# NATIONAL UNIVERSITY OF SINGAPORE MATHEMATICS SOCIETY

# PAST YEAR PAPER SOLUTIONS with credits to Teo Wei Hao

## ST2131/MA2216 Probability AY 2004/2005 Sem 1

### SECTION A

#### Question 1

(a) (i) Let  $X_i$  be i.i.d. r.v. of the lifetime in hours of the *i*-th component used,  $i \in \mathbb{Z}^+$ . We are also given that  $\mu = 100$  and  $\sigma = 30$ . Thus by Central Limit Theorem, we have  $\sum_{i=1}^{30} X_i \approx N(3000, 27000)$ . As such,

$$\mathbb{P}\left\{\sum_{i=1}^{30} X_i \ge 2900\right\} \approx \mathbb{P}\left\{Z \ge \frac{2900 - 3000}{\sqrt{27000}}\right\} = \mathbb{P}\left\{Z \ge -0.61\right\}$$
$$= 1 - 0.2709 = 0.7291.$$

(ii) Let n be the number of components required. Then we see that n > 30, and thus we can use Central Limit Theorem to get  $\sum_{i=1}^{n} X_i \approx N(100n, 900n)$ . By referring to the statistical table, we obtain,

$$\mathbb{P}\{Z \ge -1.6449\} = 0.95 \le \mathbb{P}\{\sum_{i=1}^{n} X_i \ge 2900\}$$

$$\approx \mathbb{P}\left\{Z \ge \frac{2900 - 100n}{\sqrt{900n}}\right\}.$$

Thus we conclude that  $-1.6449 \ge \frac{2900-100n}{\sqrt{900n}}$ . Let  $u = \sqrt{900n}$ , i.e.  $n = \frac{u^2}{900}$ . This give us,

$$-1.6449 \ge \frac{2900 - \frac{1}{9}u^2}{u}$$
$$\frac{1}{9}u^2 - 1.6449u - 2900 \ge 0.$$

We solve the above quadratic inequality with  $u \ge 0$ , and substituting back to get  $n \ge 31.7820$ . Therefore at least 32 components must be in stock.

(b) Let  $A_1, A_2, A_3$  be the event that the two-headed coin, the fair coin, and the biases coin is selected. Let B be the event that the selected coin shows head when flipped. Then by Baye's rule,

$$\mathbb{P}(A_1 \mid B) = \frac{\mathbb{P}(A_1 B)}{\mathbb{P}(B)} = \frac{\mathbb{P}(B \mid A_1)\mathbb{P}(A_1)}{\mathbb{P}(B \mid A_1)\mathbb{P}(A_1) + \mathbb{P}(B \mid A_2)\mathbb{P}(A_2) + \mathbb{P}(B \mid A_3)\mathbb{P}(A_3)} \\
= \frac{(1)(\frac{1}{3})}{(1)(\frac{1}{3}) + (\frac{1}{2})(\frac{1}{3}) + (\frac{3}{4})(\frac{1}{3})} = \frac{4}{9}.$$

## Question 2

(a) We have for x > 0,

$$f_X(x) = \int_{\mathbb{R}} f(x, y) \ dy = \int_0^\infty x e^{-x(y+1)} \ dy$$
  
=  $\left[ -e^{-x(y+1)} \right]_0^\infty = e^{-x}.$ 

Now for y > 0, we have,

$$f_Y(y) = \int_{\mathbb{R}} f(x,y) \ dx = \int_0^\infty x e^{-x(y+1)} \ dx$$
$$= \left[ \frac{-x}{y+1} e^{-x(y+1)} \right]_0^\infty - \int_0^\infty \frac{-1}{y+1} e^{-x(y+1)} \ dx$$
$$= \left[ \frac{-1}{(y+1)^2} e^{-x(y+1)} \right]_0^\infty = \frac{1}{(y+1)^2}.$$

Thus the marginal p.d.f. of X and Y are given by,

$$f_X(x) = \begin{cases} e^{-x}, & x > 0; \\ 0, & \text{otherwise,} \end{cases}$$

$$f_Y(y) = \begin{cases} \frac{1}{(y+1)^2}, & y > 0; \\ 0, & \text{otherwise.} \end{cases}$$

(b) When x > 0, y > 0, we have,  $f_{Y|X}(y|x) = \frac{f(x,y)}{f_X(x)} = \frac{xe^{-x(y+1)}}{e^{-x}} = xe^{-xy}$ . Thus the conditional p.d.f. of Y given that X = x is,

$$f_{Y|X}(y|x) = \begin{cases} xe^{-xy}, & x > 0, y > 0; \\ 0, & \text{otherwise.} \end{cases}$$

(c) We have,

$$E(Y \mid X) = \int_{\mathbb{R}} y f_{Y|X}(y|x) \ dy = \int_{0}^{\infty} xy e^{-xy} \ dy$$
$$= \left[ -y e^{-xy} \right]_{0}^{\infty} - \int_{0}^{\infty} -e^{-xy} \ dy$$
$$= \left[ -\frac{1}{x} e^{-xy} \right]_{0}^{\infty} = \frac{1}{x}.$$

(d) Let W = XY. Then for w > 0, we have,

$$F_{W}(w) = \mathbb{P}\{XY \le w\} = \int_{\mathbb{R}} \mathbb{P}\{xY \le w \mid X = x\} f_{X}(x) \ dx$$

$$= \int_{\mathbb{R}} \mathbb{P}\left\{Y \le \frac{w}{x} \mid X = x\right\} f_{X}(x) \ dx$$

$$= \int_{0}^{\infty} \left(\int_{0}^{\frac{w}{x}} x e^{-xy} \ dy\right) e^{-x} \ dx$$

$$= \int_{0}^{\infty} \left[-e^{-xy}\right]_{0}^{\frac{w}{x}} e^{-x} \ dx$$

$$= \int_{0}^{\infty} \left(1 - e^{-w}\right) e^{-x} \ dx$$

$$= \left(1 - e^{-w}\right) \left[-e^{-x}\right]_{0}^{\infty} = 1 - e^{-w}.$$

Thus  $f_W(w) = \frac{d}{dw} F_W(w) = e^{-w}$ . Therefore the p.d.f. of XY is,

$$f_{XY}(w) = \begin{cases} e^{-w}, & w > 0; \\ 0, & \text{otherwise.} \end{cases}$$

#### Question 3

(a) We are given that  $U \sim U(0,2\pi)$  and  $V \sim \operatorname{Exp}(1)$ . Let  $x = \sqrt{2v} \cos u$  and  $y = \sqrt{2v} \sin u$ . This give us  $u = \tan^{-1} \frac{y}{x}$  and  $v = \frac{x^2 + y^2}{2}$ . Now  $\frac{\partial x}{\partial u} = -\sqrt{2v} \sin u$ ,  $\frac{\partial x}{\partial v} = \frac{1}{\sqrt{2v}} \cos u$ ,  $\frac{\partial y}{\partial u} = \sqrt{2v} \cos u$  and  $\frac{\partial y}{\partial v} = \frac{1}{\sqrt{2v}} \sin u$ . Thus  $J(u,v) = \left(-\sqrt{2v} \sin u\right) \left(\frac{1}{\sqrt{2v}} \sin u\right) - \left(\frac{1}{\sqrt{2v}} \cos u\right) \left(\sqrt{2v} \cos u\right) = -1$ , and so |J(u,v)| = 1. Together with the fact that U and V are independent, we have,

$$f_{(X,Y)}(x,y) = \frac{1}{|J(u,v)|} f_{(U,V)}(u,v)$$

$$= (1) f_U(u) f_V(v)$$

$$= \frac{1}{2\pi} e^{-v}$$

$$= \frac{1}{2\pi} e^{-\frac{x^2 + y^2}{2}}.$$

Thus the joint p.d.f. of X and Y is given by,

$$f_{(X,Y)}(x,y) = \frac{1}{2\pi}e^{-\frac{x^2+y^2}{2}}, \quad x,y \in \mathbb{R}.$$

(b) Notice that  $f_{(X_1,X_2)}(x_1,x_2) = \left(\frac{1}{\sqrt{2\pi}}e^{-\frac{1}{2}x_1^2}\right)\left(\frac{1}{\sqrt{2\pi}}e^{-\frac{1}{2}x_2^2}\right) = \frac{1}{2\pi}e^{-\frac{x_1^2+x_2^2}{2}}$ . Then from what we found in (3a.), we see that W = V, and thus  $W \sim \text{Exp}(1)$ . This give us the p.d.f. of W to be,

$$f_W(w) = \begin{cases} e^{-w}, & w > 0; \\ 0, & \text{otherwise.} \end{cases}$$

#### **SECTION B**

### Question 4

(a) We are given that X and Y are independent r.v. such that  $X, Y \sim U(0, 1)$ . When  $0 < w \le 1$ , we have,

$$f_W(w) = \int_{\mathbb{R}} f_X(w - y) f_Y(y) \ dy = \int_0^1 f_X(w - y) (1) \ dy$$
$$= \int_0^w f_X(w - y) \ dy + \int_w^1 f_X(w - y) \ dy$$
$$= \int_0^w 1 \ dy$$
$$= w.$$

This give us  $F_W(w) = \int_{-\infty}^w f_W(w) \ dw = \int_0^w f_W(w) \ dw = \frac{1}{2}w^2$ . When  $1 < w \le 2$ , we have,

$$f_W(w) = \int_{\mathbb{R}} f_X(w - y) f_Y(y) \ dy = \int_0^1 f_X(w - y) (1) \ dy$$
$$= \int_0^{w-1} f_X(w - y) \ dy + \int_{w-1}^1 f_X(w - y) \ dy$$
$$= \int_{w-1}^1 1 \ dy$$
$$= 2 - w.$$

This give us  $F_W(w) = \int_{-\infty}^w f_W(w) \ dw = \frac{1}{2} + \int_1^w f_W(w) \ dw = 2w - \frac{1}{2}w^2 - 1$ . Therefore the c.d.f. of W is,

$$F_W(w) = \begin{cases} 0, & w \le 0; \\ \frac{1}{2}w^2, & 0 < w \le 1; \\ 2w - \frac{1}{2}w^2 - 1, & 1 < w \le 2; \\ 1, & w > 2. \end{cases}$$

(b) (i) Let  $W = X_2 + X_3$ . Using the result of (4a.) and the fact that  $X_1, X_2, X_3$  are i.i.d. r.v., we get,

$$\mathbb{P}\{X_1 > X_2 + X_3\} = \mathbb{P}\{W < X_1\} = \int_{\mathbb{R}} \mathbb{P}\{W < x_1\} f_{X_1}(x_1) dx_1 
= \int_0^1 f_W(x_1)(1) dx_1 
= \int_0^1 \frac{1}{2} x_1^2 dx_1 
= \left[\frac{1}{6} x_1^3\right]_0^1 = \frac{1}{6}.$$

(ii) Let  $A_i$  be the events that  $X_i$  is larger than the sum of the other two respectively, for i = 1, 2, 3. Let A be the event that the largest of the three is larger than the sum of the other two. Notice that  $A = A_1 \cup A_2 \cup A_3$  and the  $A_i$ 's are mutually exclusive. This give us,

$$\mathbb{P}(A) = \sum_{i=1}^{3} \mathbb{P}(A_i) = \sum_{i=1}^{3} \frac{1}{6} = \frac{1}{2}.$$

#### Question 5

(a) Let  $f_X(1) = p$  and  $\mathbb{P}\{X > 1\} = q$ . Notice that 0 < p, q < 1 such that p + q = 1. We would like to prove that  $\mathbb{P}\{X > x\} = q^x, x \in \mathbb{Z}^+$ .

Let  $P_n$  be the statement that  $\mathbb{P}\{X > n\} = q^n$ ,  $n \in \mathbb{Z}^+$ .

We have  $P_1$  given to be true.

Assume that  $P_k$  is true for some  $k \in \mathbb{Z}^+$ . Then,

$$\begin{split} \mathbb{P}\{X > k+1\} &= \mathbb{P}\{X > k+1 \mid X > k\} \mathbb{P}\{X > k\} \\ &= \mathbb{P}\{X > 1\} \mathbb{P}\{X > k\} \\ &= q \cdot q^k = q^{k+1}. \end{split}$$

i.e.  $P_{k+1}$  is true.

Therefore by Mathematical Induction, we have  $\mathbb{P}\{X > n\} = q^n$  for all  $n \in \mathbb{Z}^+$ .

Thus,  $f_X(x) = \mathbb{P}\{X > x\} - \mathbb{P}\{X > x + 1\} = q^x(1 - q) = q^x p$ , i.e.  $X \sim \text{Geom}(p)$ . Therefore the mean of X is  $\frac{1}{p}$ .

(b) Let  $Y_i$  be the r.v. of the number of rolls required to get the *i*-th new number. For example, if in 5 rolls the numbers obtained are (5, 2, 2, 1, 1), then  $Y_3 = 4$ , since 1 is the third number appearing. Let  $X_1 = Y_1$ , and  $X_i = Y_i - Y_{i-1}$  for i = 2, 3, 4, 5, 6.

Since we have a fair dice,  $X_i \sim \text{Geom}(p)$ , where  $p = \frac{\text{number of sides not obtained yet}}{\text{total sides}} = \frac{7-i}{6}$ . Also let X be the r.v. of the number of rolls before all 6 sides appeared at least once. Then we have  $X = \sum_{i=1}^{6} X_i$ . Therefore,

$$E(X) = E\left(\sum_{i=1}^{6} X_i\right) = \sum_{i=1}^{6} E(X_i)$$
$$= \sum_{i=1}^{6} \frac{6}{7-i} = \frac{147}{10},$$

i.e. 14.7 rolls are expected to get the result we wanted.

#### Question 6

- (i) Since each rolls are mutually independent, we have  $\mathbb{P}\{X_i=1\}=\mathbb{P}\{Y_i=1\}=\frac{1}{6},\ i=1,2,\ldots,n$ .
- (ii) If  $i \neq j$ , then similarly as above, we get  $\mathbb{P}\{X_i = 1, Y_j = 1\} = \mathbb{P}\{X_i = 1\}\mathbb{P}\{Y_j = 1\} = \left(\frac{1}{6}\right)\left(\frac{1}{6}\right) = \frac{1}{36}$ . If i = j, then since it is impossible to get 1 and 2 together in a roll,  $\mathbb{P}\{X_i = 1, Y_j = 1\} = 0$ .
- (iii) We have,  $Cov(X_i, Y_j) = E(X_i Y_j) E(X_i) E(Y_j) = \mathbb{P}\{X_i = 1, Y_j = 1\} \mathbb{P}\{X_i = 1\}\mathbb{P}\{Y_i = 1\}.$ Thus when  $i \neq j$ , then  $Cov(X_i, Y_j) = 0$ , else  $Cov(X_i, Y_j) = -\frac{1}{36}$ .

$$\operatorname{Cov}(X,Y) = \operatorname{Cov}\left(\sum_{i=1}^{n} X_{i}, \sum_{j=1}^{n} Y_{j}\right) = \left(\sum_{i=1}^{n} \operatorname{Cov}(X_{i}, Y_{i})\right) + \left(\sum_{\substack{1 \leq i, j \leq n \\ i \neq j}} \operatorname{Cov}(X_{i}, Y_{j})\right)$$
$$= \sum_{i=1}^{n} \left(-\frac{1}{36}\right) + 0 = -\frac{n}{36}.$$

Page: 5 of 5