

1 (a) Axioms

- (i) For all events $A \in \mathcal{A}$, $P(A) \geq 0$
- (ii) $P(\Omega) = 1$
- (iii) For any sequence of disjoint events $A_i, i=1, \dots$

$$P\left(\bigcup_{i=1}^{\infty} A_i\right) = \sum_{i=1}^{\infty} P(A_i)$$

I will accept

- (iiia) For any two disjoint events A & B .

$$P(A \cup B) = P(A) + P(B)$$

or (iiib) For any set A_1, \dots, A_n of disjoint events

$$P\left(\bigcup_{i=1}^n A_i\right) = \sum_{i=1}^n P(A_i) \quad [3 \text{ marks}]$$

(b) (i) \mathcal{A} has $2^4 = 16$ elements [1 mark]

(ii) By Axiom 3, we know that

$$P(\{A\}) + P(\{B\}) = P(\{A, B\}) \Rightarrow \underline{P(\{B\}) = 1/3}$$

$$P(\{A\}) + P(\{C\}) = P(\{A, C\}) \Rightarrow \underline{P(\{C\}) = 1/3}$$

Then

$$\underline{P(\{A, B, C\})} = P(\{A\}) + P(\{B\}) + P(\{C\}) = \underline{1/2}$$

$$P(\{A, B, C\}) + P(\{D\}) = P(\Omega) = 1 \Rightarrow \underline{P(\{D\}) = 1/2}$$

We then derive the probabilities of the other elements of \mathcal{A} .

$$P(\emptyset) = 0, \quad P(\{A, D\}) = 1/3, \quad P(\{B, C\}) = 2/3, \quad P(\{B, D\}) =$$

$$P(\{C, D\}) = 1/2, \quad P(\{A, B, D\}) = 7/12, \quad P(\{A, C, D\}) = 3/4,$$

$$P(\{B, C, D\}) = 3/4, \quad P(\Omega) = 1 \quad [5 \text{ marks}]$$

2(a)

$$P(B|A) = \frac{P(A|B) P(B)}{P(A)}.$$

[2 marks]

(b). (i) Let D be the event that the person has the disease and C be the event that they test positive. Then the information that we are given can be written as:

$$\begin{aligned} P(C|D) &= 0.95 & P(C|D^c) &= 0.05 \\ P(D) &= 0.002 \end{aligned}$$

Using Bayes' Formula.

$$P(D|C) = \frac{P(C|D) P(D)}{P(C|D) P(D) + P(C|D^c) P(D^c)}$$

$$= \frac{0.95 \times 0.002}{0.95 \times 0.002 + 0.05 \times 0.998}$$

$$= 0.0367.$$

[3 marks]

(ii). The test is not likely to be useful because the probability that a person has the disease when they test positive is still very low.
- There is a high rate of false positives.
[1 mark]

3 (a). X has a hypergeometric distribution with parameters $N = 45$, $D = 7$ and $n = 7$.

[2 marks]

(b). For $x = 0, \dots, 7$

$$p_X(x) = \frac{\binom{7}{x} \binom{38}{7-x}}{\binom{45}{7}}$$

[2 marks]

(c) M is a binomial random variable with parameters $n = 80,000,000$ and $p = \frac{\binom{7}{7} \binom{38}{0}}{\binom{45}{7}}$

$$= \frac{1}{45,379,620} \quad [2 \text{ marks}]$$

(d) $P(M > 0) = 1 - P(M = 0)$

$$= 1 - (1-p)^n$$

$$= 1 - \left(\frac{45,379,619}{45,379,620} \right)^{80,000,000} \quad [2 \text{ marks}]$$

(e) M is approximately distributed as a Poisson random variable with parameter $np =$

$$\frac{80,000,000}{45,379,620} = 1.7629$$

So $P(M > 0) = 1 - P(M = 0)$

$$= 1 - e^{-1.7629}$$

$$= 0.8285 \quad [2 \text{ marks}]$$

(f) N is the number of 'failures' before the first success in a sequence of independent Bernoulli Trials, with success probability $p = 0.8285$. So N has a geometric distribution with parameter 0.8285 . [2 marks]

(g) $E[N] = \frac{(1-p)}{p} = 0.2070$

$$V[N] = \frac{(1-p)}{p^2} = 0.2498$$

[2 marks]

4 (a) $E(X) = 1/2$
 $V(X) = 1/12$

[2 marks]

(b) X can range between 0 and 1 and.
 $Y = e^X$

So Y ranges between 1 and e .

That is $S_Y = [1, e]$. [1 mark]

(c) For $y \in [1, e]$, $F_Y(y) = P(Y \leq y)$.

$$= P(e^X \leq y)$$

$$= P(X \leq \log y)$$

$$= \log y$$

So $F_Y(y) = \begin{cases} 0 & , y \in (-\infty, 1) \\ \log y & , y \in [1, e] \\ 1 & , y \in (e, \infty) \end{cases}$

$f_Y(y) = \begin{cases} \frac{1}{y} & , y \in [1, e] \\ 0 & \text{otherwise} \end{cases}$

[3 marks]

(d). (i) $E(Y) = \int_0^1 e^x dx = [e^x]_0^1 = e - 1$.

(ii) $E(Y) = \int_1^e y \frac{1}{y} dy = [y]_1^e = e - 1$.

(e) $E(Y^2) = \int_1^e y^2 \frac{1}{y} dy = \left[\frac{y^2}{2} \right]_1^e = \frac{1}{2}(e^2 - 1)$ [2 marks]

$\therefore V(Y) = \frac{1}{2}(e^2 - 1) - (e - 1)^2 = -\frac{1}{2}e^2 + 2e - \frac{3}{2}$. [2 marks]

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$$(f) \quad Y \approx e^{1/2} + (X - 1/2)e^{1/2} + (X - 1/2)^2 e^{1/2} / 2$$

$$\text{So } E[Y] \approx e^{1/2} + 1/2 V(X) e^{1/2}$$

$$= e^{1/2} \frac{25}{24} = 1.7174$$

$$\text{cf } e^{-1} = 1.7183$$

The variance is small, so we expect the approximation to the mean to be good.

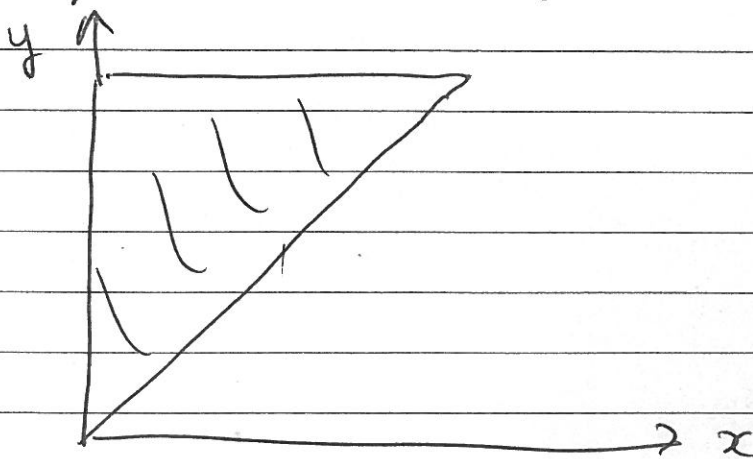
$$V(Y) \approx V(X) (e^{1/2})^2$$

$$= \frac{1}{12} \cdot e = 0.2265$$

$$\text{cf } -\frac{e^2}{2} + 2e^{-3/2} = 0.2420$$

The approximation is reasonably close. [4 marks]

5 (a) (i) X and Y are clearly dependent
since $f_{(X,Y)}(3/4, 1/4) = 0$ but $f_X(3/4) > 0$ and $f_Y(1/4) > 0$



[2 marks]

$$(ii). \quad 1 = \int_0^1 \int_0^y hxy \, dx \, dy$$

$$= \int_0^1 \left[\frac{hx^2y}{2} \right]_0^y dy$$

$$= \int_0^1 \frac{ky^3}{2} dy$$

$$= \left[\frac{ky^4}{8} \right]_0^1 = \frac{k}{8}$$

$$\therefore k = 8$$

[2 marks]

(iii) For $x \in (0,1)$

$$\begin{aligned} f_x(x) &= \int_x^1 8xy \, dy \\ &= [4xy^2]_x^1 \\ &= 4x - 4x^3 \end{aligned}$$

For $y \in (0,1)$

$$\begin{aligned} f_y(y) &= \int_0^y 8xy \, dx \\ &= [4x^2y]_0^y \\ &= 4y^3 \end{aligned}$$

[2 marks]

$$\begin{aligned} \text{(iv)} \quad f_{x|y}(y|x=1/2) &= \frac{8xy}{4x-4x^3} \Big|_{x=1/2} \\ &= \frac{4y}{3/2} \\ &= \frac{8y}{3} \end{aligned}$$

[2 marks]

$$\text{(v)} \quad P(Y \leq 3/4 \mid X \leq 1/2)$$

$$= \frac{P(Y \leq 3/4 \text{ and } X \leq 1/2)}{P(X \leq 1/2)}$$

$$= \frac{\int_0^{1/2} \int_x^{3/4} 8xy \, dy \, dx}{\int_0^{1/2} \int_x^1 8xy \, dy \, dx}$$

$$= \frac{\int_0^{1/2} [4xy^2]_x^{3/4} \, dx}{\int_0^{1/2} [4xy^2]_x^1 \, dx}$$

$$= \frac{\int_0^{1/2} \frac{9x}{4} - 4x^3 dx}{\int_0^{1/2} 4x - 4x^3 dx.}$$

$$= \frac{\left[\frac{9x^2}{8} - x^4 \right]_0^{1/2}}{\left[2x^2 - x^4 \right]_0^{1/2}}$$

$$= \frac{\frac{9}{32} - \frac{1}{16}}{\frac{1}{2} - \frac{1}{16}}$$

$$= \frac{7/32}{7/16}$$

$$= \frac{1}{2}$$

[3 marks]

(b). (i) $F_Z(z) = P(Z \leq z)$

$$= \int_{S_X} P(Z \leq z | X=x) f_X(x) dx.$$

$$= \int_{S_X} P(Y \leq z-x | X=x) f_X(x) dx.$$

$$= \int_{S_X} P(Y \leq z-x) f_X(x) dx \quad (\text{by independence})$$

$$= \int_{S_X} F_Y(z-x) f_X(x) dx.$$

[3 marks]

(ii).

$$F_Z(z) = \int_0^\infty (1 - e^{-\alpha(z-x)}) I(z-x > 0) \alpha e^{-\alpha x} dx$$

$$= \int_0^z \alpha e^{-\alpha x} - \alpha e^{-\alpha z} dx$$

$$= \left[-e^{-\alpha x} - \alpha e^{-\alpha z} x \right]_0^z$$

$$= 1 - e^{-\alpha z} - \alpha z e^{-\alpha z}$$

[2 marks]

6/ (a) $E[Z|Y=1] = E[X] = 1/2$
 $E[Z|Y=2] = E[X^2] = \int_0^1 x^2 dx = \left[\frac{x^3}{3} \right]_0^1 = 1/3$ [2 marks]

(b) $V[Z|Y=1] = V[X] = 1/12$
 $V[Z|Y=2] = V[X^2] = E[X^4] - E[X^2]^2$
 $= \int_0^1 x^4 dx - 1/9$
 $= \left[\frac{x^5}{5} \right]_0^1 - 1/9$
 $= \frac{4}{45}$ [2 marks]

(c) $E[Z|Y] = \begin{cases} 1/2 & \text{with probability } 1/2 \\ 1/3 & \text{" } 1/2 \end{cases}$
 $V[Z|Y] = \begin{cases} 1/12 & \text{" } 1/2 \\ 4/45 & \text{" } 1/2 \end{cases}$ [2 marks]

(d) $E[Z] = E[E[Z|Y]] = 1/2 (1/2 + 1/3) = \frac{5}{12}$ [2 marks]

(e) $V[Z] = E[V[Z|Y]] + V[E[Z|Y]] =$
 $= 1/2 \left[\frac{1}{12} + \frac{4}{45} \right] + 1/2 \left[\frac{1}{2} - \frac{5}{12} \right]^2 + 1/2 \left[\frac{1}{3} - \frac{5}{12} \right]^2$
 $= 1/2 \left[\frac{15+16}{180} \right] + 1/2 \left[\frac{1}{144} \right] + 1/2 \left[\frac{1}{144} \right]$
 $= \frac{31}{360} + \frac{1}{144}$
 $= \frac{248+20}{2880}$
 $= \frac{268}{2880}$
 $= \frac{67}{720}$ [2 marks]

7/ (a) $f_R(r) = re^{-r^2/2} \quad r > 0$
 $f_\theta(\theta) = \frac{1}{2\pi} \quad 0 < \theta < 2\pi$ [2 marks]

(b) $F_R(r) = \int_0^r ue^{-u^2/2} du$
 $= -e^{-u^2/2} \Big|_0^r = 1 - e^{-r^2/2} \quad r > 0$

$F_\theta(\theta) = \frac{\theta}{2\pi} \quad 0 < \theta < 2\pi$ [2 marks]

(c) Since $f_{(R,\theta)}(r,\theta) = f_R(r) f_\theta(\theta)$, we conclude that R and θ are independent. [1 mark]

(d) (i) To get a random variable with distribution $F_R(r)$, we transform U_1 according to F_R^{-1} .
So we put $R = \sqrt{-2 \log(1-U)}$ [2 marks]

(ii) We transform U_2 according to F_θ^{-1} .
So we put $\theta = 2\pi U$. [1 mark]

(e) The random variable (X,Y) is bivariate standard normal with $\rho=0$. [1 mark]

8/ (a) (i)
$$M_2(t) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-zt} e^{-\frac{z^2}{2}} dz$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z-t)^2} e^{\frac{t^2}{2}} dz$$

$$= e^{t^2/2} \underbrace{\frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-\frac{1}{2}(z-t)^2} dz}_{=1}$$

$$= e^{t^2/2}$$
 [3 marks]

(ii) If $X = \mu + \sigma Z$

$$M_X(t) = e^{\mu t} + M_2(\sigma t)$$

$$= e^{\frac{\sigma^2 t^2}{2} + \mu t}$$
 [2 marks]

(iii) If $S_n = \sum_{i=1}^n X_i$, then

$$M_{S_n}(t) = \prod_{i=1}^n M_{X_i}(t)$$

$$= \prod_{i=1}^n e^{\frac{\sigma_i^2 t^2}{2} + \mu_i t}$$

$$= \exp \left(\sum_{i=1}^n \frac{\sigma_i^2 t^2}{2} + \sum_{i=1}^n \mu_i t \right)$$

which we recognise as the moment generating function of a normal distribution with mean $\sum_{i=1}^n \mu_i$ and variance $\sum_{i=1}^n \sigma_i^2$.

$$\text{So } S_n \stackrel{d}{=} \mathcal{N} \left(\sum_{i=1}^n \mu_i, \sum_{i=1}^n \sigma_i^2 \right). \quad [4 \text{ marks}]$$

$$\begin{aligned} (d) \quad M_{A_n}(t) &= M_{\frac{S_n}{n}}(t) = M_{S_n}\left(\frac{t}{n}\right) \\ &= \exp \left(\sum_{i=1}^n \frac{\sigma_i^2 t^2}{2n^2} + \sum_{i=1}^n \mu_i \frac{t}{n} \right) \end{aligned}$$

which we recognise as the moment generating function of a normal distribution with mean $\frac{1}{n} \sum_{i=1}^n \mu_i$ and variance $\frac{1}{n^2} \sum_{i=1}^n \sigma_i^2$ [2 marks].

$$(e) \quad M_{A_n} = \exp \left(\frac{\sigma^2 t^2}{n} + \mu t \right)$$

which is the moment generating function of a normal distribution with mean μ and variance $\frac{\sigma^2}{n}$. [2 marks]

$$9/ (a) \quad q_0 = 0$$

$$q_1 = A(q_0)$$

$$= \frac{2}{5}$$

$$\begin{aligned} q_2 &= A(q_1) = \frac{1}{10} \left(\frac{8}{125} + \frac{12}{25} + \frac{4}{5} + 4 \right) \\ &= \frac{334}{625} \end{aligned}$$

[3 marks]

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(b) To get the extinction probability we solve

$$q = A(q)$$

$$\Rightarrow q = \frac{1}{10} (q^3 + 3q^2 + 2q + 4).$$

$$\Rightarrow 0 = \frac{1}{10} (q^3 + 3q^2 - 8q + 4)$$

$$\Rightarrow 0 = (q-1)(q^2 + 4q - 4)$$

$$\Rightarrow q = 1, \quad \frac{-4 \pm \sqrt{16 + 16}}{2}$$

$$\Rightarrow q = 1, \quad -2 \pm 2\sqrt{2}.$$

We observe that $-2 + 2\sqrt{2} \in (0, 1)$. This is the minimal nonnegative solution, so the extinction probability is $-2 + 2\sqrt{2}$.