

## BSc, MSci and MSc EXAMINATIONS (MATHEMATICS)

May – June 2013

This paper is also taken for the relevant examination for the Associateship of the  
Royal College of Science.

**Probability & Statistics I**

Date: Wednesday, 22 May 2013. Time: 10.00am. Time allowed: 2 hours.

This paper has FOUR questions.

Candidates should start their solutions to each question in a new main answer book

Supplementary books may only be used after the relevant main book(s) are full.

Statistical tables will not be provided.

Formula sheets are provided on pages 4 & 5

- DO NOT OPEN THIS PAPER UNTIL THE INVIGILATOR TELLS YOU TO.
- Answer all the questions. Each question carries equal weight.
- Credit will be given for all questions attempted, but extra credit will be given for complete or nearly complete answers.
- Affix one of the labels provided to each answer book that you use, but DO NOT USE THE LABEL WITH YOUR NAME ON IT.
- Calculators may not be used.

1. (a) State the axioms of probability for events defined on a sample space  $\Omega$ .
- (b) For events  $E \subseteq \Omega$  and  $F \subseteq \Omega$ , prove from the axioms that  $P(E) \geq P(E \cap F)$ .
- (c) Events  $E_1, E_2, E_3 \subseteq \Omega$  are such that,  $E_1$  and  $E_3$  are disjoint,  $E_1$  and  $E_2$  are independent and  $E_2$  and  $E_3$  are independent. Given  $P(E_i) = 1/2^i$ ,  $i = 1, 2, 3$ , determine the probability that out of  $E_1, E_2$  and  $E_3$ , only  $E_2$  occurs.
- (d) The discrete random variable  $X$  has range  $\{0, 1, 2, \dots\}$ . Prove that if  $E_{f_X}(X) = 0$  then  $P(X = 0) = 1$ .
- (e) The discrete random variable  $X$  has the following probability mass function (pmf):

$$f_X(-2) = \frac{1}{8}; \quad f_X(-1) = \frac{1}{8}; \quad f_X(0) = \frac{1}{8}; \quad f_X(1) = \frac{1}{4}, \quad f_X(2) = \frac{3}{8}.$$

- (i) Determine the pmf of  $Y = X + 2$ .
- (ii) Determine the pmf of  $Z = X^2$ .
- (f) The continuous random variables  $\alpha$  and  $\beta$  are both identically and independently distributed with a uniform distribution on the interval  $[-1, 1]$ . Consider the quadratic equation given by

$$x^2 + \alpha x + \beta = 0.$$

Determine the probability that the roots of this equation are real.

2. The continuous random variable  $X$  has Moment Generating Function (MGF),  $M_X(t)$ . Consider the random variable  $Y = \mu + \sigma X$ .
- (a) Prove that the mean and variance of  $Y$  ( $\mu_Y$  and  $\sigma_Y^2$  respectively) may be written in terms of the mean and variance of  $X$  ( $\mu_X$  and  $\sigma_X^2$  respectively) as follows:

$$\mu_Y = \mu + \sigma\mu_X; \quad \sigma_Y^2 = \sigma^2\sigma_X^2.$$

- (b) Prove that the MGF of  $Y$  is given by

$$M_Y(t) = e^{\mu t} M_X(\sigma t).$$

- (c) Express the mean and variance of  $Y$  in terms of  $M_X(t)$  and its derivatives.
- (d) Suppose that  $M_X(t) = e^{t+6t^2}$ .
  - (i) Determine  $\mu_X$  and  $\sigma_X$  using the MGF.
  - (ii) Find the mean and variance of  $Y = \frac{1}{3}(X - 1)$ .

3. (a) For discrete variables  $Y$  and  $N$  prove that

$$E_{f_N}(N) = E_{f_Y}(E_{f_{N|Y}}(N|Y))$$

- (b) Consider a sequence of independently, identically distributed random variables  $Y_i, i = 1, 2, \dots$  such that

$$Y_i = \begin{cases} 1 & \text{with probability } \theta, 0 \leq \theta \leq 1; \\ 0 & \text{otherwise,} \end{cases}$$

and let  $N$  be the discrete random variable representing the number of trials until a one is observed, i.e.  $N = n$  if:  $Y_n = 1$  and  $Y_i = 0, i = 1, \dots, n-1$ .

- (i) Determine  $E_{f_{Y_1}}(Y_1)$ .
- (ii) Derive the probability mass function of  $N$  and state the name of the distribution of  $N$ .
- (iii) Using part (a) and letting  $Y = Y_1$ , prove that

$$E_{f_N}(N) = (1 + E_{f_N}(N))(1 - \theta) + \theta,$$

hence find  $E_{f_N}(N)$ .

4. The continuous random variables  $X$  and  $Y$  have joint probability density function given by

$$f_{X,Y}(x,y) = \begin{cases} \frac{x}{3} + cy & 0 < x < 1, 0 < y < 3; \\ 0 & \text{otherwise,} \end{cases}$$

for some constant  $c$ .

- (a) Determine  $c$ .
- (b) Find the marginal density functions,  $f_X(x)$  of  $X$  and  $f_Y(y)$  of  $Y$ .
- (c) Find the marginal distribution functions,  $F_X(x)$  of  $X$  and  $F_Y(y)$  of  $Y$ .
- (d) Find the joint distribution function:

$$F_{X,Y}(x,y) = \int_0^y \int_0^x f_{X,Y}(u,v) du dv, \quad 0 < x < 1, 0 < y < 3.$$

- (e) Show that  $X$  and  $Y$  are not independent.
- (f) Determine  $P(X > 1/2 | Y > 1)$ . What would this probability be if  $X$  and  $Y$  were independent with marginals determined in part (c)?

# DISCRETE DISTRIBUTIONS

	RANGE $\mathbb{X}$	PARAMETERS	MASS FUNCTION $f_X$	CDF $F_X$	$E_{f_X} [X]$	$Var_{f_X} [X]$	MGF $M_X$
$Bernoulli(\theta)$	$\{0, 1\}$	$\theta \in (0, 1)$	$\theta^x (1 - \theta)^{1-x}$		$\theta$	$\theta(1 - \theta)$	$1 - \theta + \theta e^t$
$Binomial(n, \theta)$	$\{0, 1, \dots, n\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{n}{x} \theta^x (1 - \theta)^{n-x}$		$n\theta$	$n\theta(1 - \theta)$	$(1 - \theta + \theta e^t)^n$
$Poisson(\lambda)$	$\{0, 1, 2, \dots\}$	$\lambda \in \mathbb{R}^+$	$\frac{e^{-\lambda} \lambda^x}{x!}$		$\lambda$	$\lambda$	$\exp \{ \lambda (e^t - 1) \}$
$Geometric(\theta)$	$\{1, 2, \dots\}$	$\theta \in (0, 1)$	$(1 - \theta)^{x-1} \theta$	$1 - (1 - \theta)^x$	$\frac{1}{\theta}$	$\frac{(1 - \theta)}{\theta^2}$	$\frac{\theta e^t}{1 - e^t(1 - \theta)}$
$NegBinomial(n, \theta)$	$\{n, n + 1, \dots\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{x-1}{n-1} \theta^n (1 - \theta)^{x-n}$		$\frac{n}{\theta}$	$\frac{n(1 - \theta)}{\theta^2}$	$\left( \frac{\theta e^t}{1 - e^t(1 - \theta)} \right)^n$
or	$\{0, 1, 2, \dots\}$	$n \in \mathbb{Z}^+, \theta \in (0, 1)$	$\binom{n+x-1}{x} \theta^n (1 - \theta)^x$		$\frac{n(1 - \theta)}{\theta}$	$\frac{n(1 - \theta)}{\theta^2}$	$\left( \frac{\theta}{1 - e^t(1 - \theta)} \right)^n$

For CONTINUOUS distributions (see over), define the **GAMMA FUNCTION**

$$\Gamma(\alpha) = \int_0^{\infty} x^{\alpha-1} e^{-x} dx$$

and the LOCATION/SCALE transformation  $Y = \mu + \sigma X$  gives

$$f_Y(y) = f_X \left( \frac{y - \mu}{\sigma} \right) \frac{1}{\sigma} \quad F_Y(y) = F_X \left( \frac{y - \mu}{\sigma} \right) \quad M_Y(t) = e^{\mu t} M_X(\sigma t) \quad E_{f_Y} [Y] = \mu + \sigma E_{f_X} [X] \quad Var_{f_Y} [Y] = \sigma^2 Var_{f_X} [X]$$



**CONTINUOUS DISTRIBUTIONS**

	$\mathbb{X}$	PARAMS.	PDF	CDF	$E_{f_X}[X]$	$Var_{f_X}[X]$	MGF
<i>Uniform</i> ( $\alpha, \beta$ ) (stand. model $\alpha = 0, \beta = 1$ )	$(\alpha, \beta)$	$\alpha < \beta \in \mathbb{R}$	$\frac{1}{\beta - \alpha}$	$\frac{x - \alpha}{\beta - \alpha}$	$\frac{(\alpha + \beta)}{2}$	$\frac{(\beta - \alpha)^2}{12}$	$M_X = \frac{e^{\beta t} - e^{\alpha t}}{t(\beta - \alpha)}$
<i>Exponential</i> ( $\lambda$ ) (stand. model $\lambda = 1$ )	$\mathbb{R}^+$	$\lambda \in \mathbb{R}^+$	$\lambda e^{-\lambda x}$	$1 - e^{-\lambda x}$	$\frac{1}{\lambda}$	$\frac{1}{\lambda^2}$	$\left(\frac{\lambda}{\lambda - t}\right)$
<i>Gamma</i> ( $\alpha, \beta$ ) (stand. model $\beta = 1$ )	$\mathbb{R}^+$	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\beta^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\beta x}$		$\frac{\alpha}{\beta}$	$\frac{\alpha}{\beta^2}$	$\left(\frac{\beta}{\beta - t}\right)^\alpha$
<i>Weibull</i> ( $\alpha, \beta$ ) (stand. model $\beta = 1$ )	$\mathbb{R}^+$	$\alpha, \beta \in \mathbb{R}^+$	$\alpha \beta x^{\alpha-1} e^{-\beta x^\alpha}$	$1 - e^{-\beta x^\alpha}$	$\frac{\Gamma(1 + 1/\alpha)}{\beta^{1/\alpha}}$	$\frac{\Gamma(1 + \frac{2}{\alpha}) - \Gamma(1 + \frac{1}{\alpha})^2}{\beta^{2/\alpha}}$	
<i>Normal</i> ( $\mu, \sigma^2$ ) (stand. model $\mu = 0, \sigma = 1$ )	$\mathbb{R}$	$\mu \in \mathbb{R}, \sigma \in \mathbb{R}^+$	$\frac{1}{\sqrt{2\pi}\sigma^2} \exp\left\{-\frac{(x - \mu)^2}{2\sigma^2}\right\}$		$\mu$	$\sigma^2$	$e^{\{\mu t + \sigma^2 t^2/2\}}$
<i>Student</i> ( $\nu$ )	$\mathbb{R}$	$\nu \in \mathbb{R}^+$	$\frac{(\pi\nu)^{-\frac{1}{2}} \Gamma\left(\frac{\nu+1}{2}\right)}{\Gamma\left(\frac{\nu}{2}\right) \left\{1 + \frac{x^2}{\nu}\right\}^{(\nu+1)/2}}$		0 (if $\nu > 1$ )	$\frac{\nu}{\nu - 2}$ (if $\nu > 2$ )	
<i>Pareto</i> ( $\theta, \alpha$ )	$\mathbb{R}^+$	$\theta, \alpha \in \mathbb{R}^+$	$\frac{\alpha \theta^\alpha}{(\theta + x)^{\alpha+1}}$	$1 - \left(\frac{\theta}{\theta + x}\right)^\alpha$	$\frac{\theta}{\alpha - 1}$ (if $\alpha > 1$ )	$\frac{\alpha \theta^2}{(\alpha - 1)(\alpha - 2)}$ (if $\alpha > 2$ )	
<i>Beta</i> ( $\alpha, \beta$ )	(0, 1)	$\alpha, \beta \in \mathbb{R}^+$	$\frac{\Gamma(\alpha + \beta)}{\Gamma(\alpha)\Gamma(\beta)} x^{\alpha-1} (1 - x)^{\beta-1}$		$\frac{\alpha}{\alpha + \beta}$	$\frac{\alpha\beta}{(\alpha + \beta)^2(\alpha + \beta + 1)}$	

This paper is also taken for the relevant examination for the Associateship of the Royal College of Science .

**M1S**

**Probability & Statistics I (Solutions)**

Setter's signature

.....

Checker's signature

.....

Editor's signature

.....

seen ↓

1. (a) **Axioms of Probability**For events  $E, F \subseteq \Omega$ 

(I)  $0 \leq P(E) \leq 1$

(II)  $P(\Omega) = 1$

(III) If  $E \cap F = \phi$ , then  $P(E \cup F) = P(E) + P(F)$  (Addition rule)

3

sim. seen ↓

(b)

$$E = (E \cap F) \cup (E \cap F')$$

$$\Rightarrow P(E) = P(E \cap F) + P(E \cap F') \text{ axiom III as } (E \cap F) \cap (E \cap F') = \phi$$

$$\Rightarrow P(E) \geq P(E \cap F) \text{ axiom I ( } P(E \cap F') \geq 0 \text{)}$$

3

unseen ↓

(c)

$$\begin{aligned}
 P(\text{only } E_2) &= P(E_2 \cap E_1' \cap E_3') = P(E_2)P(E_1' | E_2)P(E_3' | E_1' \cap E_2) \\
 &= P(E_2)P(E_1')P(E_3' | E_1') \text{ independence of } E_1, E_2 \text{ and } E_2, E_3 \\
 &= \frac{1}{4} \times \frac{1}{2} \left( \frac{P(E_3' \cap E_1')}{P(E_1')} \right) = \frac{1}{8} \left( \frac{P((E_3 \cup E_1)')}{1/2} \right) \\
 &= \frac{1}{4}(1 - P(E_3 \cup E_1)) = \frac{1}{4}(1 - P(E_3) - P(E_1)) \text{ as } E_3 \cap E_1 = \phi \\
 &= \frac{1}{4} \left( 1 - \frac{1}{8} - \frac{1}{2} \right) = \frac{3}{32}.
 \end{aligned}$$

Alternatively,

$$\begin{aligned}
 P(\text{only } E_2) &= P(E_2) - P(E_2 \cap E_1) - (E_2 \cap E_3) \\
 &= P(E_2) - P(E_2 | E_1)P(E_1) + P(E_2 | E_3)P(E_3) \\
 &= P(E_2) - P(E_2)P(E_1) + P(E_2)P(E_3) \text{ independence} \\
 &= \frac{1}{4} - \frac{1}{4} \times \frac{1}{2} + \frac{1}{4} \times \frac{1}{8} = \frac{3}{32}.
 \end{aligned}$$

4

meth seen ↓

(d)

$$E_{f_X}(X) = \sum_{x=0}^{\infty} x f_X(x) = f_X(1) + 2f_X(2) + 3f_X(3) + \dots$$

$$f_X(x) \geq 0 \forall x \Rightarrow E_{f_X}(X) = 0 \Leftrightarrow f_X(x) = 0 \forall x \geq 1.$$

$$\sum_{i=0}^{\infty} f_X(x) = 1 \Rightarrow f_X(0) = P(X = 0) = 1.$$

2

sim. seen ↓

(e) (i)  $Y = X + 2$ , so range of  $Y$  is  $\{0, 1, 2, 3, 4\}$

$$f_Y(y) = P(Y = y) = P(X = y - 2) = f_X(y - 2), y \in \{0, 1, 2, 3, 4\}$$

Giving

$$f_Y(0) = f_Y(1) = f_Y(2) = \frac{1}{8}, f_Y(3) = \frac{1}{4}, f_Y(4) = \frac{3}{8}.$$

2

(ii)  $Z = X^2$ , so range of  $Z$  is  $\{0, 1, 4\}$

$$f_Z(z) = P(Z = x) = P(X = \pm\sqrt{x}), P((X = -\sqrt{x}) \cup (X = \sqrt{x})), x \in \{0, 1, 4\}.$$

Giving

$$f_Z(0) = \frac{1}{8}, f_Z(1) = \frac{3}{8}, f_Z(4) = \frac{1}{2}.$$

2

unseen ↓

(f) Roots are real if  $\alpha^2 - 4\beta \geq 0$ .

$$P(\alpha^2 - 4\beta \geq 0) = P(\beta \leq \alpha^2/4)$$

$\alpha$  and  $\beta$  are both *Uniform* $[-1, 1]$  independently so

$$f_{\alpha, \beta}(x, y) = f_{\alpha}(x)f_{\beta}(y) = \frac{1}{4}, -1 \leq x, y \leq 1$$

$$\begin{aligned} P(\beta \leq \alpha^2/4) &= \int_{-1}^1 \int_{-1}^{\alpha^2/4} \frac{1}{4} d\beta d\alpha \\ &= \frac{1}{4} \int_{-1}^1 [\beta]_{-1}^{\alpha^2/4} d\alpha = \frac{1}{4} \int_{-1}^1 \left( \frac{\alpha^2}{4} + 1 \right) d\alpha \\ &= \frac{1}{4} \left[ \frac{\alpha^3}{12} + \alpha \right]_{-1}^1 = \frac{13}{24}. \end{aligned}$$

4



2. (a)

$$\mu_Y = E_{f_Y}(Y) = E_{f_X}(\mu + \sigma X) = \mu + \sigma E_{f_X}(X) = \mu + \sigma \mu_X,$$

2

$$\begin{aligned}\sigma_Y^2 &= \text{var}_{f_Y}(Y) = \text{var}_{f_X}(\mu + \sigma X) \\ &= E_{f_X}((\mu + \sigma X)^2) - E_{f_X}^2(\mu + \sigma X) \\ &= \mu^2 + 2\mu\sigma E_{f_X}(X) + \sigma^2 E_{f_X}(X^2) - (\mu + \sigma E_{f_X}(X))^2 \\ &= \mu^2 + 2\mu\sigma E_{f_X}(X) + \sigma^2 E_{f_X}(X^2) - (\mu^2 + 2\mu\sigma E_{f_X}(X) + \sigma^2 E_{f_X}^2(X)) \\ &= \sigma^2(E_{f_X}(X^2) - E_{f_X}^2(X)) = \sigma^2 \sigma_X^2.\end{aligned}$$

Note: Can also assume that  $\text{var}(aX + b) = a^2 \text{var}(X)$  which gives much shorter solution.

3

(b)

$$\begin{aligned}M_Y(t) &= E_{f_Y}(e^{tY}) = E_{f_X}(e^{t(\mu + \sigma X)}) \\ &= e^{\mu t} E_{f_X}(e^{t\sigma X}) = e^{\mu t} E_{f_X}(e^{t\sigma X}) = e^{\mu t} M_X(\sigma t)\end{aligned}$$

3

(c) We have  $\mu_X = M'_X(0)$  and  $\sigma_X^2 = E_{f_X}(X^2) - E_{f_X}^2(X) = M''_X(0) - (M'_X(0))^2$ , so, from (a)

$$\begin{aligned}\mu_Y &= \mu + \sigma M'_X(0) \\ \sigma_Y^2 &= \sigma^2(M''_X(0) - (M'_X(0))^2).\end{aligned}$$

Alternatively,

$$\begin{aligned}M'_Y(t) &= \mu e^{\mu t} M_X(\sigma t) + \sigma e^{\mu t} M'_X(\sigma t) \\ \Rightarrow \mu_Y &= M'_Y(0) = \mu M_X(0) + \sigma M'_X(0) = \mu + \sigma M'_X(0),\end{aligned}$$

and,

$$\begin{aligned}M''_Y(t) &= \mu^2 e^{\mu t} M_X(\sigma t) + \mu \sigma e^{\mu t} M'_X(\sigma t) + \sigma \mu e^{\mu t} M'_X(\sigma t) + \sigma^2 e^{\mu t} M''_X(\sigma t) \\ \Rightarrow M''_Y(0) &= \mu^2 M_X(0) + 2\mu\sigma M'_X(0) + \sigma^2 M''_X(0) \\ &= \mu^2 + 2\mu\sigma M'_X(0) + \sigma^2 M''_X(0) \\ \Rightarrow \sigma_Y^2 &= E_{f_Y}(Y^2) - E_{f_Y}^2(Y) \\ &= M''_Y(0) - (M'_Y(0))^2 \\ &= \mu^2 + 2\mu\sigma M'_X(0) + \sigma^2 M''_X(0) - (\mu + \sigma M'_X(0))^2 \\ &= \sigma^2(M''_X(0) - [M'_X(0)]^2).\end{aligned}$$

4

unseen ↓

(d)  $M_X(t) = e^{t+6t^2}$ , so  $X \sim N(1, 12)$ .

(i)

$$M'_X(t) = (1 + 12t)e^{t+6t^2} \Rightarrow \mu_X = M'_X(0) = 1.$$

2

$$M''_X(t) = (1 + 12t)^2 e^{t+6t^2} + 12e^{t+6t^2}$$

$$\sigma_X^2 = M''_X(0) - (M'_X(0))^2$$

$$= 1 + 12 - 1 = 12.$$

$$\Rightarrow \sigma_X = \sqrt{12} = 2\sqrt{3}.$$

3

as expected.

(ii)  $Y = (X - 1)/3 \Rightarrow \mu = 1/3, \sigma = -1/3,$

$$\mu_Y = \mu + \sigma\mu_X = \frac{1}{3} - \frac{1}{3} = 0.$$

$$\sigma_Y^2 = \sigma^2\sigma_X^2 = \frac{1}{9} \times 12 = \frac{4}{3}.$$

3

3. (a)

$$\begin{aligned}
 E_{f_Y}(E_{f_{N|Y}}(N | Y = y)) &= \sum_y E_{f_{N|Y}}(N | Y = y) f_Y(y) \\
 &= \sum_y \sum_n n f_{N|Y}(n | Y = y) f_Y(y) \\
 &= \sum_n n \sum_y f_{N,Y}(n, y) = \sum_n n f_N(n) \\
 &= E_{f_N}(N).
 \end{aligned}$$

4

(b) (i)

$$\begin{aligned}
 E_{f_{Y_1}}(Y_1) &= \sum_{y=0}^1 y f_{Y_1}(y) = 0 \times f_{Y_1}(0) + 1 \times f_{Y_1}(1) \\
 &= f_{Y_1}(1) = \theta.
 \end{aligned}$$

2

(ii)

$$\begin{aligned}
 f_N(n) &= P(N = n) = P(Y_1 = 0 \cap \dots \cap Y_{n-1} = 0 \cap Y_n = 1) \\
 &= (1 - \theta)^{n-1} \theta \quad n = 1, 2, \dots \quad \text{as } Y_i\text{s independent,}
 \end{aligned}$$

3

i.e.  $N \sim \text{Geometric}(\theta)$ 

1

unseen ↓

(iii) from (a)

$$E_{f_N}(N) = E_{f_{Y_1}}(E_{f_{N|Y_1}}(N | Y_1 = y))$$

2

$$\begin{aligned} E_{f_N}(N) &= \sum_{y=0}^1 E_{f_{N|Y_1}}(N | Y_1 = y) f_{Y_1}(y) \\ &= E_{f_{N|Y_1}}(N | Y_1 = 0)(1 - \theta) + E_{f_{N|Y_1}}(N | Y_1 = 1)(\theta). \end{aligned}$$

$$\begin{aligned} E_{f_{N|Y_1}}(N | Y_1 = 0) &= \sum_{n=1}^{\infty} n f_{N|Y_1}(n | Y_1 = 0) \\ &= \sum_{n=2}^{\infty} n \frac{P(N = n \cap Y_1 = 0)}{P(Y_1 = 0)} = \sum_{n=2}^{\infty} n \frac{(1 - \theta)^{n-1} \theta}{(1 - \theta)} \\ &= \sum_{n=2}^{\infty} n (1 - \theta)^{n-2} \theta. \end{aligned}$$

Let  $M = N + 1$ , then  $E_{f_M}(M) = E_{f_N}(N) + 1$  and

$$f_M(m) = (1 - \theta)^{m-2} \theta, \quad m = 2, 3, \dots$$

So,

$$E_{f_{N|Y_1}}(N | Y_1 = 0) = E_{f_M}(M) = E_{f_N}(N) + 1.$$

Also.  $f_{N|Y_1}(n | Y_1 = 1) = 1, n = 1$ , and  $f_{N|Y_1}(n | Y_1 = 1) = 0, n > 1$ , so  $E_{f_{N|Y_1}}(N | Y_1 = 1) = 1$ .

Giving,

$$E_{f_N}(N) = (1 + E_{f_N}(N))(1 - \theta) + \theta,$$

as required.

Could also use argument (if well explained) that if  $Y_1 = 0$ , the game starts over.

Hence,

5

$$\begin{aligned} E_{f_N}(N) &= (1 + E_{f_N}(N))(1 - \theta) + \theta \\ \Rightarrow E_{f_N}(N) (1 - (1 - \theta)) &= (1 - \theta) + \theta \\ \Rightarrow E_{f_N}(N) &= \frac{1}{\theta}, \end{aligned}$$

as expected given  $N \sim \text{Geometric}(\theta)$ .

3

4. (a)

$$\begin{aligned}
 \int_0^1 \int_0^3 \left( \frac{x}{3} + cy \right) dy dx &= \int_0^1 \left[ \frac{xy}{3} + \frac{cy^2}{2} \right]_0^3 dx \\
 &= \int_0^1 \left( x + \frac{9c}{2} \right) dx = \left[ \frac{x^2}{2} + \frac{9cx}{2} \right]_0^1 \\
 &= \frac{1}{2} + \frac{9c}{2}, \\
 \Rightarrow \frac{1}{2} + \frac{9c}{2} &= 1 \Rightarrow c = \frac{1}{9}.
 \end{aligned}$$

2

(b)

$$\begin{aligned}
 f_X(x) &= \frac{1}{9} \int_0^3 3x + y dy = \frac{1}{9} \left[ 3xy + \frac{y^2}{2} \right]_0^3 \\
 &= x + \frac{1}{2}, \quad 0 < x < 1.
 \end{aligned}$$

2

$$\begin{aligned}
 f_Y(y) &= \frac{1}{9} \int_0^1 3x + y dx = \frac{1}{9} \left[ \frac{3x^2}{2} + yx \right]_0^1 \\
 &= \frac{1}{9} \left( y + \frac{3}{2} \right), \quad 0 < y < 3.
 \end{aligned}$$

2

(c)  $F_X(x) = 0, x \leq 0; F_X(x) = 1, x \geq 1;$ 

$$\begin{aligned}
 F_X(x) &= \int_0^x f_X(v) dv = \int_0^x v + \frac{1}{2} dv \\
 &= \left[ \frac{v^2}{2} + \frac{v}{2} \right]_0^x = \frac{x^2}{2} + \frac{x}{2} = \frac{x}{2}(x+1), \quad 0 < x < 1.
 \end{aligned}$$

2

 $F_Y(y) = 0, y \leq 0; F_Y(y) = 1, y \geq 3;$ 

$$\begin{aligned}
 F_Y(y) &= \int_0^y f_Y(v) dv = \frac{1}{9} \int_0^y \left( v + \frac{3}{2} \right) dv \\
 &= \frac{1}{9} \left[ \frac{v^2}{2} + \frac{3v}{2} \right]_0^y = \frac{1}{9} \left( \frac{y^2}{2} + \frac{3y}{2} \right) = \frac{y}{18}(y+3), \quad 0 < y < 3.
 \end{aligned}$$

2



(d)

$$\begin{aligned}F_{X,Y}(x,y) &= \frac{1}{9} \int_0^x \int_0^y 3u + v \, dv \, du \\&= \frac{1}{9} \int_0^x \left[ 3uv + \frac{v^2}{2} \right]_0^y \, du \\&= \frac{1}{9} \int_0^x 3uy + \frac{y^2}{2} \, du = \frac{1}{9} \left[ \frac{3u^2 y}{2} + \frac{y^2 u}{2} \right]_0^x \\&= \frac{1}{9} \left( \frac{3x^2 y}{2} + \frac{xy^2}{2} \right) = \frac{xy}{18} (3x + y), \quad 0 < x < 1; 0 < y < 1.\end{aligned}$$

3

(e)

$$f_X(x)f_Y(y) = \frac{1}{9} \left( x + \frac{1}{2} \right) \left( y + \frac{3}{2} \right) \neq f_{X,Y}(x,y).$$

NOT independent.

Alternatively,

$$F_X(x)F_Y(y) = \frac{x}{2}(x+1) \frac{y}{18}(y+3) \neq F_{X,Y}(x,y).$$

1

unseen ↓

(f)

$$\begin{aligned}P(X > 0.5 \mid Y > 1) &= \frac{P(X > 0.5 \cap Y > 1)}{P(Y > 1)} \\P(X > 0.5 \cap Y > 1) &= 1 - (F_X(0.5) + F_Y(1) - F_{X,Y}(0.5, 1)) \\&= 1 - \left( \frac{1}{4} \left( \frac{1}{2} + 1 \right) + \frac{1}{18}(1+3) - \frac{1}{36} \left( \frac{3}{2} + 1 \right) \right) \\&= 1 - \left( \frac{3}{8} + \frac{2}{9} - \frac{5}{72} \right) = 1 - \left( \frac{27+16-5}{72} \right) \\&= 1 - \frac{19}{36} = \frac{17}{36} \\ \Rightarrow P(X > 0.5 \mid Y > 1) &= \frac{17/36}{1-2/9} = \frac{17/36}{7/9} = \frac{17}{28}.\end{aligned}$$

Alternatively,

$$\begin{aligned}P(X > 0.5 \cap Y > 1) &= \frac{1}{9} \int_{0.5}^1 \int_1^3 3x + y \, dy \, dx \\&= \frac{1}{9} \int_{0.5}^1 \left[ 3xy + \frac{y^2}{2} \right]_1^3 \, dx = \frac{1}{9} \int_{0.5}^1 \left( 9x + \frac{9}{2} \right) - \left( 3x + \frac{1}{2} \right) \, dx \\&= \frac{1}{9} \int_{0.5}^1 6x + 4 \, dx = \frac{1}{9} [3x^2 + 4x]_{0.5}^1 \\&= \frac{1}{9} \left( (3+4) - \left( \frac{3}{4} + 2 \right) \right) = \frac{1}{9} \left( 7 - \frac{11}{4} \right) = \frac{17}{36}.\end{aligned}$$

5

If independent, then

$$P(X > 0.5 \mid Y > 1) = P(X > 0.5) = 1 - F_X(0.5) = 1 - \frac{3}{8} = \frac{5}{8}.$$

1