

# MATH-232: Midterm 2022 - Solutions B

May 15, 2022

**Exercise 1.** At the end of a video game, the player must kill a monster sampled randomly as follows:

- 4 times out of 10, it is a *dragon*,
- 1 time out of 10, it is a *troll*,
- the rest of the time, it is a *giant*.

When the monster dies, the player gets a chance to get a *ruby*:

- the dragon gives a ruby 1 time out of 3,
- the troll gives a ruby 1 time out of 6,
- the giant always gives a ruby.

We assume that the game is very easy; thus the players always succeed at killing the monster. Two different games are independent.

Alice plays the game. We denote  $D$  (respectively  $T$ ,  $G$ ) the event “the monster is a dragon” (respectively a troll, a giant). We denote  $R$  the event “Alice wins a ruby”, and  $p = \mathbb{P}(R)$ .

1. Compute  $p$ .
2. Alice won a ruby! What is the probability that the monster was a troll?
3. Bob decides to play 8 games. We denote  $S$  the number of rubies that he gets.
  - (a) What is the law of  $S$ ?
  - (b) What is the expectation of  $S$ ?
  - (c) What is the probability that Bob wins exactly 3 rubies?
  - (d) What is the probability that Bob wins at least 1 ruby?
4. Charlie decides to play until he wins one ruby. We denote  $X$  the number of games Charlie plays.
  - (a) What is the law of  $X$ ?
  - (b) What is the expectation of  $X$ ?
  - (c) What is the probability that Charlie does exactly 5 games?

*Solution.* (8 points)

1. (1.5pt) Using the law of total probability we write,

$$\begin{aligned} p = \mathbb{P}(R) &= \mathbb{P}(R|D)\mathbb{P}(D) + \mathbb{P}(R|T)\mathbb{P}(T) + \mathbb{P}(R|G)\mathbb{P}(G) \\ &= \frac{1}{3} \cdot \frac{4}{10} + \frac{1}{6} \cdot \frac{1}{10} + 1 \cdot \frac{5}{10} \\ &= \frac{1}{10} \cdot \frac{8 + 1 + 30}{6} \\ &= \frac{39}{60} = \frac{13}{20}. \end{aligned}$$

2. (1pt) Using Bayes rule,

$$\mathbb{P}(T|R) = \frac{\mathbb{P}(R|T)\mathbb{P}(T)}{p} = \frac{1}{6} \cdot \frac{1}{10} \cdot \frac{60}{39} = \frac{1}{39}.$$

3. (3pt)

a) (1pt) Since  $S$  is the sum of 8 independent events, with probability of success  $p$ , it follows that  $S \sim \text{Binomial}(p, 8)$

b) (0.5pt)  $\mathbb{E}[S] = 8p$

c) (1pt)  $\mathbb{P}(S = 3) = \binom{8}{3}p^3(1-p)^5$

d) (0.5pt)  $\mathbb{P}(S \geq 1) = 1 - \mathbb{P}(S = 0) = 1 - (1-p)^8$

4. (2.5pt)

a) (1pt) Since  $X$  denotes the number of trials to get the first success of independent events of probability  $p$ ,  $X \sim \text{Geom}(p)$

b) (0.5pt)  $\mathbb{E}[X] = 1/p = 60/39$

c) (1pt)  $\mathbb{P}(X = 5) = (1-p)^4p$ .

□

**Exercise 2.** Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be the function defined by the formula

$$f(x) = \begin{cases} 0 & \text{if } x \geq 0, \\ -9x \exp(3x) & \text{if } x < 0. \end{cases}$$

1. Justify that  $f$  is a probability density.

2. What is the expectation of a random variable  $X$  with density  $f$ ?

*Solution.* (2.5 points)

1. (1.5pt)  $f(x) \geq 0$  for all  $x$ . One can check by integration by part that  $\int_{-\infty}^{\infty} f(x)dx = 1$ .

2. (1pt)

$$\mathbb{E}[X] = \int_{-\infty}^0 -9x^2 \exp(3x)dx \tag{1}$$

$$= -3x^2 \exp(3x) \Big|_{-\infty}^0 + \int_{-\infty}^0 6x \exp(3x)dx \tag{2}$$

$$= -\frac{2}{3} \cdot \int_{-\infty}^0 -9x \exp(3x)dx \tag{3}$$

$$= -\frac{2}{3} \tag{4}$$

□

**Exercise 3.** Let  $(X, Y)$  be a joint random variable whose probability mass function is given by the following table:

		Y	
		-1	1
X	0	0.6	0.2
	1	0.1	0.1

1. What are the marginal laws of  $X$  and  $Y$ ?
2. Are  $X$  and  $Y$  independent? (As always, justify your answer.)
3. Compute  $\mathbb{E}[X]$ ,  $\mathbb{E}[Y]$ ,  $\mathbb{E}[X + Y]$  and  $\mathbb{E}[XY]$ .

*Solution:* (4 points)

1. (1pt)

$$f_Y(y) = \begin{cases} 0.7 & \text{if } y = -1, \\ 0.3 & \text{if } y = 1 \end{cases} \quad (5)$$

$$f_X(x) = \begin{cases} 0.8 & \text{if } x = 0, \\ 0.2 & \text{if } x = 1 \end{cases} \quad (6)$$

2. (0.5pt) No. For instance,

$$f_{X,Y}(0, -1) = 0.6 \quad (7)$$

$$f_X(0) \cdot f_Y(-1) = 0.8 \cdot 0.7 = 0.56. \quad (8)$$

Since,  $f_{X,Y}(0, -1) \neq f_X(0) \cdot f_Y(-1)$ , it follows that  $X$  and  $Y$  are not independent.

3. (2.5pt)

$$\mathbb{E}[X] = 0.2 \cdot 1 = 0.2 \quad (9)$$

$$\mathbb{E}[Y] = 0.3 \cdot 1 + 0.7 \cdot -1 = -0.4 \quad (10)$$

$$\mathbb{E}[X + Y] = \mathbb{E}[X] + \mathbb{E}[Y] = -0.2 \quad (11)$$

$$\mathbb{E}[X \cdot Y] = -1 \cdot 0.1 + 1 \cdot 0.1 = 0. \quad (12)$$

□

**Exercise 4.** Let  $F : \mathbb{R} \rightarrow [0, 1]$  be the function defined by

$$F(x) = \begin{cases} 0 & \text{if } x < 0, \\ \frac{x^2}{4} & \text{if } 0 \leq x \leq 2, \\ 1 & \text{if } x > 2. \end{cases}$$

1. Justify that  $F$  is a cumulative distribution function.
2. Justify that the law associated to  $F$  is continuous.
3. Compute the density  $f$  associated to this law.
4. Let  $X$  be a random variable with cumulative distribution function  $F$ .
  - (a) Compute  $\mathbb{P}(X > 1/2)$ .
  - (b) Compute  $\mathbb{E}[X]$ .
  - (c) Compute the variance of  $X$ .

*Solution.* (5 points)

1. (1pt) One can verify that  $F$  is such that

- $\lim_{x \rightarrow -\infty} F(x) = 0$ ;
- $\lim_{x \rightarrow \infty} F(x) = 1$ ;
- $F$  is non-decreasing;

-  $F$  is continuous on the right.

Alternatively, one can check that  $f(x) = \frac{d}{dx}F(x)$  is a valid density function.

2. (0.5pt) Since  $F(x)$  is continuous, it follows that the law associated to  $F$  is continuous.

3. (1pt)

$$f(x) = \begin{cases} 0 & \text{if } x < 0, \\ x/2 & \text{if } 0 \leq x \leq 2, \\ 0 & \text{if } x > 2. \end{cases} \quad (13)$$

4. (2.5pt) One can compute that:

a) (0.5pt)  $\mathbb{P}(X > 1/2) = 15/16$

b) (1pt)  $\mathbb{E}[X] = 4/3$

c) (1pt)  $\text{Var}(X) = \mathbb{E}[X^2] - \mathbb{E}[X]^2 = 2/9$ .

□

**Exercise 5.** *This exercise is significantly harder than the previous ones, but not worth many points. Try it once you have finished the rest of the midterm.*

*We remind that:*

- an exponential random variable with rate  $\lambda$  has density  $f(x) = \lambda e^{-\lambda x} I(x > 0)$ ,
- a gamma random variable with shape parameter  $\alpha$  and rate  $\lambda$  has density

$$f(x) = \frac{\lambda^\alpha}{\Gamma(\alpha)} x^{\alpha-1} e^{-\lambda x} I(x > 0),$$

- if  $n$  is a non-negative integer,  $\Gamma(n+1) = n!$ , and
- a Poisson random variable with rate  $\lambda$  has probability mass function  $f(n) = e^{-\lambda} \lambda^n / n!$ ,  $n = 0, 1, \dots$ .

Let  $X_1, X_2, X_3, \dots$  be an infinite sequence of independent random variables of exponential law with rate parameter 1. We denote  $T_0 = 0$  and for all  $n \geq 1$ ,  $T_n = X_1 + \dots + X_n$ . For all  $t \geq 0$ , we denote  $N_t = \max\{n \geq 0 \mid T_n \leq t\}$ .

1. (a) Compute the cumulative distribution function of  $T_2$ . Deduce that  $T_2$  is a gamma random variable with shape parameter 2 and rate 1. (Like everywhere else, a proof is required.)

(b) Let  $n \geq 1$ . Using the same method, compute the law of  $T_n$ .

2. Let  $t > 0$ . Compute the law of  $N_t$ .

3. Let  $n \geq 1$ .

(a) Compute the joint law of  $(T_1, \dots, T_n)$ .

(b) Compute the conditional law of  $(T_1, \dots, T_n)$  given that  $N_t = n$ .

*Solution.* (5 points)

1. (2pt)

a) (1pt) The CDF of  $T_2$  can be computed as follows:

$$F_{T_2}(t) = \mathbb{P}(X_1 + X_2 \leq t) = \int_0^t dx_1 \int_0^{t-x_1} dx_2 f_{(X_1, X_2)}(x_1, x_2) \quad (14)$$

$$\stackrel{(\text{indep.})}{=} \int_0^t dx_1 \int_0^{t-x_1} dx_2 f_{X_1}(x_1) f_{X_2}(x_2) \quad (15)$$

$$= \int_0^t dx_1 e^{-x_1} \int_0^{t-x_1} dx_2 e^{-x_2} \quad (16)$$

$$= \int_0^t dx_1 e^{-x_1} (1 - e^{-(t-x_1)}) \quad (17)$$

$$= \int_0^t dx_1 (e^{-x_1} - e^{-t}) \quad (18)$$

$$= 1 - e^{-t}(t + 1). \quad (19)$$

Deriving the above with respect to  $t$ , we obtain

$$\frac{d}{dt} F_{T_2}(t) = -e^{-t} + e^{-t}(t + 1) = te^{-t} \quad (20)$$

that is the pdf of a Gamma(2,1) random variable.

b) (1pt) We show by induction that  $T_n \sim \text{Gamma}(n, 1)$ . The base case is given by point a). The inductive step is as follows: assume that  $T_n \sim \text{Gamma}(n, 1)$ , then

$$F_{T_{n+1}}(t) = \mathbb{P}(T_n + X_{n+1} \leq t) = \int_0^t dy \int_0^{t-y} dx f_{(T_n, X_{n+1})}(y, x) \quad (21)$$

$$= \int_0^t dy \int_0^{t-y} dx f_{T_n}(y) f_{X_{n+1}}(x) \quad (22)$$

$$= \int_0^t dy \frac{1}{(n-1)!} y^{n-1} e^{-y} \int_0^{t-y} dx e^{-x} \quad (23)$$

$$= \int_0^t dy \frac{1}{(n-1)!} y^{n-1} e^{-y} (1 - e^{-(t-y)}) \quad (24)$$

$$= \int_0^t dy \frac{1}{(n-1)!} y^{n-1} e^{-y} - e^{-t} \int_0^t dy \frac{1}{(n-1)!} y^{n-1}. \quad (25)$$

Deriving the above with respect to  $t$ , we obtain

$$\frac{d}{dt} F_{T_{n+1}}(t) = \frac{1}{(n-1)!} t^{n-1} e^{-t} - e^{-t} \frac{1}{(n-1)!} t^{n-1} + e^{-t} \int_0^t dy \frac{1}{(n-1)!} y^{n-1} \quad (26)$$

$$= \frac{1}{n!} t^n e^{-t}, \quad (27)$$

that is the pdf of a Gamma distribution with shape parameter  $n + 1$  and scale parameter 1.

2. (2pt) For all  $n \in \mathbb{N}$ , we have  $\{N_t = n\} = \{T_n \leq t, T_{n+1} > t\}$ , since the sequence  $(T_n)_{n \geq 0}$  is increasing. Note that  $T_n$  and  $T_{n+1}$  are not independent, since they both depend on  $X_1, \dots, X_n$ .

However,  $T_n$  and  $X_{n+1}$  are independent. Thus, for  $n \geq 1$ , we get

$$\mathbb{P}(N_t = n) = \mathbb{P}(T_n \leq t, T_n + X_{n+1} > t) \quad (28)$$

$$= \mathbb{P}(T_n \leq t, X_{n+1} > t - T_n) \quad (29)$$

$$= \int_0^t dy \int_{t-y}^{\infty} dx f_{(T_n, X_{n+1})}(y, x) \quad (30)$$

$$= \int_0^t dy f_{T_n}(y) \int_{t-y}^{\infty} dx f_{X_{n+1}}(x) \quad (31)$$

$$= \int_0^t dy \frac{1}{(n-1)!} y^{n-1} e^{-y} \int_{t-y}^{\infty} dx e^{-x} \quad (32)$$

$$= \int_0^t dy \frac{1}{(n-1)!} y^{n-1} e^{-y} e^{-(t-y)} \quad (33)$$

$$= e^{-t} \int_0^t \frac{1}{(n-1)!} y^{n-1} = \frac{1}{n!} t^n e^{-t}, \quad (34)$$

thus  $N_t \sim \text{Poisson}(t)$ .

3. (1pt)

a) (0.5pt) First, consider the case where  $n = 2$ .

$$f_{(T_1, T_2)}(t_1, t_2) = f_{(X_1, X_1 + X_2)}(t_1, t_2) \quad (35)$$

$$= f_{X_1}(t_1) f_{X_1 + X_2 | X_1}(t_2 | t_1) \quad (36)$$

$$= f_{X_1}(t_1) f_{X_2 | X_1}(t_2 - t_1 | t_1). \quad (37)$$

Since  $X_1$  and  $X_2$  are independent we can write  $f_{X_2 | X_1}(t_2 - t_1 | t_1) = f_{X_2}(t_2 - t_1)$ , and plug in the density functions of  $X_1$  and  $X_2$

$$f_{(T_1, T_2)}(t_1, t_2) = f_{X_1}(t_1) f_{X_2}(t_2 - t_1) \quad (38)$$

$$= e^{-t_1} \mathbf{1}(t_1 > 0) e^{-(t_2 - t_1)} \mathbf{1}(t_2 > t_1) \quad (39)$$

$$= e^{-t_2} \mathbf{1}(0 < t_1 < t_2). \quad (40)$$

We now prove by induction that for general  $n$ ,

$$f_{(T_1, \dots, T_n)}(t_1, \dots, t_n) = e^{-t_n} \mathbf{1}(0 < t_1 < t_2 < \dots < t_n) \quad (41)$$

The base case is given by the above. For the inductive step, assume that (41) holds for  $n$ . Then,

$$f_{(T_1, \dots, T_n, T_{n+1})}(t_1, \dots, t_n, t_{n+1}) = f_{(T_1, \dots, T_n, T_n + X_{n+1})}(t_1, \dots, t_n, t_{n+1}) \quad (42)$$

$$= f_{(T_1, \dots, T_n)}(t_1, \dots, t_n) f_{T_n + X_{n+1} | (T_1, \dots, T_n)}(t_{n+1} | t_1, \dots, t_n) \quad (43)$$

$$= f_{(T_1, \dots, T_n)}(t_1, \dots, t_n) f_{X_{n+1} | (T_1, \dots, T_n)}(t_{n+1} - t_n | t_1, \dots, t_n). \quad (44)$$

Observe one more time that  $X_{n+1}$  is independent from  $T_1, \dots, T_n$ , which gives

$$f_{(T_1, \dots, T_n, T_{n+1})}(t_1, \dots, t_n, t_{n+1}) = f_{(T_1, \dots, T_n)}(t_1, \dots, t_n) f_{X_{n+1}}(t_{n+1} - t_n) \quad (45)$$

$$= e^{-t_n} \mathbf{1}(0 < t_1 < t_2 < \dots < t_n) e^{-(t_{n+1} - t_n)} \mathbf{1}(t_{n+1} > t_n) \quad (46)$$

$$= e^{-t_{n+1}} \mathbf{1}(0 < t_1 < t_2 < \dots < t_{n+1}), \quad (47)$$

which concludes the proof.

b) (0.5pt) Recall that  $\{N_t = n\} = \{T_n \leq t, T_{n+1} > t\}$ . We first compute the joint distribution of  $(T_1, \dots, T_n, N_t)$  :

$$f_{(T_1, \dots, T_n, N_t)}(t_1, \dots, t_n, n) = \int_0^\infty dt_{n+1} f_{(T_1, \dots, T_n, T_{n+1}, N_t)}(t_1, \dots, t_n, t_{n+1}, n) \quad (48)$$

$$= \int_0^\infty dt_{n+1} f_{(T_1, \dots, T_n, T_{n+1})}(t_1, \dots, t_n, t_{n+1}) \mathbf{1}(t_n \leq t, t_{n+1} > t) \quad (49)$$

$$= \int_0^\infty dt_{n+1} e^{-t_{n+1}} \mathbf{1}(0 < t_1 < \dots < t_{n+1}) \mathbf{1}(t_n \leq t, t_{n+1} > t) \quad (50)$$

$$= \int_t^\infty dt_{n+1} e^{-t_{n+1}} \mathbf{1}(0 < t_1 < \dots < t_n \leq t) \quad (51)$$

$$= e^{-t} \mathbf{1}(0 < t_1 < t_2 < \dots < t_n \leq t). \quad (52)$$

Applying the formula for the conditional probability we get

$$\begin{aligned} f_{(T_1, \dots, T_n) | N_t}(t_1, \dots, t_n | n) &= \frac{f_{(T_1, \dots, T_n, N_t)}(t_1, \dots, t_n, n)}{\mathbb{P}(N_t = n)} \\ &= \frac{e^{-t} \mathbf{1}(0 < t_1 < t_2 < \dots < t_n \leq t)}{e^{-t} t^n / n!} \\ &= n! t^{-n} \mathbf{1}(0 < t_1 < t_2 < \dots < t_n \leq t). \end{aligned}$$

□