using MGFs, and the definition of  $Y_1 = X_1 + X_3$ ).

[2 MARKS]

(d) We have for the mgf for X, by using the mgf of the standard normal  $\exp\{t^2/2\}$ 

$$M_X(\mathbf{t}) = E_{f_X}\left[\exp\left\{\mathbf{t}^T X\right\}\right] = E_{f_X}\left[\exp\left\{t_1 X_1 + t_2 X_2 + t_3 X_3\right\}\right] = \prod_{i=1}^3 \exp\left\{t_i X_i\right\}$$
(by indefinition)

(by independence)

$$= \prod_{i=1}^{3} M_{X_{i}}(t_{i}) = \exp\left\{\frac{t_{1}^{2}}{2}\right\} \exp\left\{\frac{t_{2}^{2}}{2}\right\} \exp\left\{\frac{t_{3}^{2}}{2}\right\} = \exp\left\{\frac{\mathbf{t}^{T}\mathbf{t}}{2}\right\}$$
$$M_{Y}(\mathbf{t}) = E_{f_{Y}}\left[\exp\left\{\mathbf{t}^{T}Y\right\}\right] = E_{f_{X}}\left[\exp\left\{\mathbf{t}^{T}(AX)\right\}\right]$$
$$= E_{f_{X}}\left[\exp\left\{\mathbf{t}^{T}(A^{T})^{T}X\right\}\right] = E_{f_{X}}\left[\exp\left\{\left(A^{T}\mathbf{t}\right)^{T}X\right\}\right]$$
$$= M_{X}\left(\left(A^{T}\mathbf{t}\right)\right) = \exp\left\{\frac{\left(A^{T}\mathbf{t}\right)^{T}(A^{T}\mathbf{t})}{2}\right\} = \exp\left\{\frac{\mathbf{t}^{T}(AA^{T})\mathbf{t}}{2}\right\} = \exp\left\{\frac{\mathbf{t}^{T}\Sigma\mathbf{t}}{2}\right\}$$

[2 MARKS, 3 MARKS]

(e) We have

$$V = Y^{T} \Sigma^{-1} Y = (AX)^{T} (AA^{T})^{-1} AX = X^{T} A^{T} (A^{T})^{-1} A^{-1} AX = X^{T} (A^{T} (A^{T})^{-1}) (A^{-1}A) X$$
$$= X^{T} X = X_{1}^{2} + X_{2}^{2} + X_{3}^{2}.$$

Now,  $X_1, X_2$  and  $X_3$  are all Normal(0, 1), and thus for i = 1, 2, 3, from lecture notes

$$V_i = X_i^2 \sim \chi_1^2 \equiv Gamma\left(\frac{1}{2}, \frac{1}{2}\right) \qquad \therefore \qquad V = \sum_{i=1}^3 V_i \sim Gamma\left(\frac{3}{2}, \frac{1}{2}\right) \equiv \chi_3^2$$

using mgfs, or the result for the addition of independent Gamma random variables with the same  $\beta$  parameter.

2. (a) We have for y > 0

$$F_{Y_n}(y) = \{F_X(y)\}^n = \left\{\frac{(y+1)^2 - 1}{(y+1)^2}\right\}^n = \left\{1 - \frac{1}{(y+1)^2}\right\}^n$$

with  $F_{Y_n}(y) = 0$  for  $y \leq 0$ .

(i) For any (fixed)  $y > 0, 0 < (y+1)^{-2} < 1$ , and hence

$$\left\{1 - \frac{1}{\left(y+1\right)^2}\right\}^n \to 0 \qquad \text{as } n \to \infty \qquad \therefore F_{Y_n}\left(y\right) \to F(y) = 0$$

for all  $y \in \mathbb{R}$ , which is **not** a probability distribution function, so the limiting distribution **does not exist**.

[5 MARKS]

(ii) For the transformed variable  $Z_n = Y_n / \sqrt{n}$ , for z > 0,

$$F_{Z_n}(z) = P\left[Z_n \le z\right] = P\left[Y_n/\sqrt{n} \le z\right] = P\left[Y_n \le \sqrt{n}z\right] = F_{Y_n}\left(\sqrt{n}z\right).$$
  
$$\therefore F_{Z_n}(z) = \left\{1 - \frac{1}{\left(\sqrt{n}z + 1\right)^2}\right\}^n = \left\{1 - \frac{1}{\left(nz^2 + 2\sqrt{n}z + 1\right)}\right\}^n = \left\{1 - \frac{1}{n}\frac{1}{\left(z^2 + 2z/\sqrt{n} + 1/n\right)}\right\}^n$$

In the limit as  $n \to \infty$ , terms in the denominator of the bracketed expression tend to zero with n at a fast enough rate to preserve the result that

$$\lim_{n \to \infty} \left\{ 1 - \frac{1}{n} \frac{1}{(z^2 + 2z/\sqrt{n} + 1/n)} \right\}^n = \lim_{n \to \infty} \left\{ 1 - \frac{1}{nz^2} \right\}^n = e^{-1/z^2}$$

as  $n \to \infty$ , as in the Central Limit Theorem proof from lectures. Thus, for z > 0

$$F_{Z_n}(z) \to F_Z(z) = \exp\left\{-\frac{1}{z^2}\right\} \qquad (\text{and } F_{Z_n}(z) \to 0 \text{ for } z \le 0)$$

which is a valid cdf, the limiting distribution exists (and, in fact, is continuous).

[5 MARKS]

M2S1 ASSESSED COURSEWORK 3: SOLUTIONS - page 2 of 3

(b) Now suppose that  $X_1, ..., X_n$  are independent *Exponential* (1) random variables.

(i) We have

$$F_{Y_n}(y) = \{F_X(y)\}^n = \{1 - e^{-y}\}^n \quad \text{for } z > 0$$

and is zero for  $z \leq 0$ .

(ii) For  $Z_n = Y_n - a$ , we have that the range of  $Z_n$  is  $(-a, \infty)$ 

$$F_{Z_n}(z) = P[Z_n \le z] = P[Y_n - a \le z] = P[Y_n \le z + a] = F_{Y_n}(z + a)$$

and hence

$$F_{Z_n}(z) = \left\{1 - e^{-(z+a)}\right\}^n$$

for z > -a, and  $F_{Z_n}(z) = 0$  for  $z \le -a$ .

[2 MARKS]

[2 MARKS]

[2 MARKS]

(iii) For constants  $\{a_n\}$ 

$$F_{Z_n}(z) = \left\{1 - e^{-(z+a_n)}\right\}^n = \left\{1 - e^{-a_n}e^{-z}\right\}^n = \left\{1 - A_n e^{-z}\right\}^n$$

where  $A_n = e^{-a_n}$ . Now, if  $A_n = 1/n$  (so that  $a_n = \log n$ ), we have, for z > 0,

$$F_{Z_n}(z) = \left\{1 - \frac{e^{-z}}{n}\right\}^n \to \exp\left\{-e^{-z}\right\}$$

as  $n \to \infty$ . This is a valid cdf for a random variable Z on  $\mathbb{R}$ . Hence, for large n

$$F_{Z_n}(z) = P[Z_n \le z] \approx \exp\left\{-e^{-z}\right\}$$

and therefore

 $P[Y_n \le y_0] = P[Z_n - a_n \le y_0] = P[Z_n \le y_0 + a_n] = F_{Z_n}(y_0 + a_n) \approx \exp\{-\exp\{-(y_0 + a_n)\}\}$ so that

$$P[Y_n > y_0] \approx 1 - \exp\{-\exp\{-(y_0 + a_n)\}\}.$$

(iv) Here we have the maximum **observed** between-earthquake time to be 1021 days. Let  $X_1, ..., X_n$  (n = 199) be the random between-earthquake times and  $Y_n$  be the (random) maximum value. We have from above that

$$F_{Y_n}(y) = \{F_X(y)\}^n = \left\{1 - e^{-\lambda y}\right\}^n \qquad \therefore \qquad P[Y_n > y] = 1 - \left\{1 - e^{-\lambda y}\right\}^n$$

In this case,  $\lambda = 0.01$ , so

$$P[Y_n > y] = 1 - \{1 - e^{-0.01y}\}^n$$

But we have **observed**  $Y_n = 1021$ , and under this model

$$P[Y_n > 1021] = 1 - \left\{1 - e^{-0.01 \times 1021}\right\}^{199} = 0.0073$$

which is an unlikely outcome, given the proposed model. Hence it is probable that the proposed model is incorrect.

Re-doing the calculation with n = 200 (this is acceptable, the question was a little ambiguous), we have again  $P[Y_n > 1021] = 0.0073$ , so there is no significant difference in the conclusion.

[4 MARKS]