

Duration: 180 minutes

Exam (Época Extraordinária)

- Please justify all your answers.
- This test has THREE PAGES and SIX GROUPS. The total of points is 40.0.

Group 1 — Introduction to Stochastic Processes

2.5 points

Consider a stochastic process  $\{X(t) : t \in \mathbb{R}\}$  where  $X(t) = U \cos(t) + V \sin(t)$  and  $U$  and  $V$  are two r.v.

- (a) Show that the condition  $E(U) = E(V) = 0$  is necessary for  $\{X(t) : t \in \mathbb{R}\}$  to be a second order weakly stationary process. (1.0)

• Stochastic process

$$\{X(t) : t \geq 0\}$$

$$X(t) = U \cos(t) + V \sin(t)$$

$U, V$  r.v.

• Requested proof

In case  $\{X(t) : t \in \mathbb{R}\}$  is a second order weakly stationary process, then its mean function,

$$\begin{aligned} E[X(t)] &= E[U \cos(t) + V \sin(t)] \\ &= E(U) \times \cos(t) + E(V) \sin(t), \end{aligned}$$

cannot depend on  $t$ , i.e., has to be constant. This is possible only if  $E[X(t)] = 0$ , for all  $t \geq 0$ , which in turn implies that

$$\begin{aligned} E(U) &= E(V) \\ &= 0. \quad \checkmark \end{aligned}$$

- (b) Prove that if  $\{X(t) : t \in \mathbb{R}\}$  is a second order weakly stationary process then  $U$  and  $V$  are uncorrelated r.v. with equal variance. (1.5)

**Hint:** Recall that  $X(0) = U$ ;  $X(\frac{\pi}{2}) = V$ ;  $\cos(x)\cos(y) + \sin(x)\sin(y) = \cos(x-y)$ ; and  $\sin(x)\cos(y) - \cos(x)\sin(y) = \sin(x+y)$ .

• Requested proof

We just proved that if  $\{X(t) : t \in \mathbb{R}\}$  is a second order weakly stationary process then  $E(U) = E(V) = 0$ . Moreover,  $V[X(t)]$  has to be constant, namely

$$\begin{aligned} V[X(0)] &= V\left[X\left(\frac{\pi}{2}\right)\right] \\ E[X^2(0)] - E^2[X(0)] &= E\left[X^2\left(\frac{\pi}{2}\right)\right] - E^2\left[X\left(\frac{\pi}{2}\right)\right] \\ E[X^2(0)] &= E\left[X^2\left(\frac{\pi}{2}\right)\right] \\ E(U^2) &= E(V^2) \\ V(U) &= V(V). \quad \checkmark \end{aligned}$$

In addition, the autocovariance function,

$$\begin{aligned} Cov(X(t), X(t+s)) &= E[X(t)X(t+s)] - E[X(t)]E[X(t+s)] \\ &\stackrel{E[X(t)]=E[X(t+s)]=0}{=} E[X(t)X(t+s)] \\ &= E\{[U \cos(t) + V \sin(t)] \times [U \cos(t+s) + V \sin(t+s)]\} \\ &= E[U^2 \cos(t) \cos(t+s) + UV \cos(t) \sin(t+s) \\ &\quad + UV \sin(t) \cos(t+s) + V^2 \sin(t) \sin(t+s)] \\ &\stackrel{E(U^2)=E(V^2)}{=} E(U^2) \times [\cos(t) \cos(t+s) + \sin(t) \sin(t+s)] \\ &\quad + E(UV) \times [\cos(t) \sin(t+s) + \sin(t) \cos(t+s)] \end{aligned}$$

$$\begin{aligned} Cov(X(t), X(t+s)) &= E(U^2) \times \cos(t-t-s) + E(UV) \times \sin(t+t+s) \\ &= E(U^2) \times \cos(-s) + E(UV) \times \sin(2t+s) \\ &= E(U^2) \times \cos(s) + E(UV) \times \sin(2t+s), \end{aligned}$$

cannot not depend on  $t$ , it has to depend exclusively on the time lag  $s$ . This implies that  $E(UV) = 0$ .

If we combine this result and the fact that  $E(U) = E(V) = 0$ , we conclude that

$$\begin{aligned} E(UV) &= E(UV) - E(U) \times E(V) \\ &= Cov(U, V) \\ &= 0, \end{aligned}$$

and this means that  $U$  and  $V$  are two uncorrelated r.v.  $\checkmark$

Group 2 — Poisson Processes

9.5 points

1. One estimates that meteors enter the atmosphere in a specific region of the globe according to a Poisson process with rate  $\lambda$  equal to 100 meteors per hour and that 1% of those meteors are visible to the *naked eye* as shooting stars.

- (a) What is the probability that an observer sees at least two shooting stars in 30 minutes? (1.0)

• Stochastic process

$$\{N(t) : t \geq 0\} \sim PP(\lambda = 100)$$

$N(t)$  = number of meteors entering the atmosphere by time (in hours)  $t$

• Split process

$$\{N_{ss}(t) : t \geq 0\} \sim PP(\lambda_{ss} = \lambda p = 100 \times 0.01 = 1)$$

$N_{ss}(t)$  = number of meteors seen as shooting stars by time (in hours)  $t$

$$N_{ss}(t) \sim \text{Poisson}(\lambda p t = t)$$

• Requested probability

$$\begin{aligned} P[N_{ss}(t) \geq 2] &= 1 - P[N_{ss}(t) \leq 1] \\ &= 1 - F_{\text{Poisson}(t)}(1) \\ &= 1 - F_{\text{Poisson}(0.5)}(1) \\ &\stackrel{\text{tables}}{=} 1 - 0.9098 \\ &= 0.0902. \end{aligned}$$

- (b) Find the expected waiting time until a fifth meteor is visible to the *naked eye* as a shooting star. What is the probability that this waiting time does not exceed its expected value? (1.0)

• Relevant r.v. and its distribution

$S_n$  = arrival time of the  $n$ th meteor seen as a shooting star

$$S_n \stackrel{\text{form.}}{\sim} \text{gamma}(n, \lambda_{ss})$$

$$n = 5$$

$$\lambda_{ss} = \lambda p = 1$$

• Requested expected value and probability

$$E(S_n) \stackrel{\text{form.}}{=} \frac{n}{\lambda_{ss}} = 5$$

$$\begin{aligned} P[S_n \leq E(S_n)] &= F_{\text{Erlang}(n, \lambda_{ss})}(E(S_n)) \\ &= [1 - F_{\text{Poisson}(\lambda_{ss} \times E(S_n))}(n-1)] \\ &= 1 - F_{\text{Poisson}(n)}(n-1) \\ &\stackrel{n=5}{=} 1 - F_{\text{Poisson}(5)}(4) \\ &\stackrel{\text{tables}}{=} 1 - 0.4405 = 0.5595. \end{aligned}$$

- (c) Find the joint probability that the cumulative number of meteors seen as shooting stars is equal to 6 at time 6, 15 at time 15, and 27 at time 27. (1.0)

• **Relevant fact**

$\{N_{ss}(t) : t \geq 0\}$  has independent and stationary increments

• **Requested probability**

$$P[N_{ss}(6) = 6, N_{ss}(15) = 15, N_{ss}(27) = 27]$$

$$\stackrel{indep. incr.}{=} P[N_{ss}(6) = 6, N_{ss}(15) - N_{ss}(6) = 15 - 6, N_{ss}(27) - N_{ss}(15) = 27 - 15]$$

$$\stackrel{station. incr.}{=} P[N_{ss}(6) = 6] \times P[N_{ss}(15) - N_{ss}(6) = 15 - 6] \times P[N_{ss}(27) - N_{ss}(15) = 27 - 15]$$

$$\stackrel{station. incr.}{=} P[N_{ss}(6) = 6] \times P[N_{ss}(15 - 6) = 15 - 6] \times P[N_{ss}(27 - 15) = 27 - 15]$$

$$\stackrel{station. incr.}{=} P[N_{ss}(6) = 6] \times P[N_{ss}(9) = 9] \times P[N_{ss}(12) = 12]$$

$$\stackrel{N(t) \sim \text{Poisson}(\lambda t)}{=} e^{-6} \frac{6^6}{6!} \times e^{-9} \frac{9^9}{9!} \times e^{-12} \frac{12^{12}}{12!}$$

$$\approx 0.002420.$$

[Using the tables of the Poisson c.d.f. we would get  $(0.7440 - 0.6063) \times (0.5874 - 0.4557) \times (0.5760 - 0.4616) \approx 0.002075$ .]

- (d) Suppose 4 meteors were seen as shooting stars during the first 4 hours. Obtain the probability that at most 2 arrived during the first hour? (0.5)

• **Relevant r.v.**

$$(N(s) | N(t) = n) \stackrel{form.}{\sim} \text{binomial}(n, s/t), 0 < s < t$$

• **Requested probability**

$$P[N(s) \leq 2 | N(4) = 4] \stackrel{tables}{=} F_{\text{binomial}(n=4, s/t=1/4)}(2) = 0.9492.$$

2. Let  $\{N_1(t) : t \geq 0\}$  and  $\{N_2(t) : t \geq 0\}$  be two independent nonhomogeneous Poisson processes, with intensity functions  $\lambda_1(t)$  and  $\lambda_2(t)$ , respectively. (2.5)

Prove that merging these two stochastic processes results in a nonhomogeneous Poisson process with intensity function  $\lambda_1(t) + \lambda_2(t)$ .

**Hint:**  $\{N(t) : t \geq 0\} \sim \text{NHPP}(\lambda(t))$  iff: i)  $N(0) = 0$ ; ii)  $\{N(t) : t \geq 0\}$  has independent increments; iii)  $P[N(t+h) - N(t) = 1] = \lambda(t) \times h + o(h)$ ,  $t \geq 0$ ; iv)  $P[N(t+h) - N(t) \geq 2] = o(h)$ ,  $t \geq 0$ . Moreover, a function  $f$  is said to be  $o(h)$  iff  $\lim_{h \rightarrow 0} \frac{f(h)}{h} = 0$ .

• **Stochastic processes**

$$\{N_i(t) : t \geq 0\} \sim \text{NHPP}(\lambda_i(t)), \quad i = 1, 2$$

$$\{N_1(t) : t \geq 0\} \perp\!\!\!\perp \{N_2(t) : t \geq 0\}$$

• **Merged stochastic process**

$$\{N(t) = N_1(t) + N_2(t) : t \geq 0\}$$

• **Important result**

$$N_i(t+s) - N_i(s) \stackrel{indep.}{\sim} \text{Poisson}(m_i(t+s) - m_i(s)), \quad i = 1, 2$$

$$m_i(t) = \int_0^t \lambda_i(z) dz, \quad i = 1, 2$$

• **Requested proof**

Let us take advantage of the hint and capitalize on the properties of  $\{N_i(t) : t \geq 0\}$ ,  $i = 1, 2$ .

- i) Is  $N(0) = 0$ ?

$$N(0) = N_1(0) + N_2(0) = 0. \quad \checkmark$$

- ii) Has  $\{N(t) : t \geq 0\}$  independent increments?

Let  $0 < t_1 < t_2 < \dots < t_n$  ( $n = 2, 3, \dots$ ) and  $0 \leq i_1 \leq i_2 \leq \dots \leq i_n$ .

Since  $\{N_1(t) : t \geq 0\}$  and  $\{N_2(t) : t \geq 0\}$  are two independent stochastic processes, in any case with independent increments, the joint probability

$$P[N(t_1) = i_1, N(t_2) - N(t_1) = i_2 - i_1, \dots, N(t_n) - N(t_{n-1}) = i_n - i_{n-1}] = \star$$

is equal to

$$\begin{aligned} \star &= P[N_1(t_1) + N_2(t_1) = i_1, N_1(t_2) + N_2(t_2) - N_1(t_1) - N_2(t_1) = i_2 - i_1, \dots, \\ &\quad N_1(t_n) + N_2(t_n) - N_1(t_{n-1}) - N_2(t_{n-1}) = i_n - i_{n-1}] \\ &= P\{N_1(t_1) + N_2(t_1) = i_1, [N_1(t_2) - N_1(t_1)] + [N_2(t_2) - N_2(t_1)] = i_2 - i_1, \dots, \\ &\quad [N_1(t_n) - N_1(t_{n-1})] + [N_2(t_n) - N_2(t_{n-1})] = i_n - i_{n-1}\} \\ &= P[N_1(t_1) + N_2(t_1) = i_1] \times P[N_1(t_2) + N_2(t_2) - N_1(t_1) - N_2(t_1) = i_2 - i_1] \times \dots \\ &\quad \times P[N_1(t_n) + N_2(t_n) - N_1(t_{n-1}) - N_2(t_{n-1}) = i_n - i_{n-1}] \\ &= P[N(t_1) = i_1] \times P[N(t_2) - N(t_1) = i_2 - i_1] \times \dots \\ &\quad \times P[N(t_n) - N(t_{n-1}) = i_n - i_{n-1}], \end{aligned}$$

that is, the r.v.  $N(t_1)$ ,  $N(t_2) - N(t_1)$ ,  $\dots$ ,  $N(t_n) - N(t_{n-1})$  are independent, thus  $\{N(t) : t \geq 0\}$  also has independent increments.  $\checkmark$

- iii)  $P[N(t+h) - N(t) = 1] = [\lambda_1(t) + \lambda_2(t)] \times h + o(h)$ ,  $t \geq 0$ ?

In other words, we have to verify that  $\lim_{h \rightarrow 0} \frac{P[N(t+h) - N(t) = 1]}{h} = \lambda_1(t) + \lambda_2(t)$ .

There will be exactly one event in the merged stochastic process in the interval  $(t, t+h]$  if there is:

- \* exactly one event of the  $N$ -process and 0 events of the  $N$ -process, with probability

$$\begin{aligned} P[N_1(t+h) - N_1(t) = 1, \quad N_2(t+h) - N_2(t) = 0] \\ = P[N_1(t+h) - N_1(t) = 1] \times P[N_2(t+h) - N_2(t) = 0]; \end{aligned}$$

- \* or exactly one event of the  $N$ -process and 0 events of the  $N$ -process, with probability

$$\begin{aligned} P[N_1(t+h) - N_1(t) = 0, \quad N_2(t+h) - N_2(t) = 1] \\ = P[N_1(t+h) - N_1(t) = 0] \times P[N_2(t+h) - N_2(t) = 1]; \end{aligned}$$

Since these two events are mutually exclusive, we get

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{P[N(t+h) - N(t) = 1]}{h} &= \lim_{h \rightarrow 0} \frac{P[N_1(t+h) - N_1(t) = 1] \times P[N_2(t+h) - N_2(t) = 0]}{h} \\ &\quad + \lim_{h \rightarrow 0} \frac{P[N_1(t+h) - N_1(t) = 0] \times P[N_2(t+h) - N_2(t) = 1]}{h} \\ &= \lim_{h \rightarrow 0} \frac{e^{-[m_1(t+h) - m_1(t)]} [m_1(t+h) - m_1(t)] \times e^{-[m_2(t+h) - m_2(t)]}}{h} \\ &\quad + \lim_{h \rightarrow 0} \frac{e^{-[m_1(t+h) - m_1(t)]} \times e^{-[m_2(t+h) - m_2(t)]} [m_2(t+h) - m_2(t)]}{h} \\ &= \lim_{h \rightarrow 0} \frac{m_1(t+h) - m_1(t)}{h} + \lim_{h \rightarrow 0} \frac{m_2(t+h) - m_2(t)}{h} \\ &= \lambda_1(t) + \lambda_2(t). \quad \checkmark \end{aligned}$$

- iv)  $P[N(t+h) - N(t) \geq 2] = o(h)$ ,  $t \geq 0$ ?

Equivalently, we have to verify that  $\lim_{h \rightarrow 0} \frac{P[N(t+h) - N(t) \geq 2]}{h} = 0$ . Similarly and invoking L'Hôpital rule,

$$\begin{aligned} \lim_{h \rightarrow 0} \frac{P[N(t+h) - N(t) \geq 2]}{h} &= \lim_{h \rightarrow 0} \frac{1 - P[N(t+h) - N(t) = 0, 1]}{h} \\ &\stackrel{ii)}{=} \lim_{h \rightarrow 0} \frac{1 - P[N(t+h) - N(t) = 0]}{h} - [\lambda_1(t) + \lambda_2(t)] \\ &= \lim_{h \rightarrow 0} \frac{1 - P[N_1(t+h) - N_1(t) = 0] \times P[N_2(t+h) - N_2(t) = 0]}{h} \\ &\quad - [\lambda_1(t) + \lambda_2(t)] \\ &= \lim_{h \rightarrow 0} \frac{1 - e^{-[m_1(t+h) - m_1(t)]} \times e^{-[m_2(t+h) - m_2(t)]}}{h} - [\lambda_1(t) + \lambda_2(t)] \end{aligned}$$

$$\lim_{h \rightarrow 0} \frac{P[N(t+h) - N(t) \geq 2]}{h} \stackrel{\text{Hrule}}{=} \lim_{h \rightarrow 0} \frac{\frac{d[1 - e^{-[m_1(t+h) - m_1(t)]} \times e^{-[m_2(t+h) - m_2(t)]}]}{dh}}{1} - [\lambda(t) + \mu(t)]$$

$$= \lim_{h \rightarrow 0} \lambda_1(t+h) e^{-[m_1(t+h) - m_1(t)] - [m_2(t+h) - m_2(t)]} + \lim_{h \rightarrow 0} \lambda_2(t+h) e^{-[m_1(t+h) - m_1(t)] - [m_2(t+h) - m_2(t)]} - [\lambda_1(t) + \lambda_2(t)]$$

$$= 0. \quad \checkmark$$

Thus,  $\{N^*(t) : t \geq 0\}$  is indeed a NHPP with intensity function  $\lambda_1(t) + \lambda_2(t)$ .  $\checkmark$

3. Admit that: buses arrive at a music festival in accordance with a Poisson process  $\{N(t) : t \geq 0\}$  with rate  $\lambda$ ; the numbers of *festivalgoers* in each bus are independent and identically distributed to  $Y \sim \text{geometric}(p)$ ; these r.v. are independent of  $N(t)$ .

Let  $X(t)$  represent the number of *festivalgoers* who have arrived by time  $t$ .

(a) Obtain the mean and variance of  $X(t)$ . (1.0)

• **Compound PP**

$$\{X(t) = \sum_{i=1}^{N(t)} Y_i : t > 0\} \sim \text{CpPP}(\lambda, Y)$$

$$\{N(t) : t > 0\} \sim \text{PP}(\lambda)$$

$Y_i$  = number of festivalgoers in bus  $i$

$$Y_i \stackrel{i.i.d.}{\sim} Y, \quad i \in \mathbb{N}$$

$Y \sim \text{geometric}(p)$

• **Requested mean and variance**

$$E[X(t)] \stackrel{\text{form.}}{=} \lambda t \times E(Y)$$

$$\stackrel{\text{form.}}{=} \frac{\lambda t}{p}$$

$$V[X(t)] \stackrel{\text{form.}}{=} \lambda t \times E(Y^2)$$

$$= \lambda t \times [V(Y) + E^2(Y)]$$

$$\stackrel{\text{form.}}{=} \lambda t \times \left( \frac{1-p}{p^2} + \frac{1}{p^2} \right)$$

$$= \frac{\lambda t(2-p)}{p^2}.$$

(b) Derive  $P[X(t) = 0]$  and a simplified expression for  $P[X(t) = x]$ ,  $x \in \mathbb{N}$ . (1.5)

• **Requested probability**

Since  $Y$  is a positive integer r.v.,

$$P[X(t) = 0] = P[N(t) = 0]$$

$$= e^{-\lambda t}.$$

• **Requested p.f.**

Recall that  $(X(t) | N(t) = n) \sim \text{negBinomial}(n, p)$  and the number of buses cannot exceed the number of festivalgoers, hence the application of the total probability law leads to

$$P[X(t) = x] = \sum_{n=1}^x P[X(t) = x | N(t) = n] \times P[N(t) = n]$$

$$\stackrel{\text{form.}}{=} \sum_{n=1}^x \binom{x-1}{n-1} p^n (1-p)^{x-n} \times \frac{e^{-\lambda t} (\lambda t)^n}{n!}$$

$$= e^{-\lambda t} (1-p)^x \sum_{n=1}^x \binom{x-1}{n-1} \frac{\left(\frac{\lambda t p}{1-p}\right)^n}{n!}, \quad x \in \mathbb{N}.$$

• **[Comment]**

$X(t)$  is said to have a *geometric Poisson distribution* or a *Pólya-Aeppli distribution*. This distribution is defined in *Mathematica* with a slightly different parametrization:  $p' = 1 - p$ .

(c) Consider  $\lambda = 1$  (bus per minute),  $p = 0.025$ ,  $t = 720$  minutes, and  $x = 28000$ . (1.0)  
Use the central limit theorem to compute an approximate value to  $P\{X(t) \leq x\}$ .

• **Requested probability**

We can apply the CLT to provide the following approximate value:

$$P\{X(t) \leq x\} \approx \Phi\left(\frac{x - E[X(t)]}{\sqrt{V[X(t)]}}\right)$$

$$= \Phi\left(\frac{x - \frac{\lambda t}{p}}{\sqrt{\frac{\lambda t(2-p)}{p^2}}}\right)$$

$$= \Phi\left(\frac{28000 - \frac{1 \times 720}{0.025}}{\sqrt{\frac{1 \times 720 \times (2 - 0.025)}{0.025^2}}}\right)$$

$$= \Phi\left(\frac{28000 - 28800}{\sqrt{2275200}}\right)$$

$$\approx \Phi(-0.53)$$

$$\approx 1 - \Phi(0.53)$$

$$\stackrel{\text{tables}}{=} 1 - 0.7019$$

$$= 0.2981.$$

**Group 3 — Renewal Processes**

8.0 points

1. Admit that:  $N(t)$  represents the cumulative number of covid-19 cases in a region in the interval  $(0, t]$ ;  $\{N(t) : t \geq 0\}$  is a renewal process with renewal function

$$m(t) = \frac{t}{3} + \frac{e^{-\frac{3t}{2}} \left[ \sin\left(\frac{\sqrt{3}t}{2}\right) + \sqrt{3} \cos\left(\frac{\sqrt{3}t}{2}\right) \right]}{3\sqrt{3}} - \frac{1}{3}, \quad t \geq 0.$$

(a) Derive the inter-renewal distribution of  $\{N(t) : t \geq 0\}$ . (1.5)

**Hint:**  $\tilde{m}(s) = \int_0^{+\infty} e^{-st} dm(t) = \frac{1}{s^3 + 3s^2 + 3s}$ ,  $s > 0$ .

• **Renewal process**

$$\{N(t) : t \geq 0\}$$

$N(t)$  = cumulative number of covid-19 cases until time  $t$

• **Inter-renewal times**

$$X_i \stackrel{i.i.d.}{\sim} X, \quad i \in \mathbb{N}$$

• **LST of the renewal function**

$$\tilde{m}(s) = \int_0^{+\infty} e^{-st} dm(t) = \frac{1}{s^3 + 3s^2 + 3s}$$

• **Deriving the inter-renewal distribution**

Since  $\tilde{m}(s) \stackrel{\text{form.}}{=} \frac{\tilde{F}(s)}{1 - \tilde{F}(s)}$  the LST of the inter-renewal distribution can be obtained in terms of the one of  $m$ :

$$\tilde{F}(s) = \int_0^{+\infty} e^{-sx} dF(x)$$

$$\stackrel{\text{form.}}{=} \frac{\tilde{m}(s)}{1 + \tilde{m}(s)}$$

$$\begin{aligned}\tilde{F}(s) &= \frac{1}{s^3 + 3s^2 + 3s} \\ &= \frac{1}{1 + \frac{1}{s^3 + 3s^2 + 3s}} \\ &= \frac{1}{\frac{s^3 + 3s^2 + 3s + 1}{s^3 + 3s^2 + 3s}} \\ &= \frac{1}{s^3 + 3s^2 + 3s + 1} \\ &= \left(\frac{1}{1+s}\right)^3 \\ &\equiv M_{\text{gamma}(3,1)}(-s),\end{aligned}$$

i.e., the IRT are  $X_i \stackrel{i.i.d.}{\sim} X \sim \text{gamma}(3, 1)$ ,  $i \in \mathbb{N}$ .

(b) Compute the expected number of covid-19 cases in the interval  $(t, 2t]$ , where  $t = \frac{\pi}{\sqrt{3}}$ . (1.0)

• Requested value

$$\begin{aligned}m(2t) - m(t) &= \left\{ \frac{2t}{3} + \frac{e^{-3t} [\sin(\sqrt{3}t) + \sqrt{3} \cos(\sqrt{3}t)] - \frac{1}{3}}{3\sqrt{3}} \right\} \\ &\quad - \left\{ \frac{t}{3} + \frac{e^{-\frac{3t}{2}} [\sin(\frac{\sqrt{3}t}{2}) + \sqrt{3} \cos(\frac{\sqrt{3}t}{2})] - \frac{1}{3}}{3\sqrt{3}} \right\} \\ &= \frac{t}{3} + \frac{e^{-3t} [\sin(\sqrt{3}t) + \sqrt{3} \cos(\sqrt{3}t)] - e^{-\frac{3t}{2}} [\sin(\frac{\sqrt{3}t}{2}) + \sqrt{3} \cos(\frac{\sqrt{3}t}{2})]}{3\sqrt{3}} \\ &\stackrel{t = \frac{\pi}{\sqrt{3}}}{=} \frac{\pi}{3\sqrt{3}} + \frac{e^{-\sqrt{3}\pi} [\sin(\pi) + \sqrt{3} \cos(\pi)] - e^{-\frac{\sqrt{3}\pi}{2}} [\sin(\frac{\pi}{2}) + \sqrt{3} \cos(\frac{\pi}{2})]}{3\sqrt{3}} \\ &= \frac{\pi - \sqrt{3}e^{-\sqrt{3}\pi} - e^{-\frac{\sqrt{3}\pi}{2}}}{3\sqrt{3}} \\ &\approx 0.590487.\end{aligned}$$

(c) Consider time in hours and the time origin the beginning of 2020. (1.5)

Provide an approximate value to the expected number of covid-19 cases bound to occur on the last day of the second week.

• Expected IRT

$$\mu = E(X) = E[\text{gamma}(3, 1)] = \frac{3}{1} = 3$$

• Requested expected value

Since the inter-renewal distribution is non-lattice, we can apply Blackwell's theorem and state that, for very large  $t$  (such as  $t = 13 \times 24 = 576$  hours), the expected number of covid-19 cases in the interval  $(t, t+a]$  (for  $a = 24$  hours),  $E[N(t+a) - N(t)]$ , can be approximated as follows:

$$\begin{aligned}E[N(t+a) - N(t)] &\approx \lim_{z \rightarrow +\infty} [m(z+a) - m(z)] \\ &\equiv \frac{a}{\mu} \\ &= \frac{24}{3} \\ &= 8.\end{aligned}$$

(d) Admit an inspection was made on January 1, 2021. (2.0)

Obtain an approximate value to the probability that the first covid-19 case after this inspection occurred after January 2, 2021.

• C.d.f. of the IRT

$$\begin{aligned}F(u) &= F_{\text{gamma}(3,1)}(u) \\ \text{form.} &= 1 - F_{\text{Poisson}}(u)(3-1)\end{aligned}$$

$$\begin{aligned}F(u) &= 1 - \sum_{i=0}^2 \frac{e^{-u} u^i}{i!} \\ &= 1 - e^{-u} \left(1 + u + \frac{u^2}{2}\right)\end{aligned}$$

• Recurrence time

$Y(t) \stackrel{\text{form}}{=} S_{N(t)+1} - t =$  time until the first covid-19 case after the inspection at time  $t$

• Requested probability (approximate value)

Since the value of  $t = 365 \times 24 = 8760$  hours is rather large, we can provide the following approximate value to  $P[S_{N(t)+1} > t+x] = \star$

$$\begin{aligned}\star &= 1 - P[S_{N(t)+1} - t \leq x] \\ &= 1 - F_{Y(t)}(x)\end{aligned}$$

$$\approx 1 - \lim_{z \rightarrow +\infty} P[Y(z) \leq x]$$

$$\stackrel{\text{form.}}{=} 1 - \frac{\int_0^x [1 - F(u)] du}{E(X)}$$

$$= 1 - \frac{1}{E(X)} \times \int_0^x e^{-u} \left(1 + u + \frac{u^2}{2}\right) du$$

$$= 1 - \frac{1}{E(X)} \times \int_0^x [f_{\text{gamma}(1,1)}(u) + f_{\text{gamma}(2,1)}(u) + f_{\text{gamma}(3,1)}(u)] du$$

$$= 1 - \frac{1}{E(X)} \times [F_{\text{gamma}(1,1)}(x) + F_{\text{gamma}(2,1)}(x) + F_{\text{gamma}(3,1)}(x)]$$

$$\stackrel{\text{form.}}{=} 1 - \frac{[1 - F_{\text{Poisson}}(x)(1-1)] + [1 - F_{\text{Poisson}}(x)(2-1)] + [1 - F_{\text{Poisson}}(x)(3-1)]}{E(X)}$$

$$\stackrel{x=1}{=} 1 - \frac{3 - F_{\text{Poisson}}(1)(0) - F_{\text{Poisson}}(1)(1) - F_{\text{Poisson}}(1)(2)}{E(X)}$$

$$\stackrel{\text{tables}}{=} 1 - \frac{3 - 0.3679 - 0.7358 - 0.9197}{3}$$

$$\approx 0.674467.$$

[The result obtained using Mathematica is 0.674446.]

2. The duration of the consecutive phone calls made by a seller are independent r.v. with common c.d.f. (2.0)

$$F_X(x) = \begin{cases} 0, & x \leq 0 \\ x^2(3-2x), & 0 < x < \tau \\ 1, & x \geq \tau, \end{cases}$$

where  $\tau$  is a constant in  $(0, \frac{1}{2}]$ . If the seller manages to persuade a customer to acquire a product before time  $\tau$ , then the seller gets a reward of one monetary unit.

Derive the expected reward per time unit in the long-run.

**Hint:** Recall that for any non-negative r.v.  $X$ ,  $E(X) = \int_0^{+\infty} [1 - F_X(x)] dx$ .

• Renewal process

$$\{N(t) : t \geq 0\}$$

$N(t) =$  number of phone calls by time  $t$

• IRT

$$X_n \stackrel{i.i.d.}{\sim} X, \quad n \in \mathbb{N}$$

$$F(x) = F_X(x) = \begin{cases} 0, & x \leq 0 \\ x^2(3-2x), & 0 < x < \tau \\ 1, & x \geq \tau \end{cases}$$

• Reward renewal process

$$\{R(t) = \sum_{n=1}^{N(t)} R_n : t \geq 0\}$$

$R(t)$  = total amount of rewards got by the seller until time  $t$

$$R_n = \begin{cases} 1, & \text{if } X_n < \tau \text{ (i.e., during the } n^{\text{th}} \text{ phone call the seller persuaded the customer} \\ & \text{to acquire a product before time } \tau) \\ 0, & \text{otherwise} \end{cases}$$

$(X_n, R_n) \stackrel{i.i.d.}{\sim} (X, R), n \in \mathbb{N}$

• **Expected IRT**

$$\begin{aligned} E(X) &\stackrel{X \geq 0}{=} \int_0^{+\infty} [1 - F_X(x)] dx \\ &= \int_0^\tau [1 - x^2(3 - 2x)] dx \\ &= \int_0^\tau (1 - 3x^2 + 2x^3) dx \\ &= \left( x - x^3 + \frac{x^4}{2} \right) \Big|_0^\tau \\ &= \tau - \tau^3 + \frac{\tau^4}{2} \\ &= \tau \left( 1 - \tau^2 + \frac{\tau^3}{2} \right) \end{aligned}$$

• **Expected reward per phone call**

$$\begin{aligned} E(R) &= 1 \times P(X < \tau) + 0 \times P(X \geq \tau) \\ &= \lim_{h \rightarrow 0^+} F_X(\tau - h) \\ &= \tau^2(3 - 2\tau) \end{aligned}$$

• **Expected reward per time unit in the long-run**

Since  $E(X), E(R) < +\infty$ , we can invoke the ERT for renewal reward processes and add that the expected reward per time unit in the long-run is given by

$$\begin{aligned} \lim_{t \rightarrow +\infty} \frac{E[R(t)]}{t} &= \frac{E(R)}{E(X)} \\ &= \frac{\tau^2(3 - 2\tau)}{\tau \left( 1 - \tau^2 + \frac{\tau^3}{2} \right)} \\ &= \frac{2(3 - 2\tau)\tau}{\tau^3 - 2\tau^2 + 2}. \end{aligned}$$

• **[Note: the value of  $\tau$  ( $\tau \in (0, \frac{1}{2}]$ ) that maximizes the expected reward per time unit in the long-run,  $h(\tau) = \frac{E(R)}{E(X)}$ , is  $\tau = \frac{1}{2}$ .]**

**Group 4 — Renewal Processes (cont'd)**

2.5 points

Let  $\{X_n : n \in \mathbb{N}_0\}$  be an irreducible, aperiodic, positive recurrent DTMC with limiting distribution  $\{\pi_j : j \in \mathcal{S}\}$ , where  $\mathcal{S}$  represents the state space of this DTMC. (2.5)

Use regenerative processes to show that the mean inter-visit time to state  $j$  is given by  $\frac{1}{\pi_j}$ .

• **Proof**

Consider an irreducible, aperiodic, positive recurrent DTMC that is initially in state  $i$ .

By the Markovian property, each time the process reenters state  $i$  the DTMC restarts probabilistically. Thus returns to state  $i$  are renewals and constitute the beginnings of new cycles.

The expected length of each cycle is  $\mu_{ii}$  and the IRT are constant an equal to 1.

When we are dealing with a regenerative renewal process  $\{X(t) : t \geq 0\}$ , we have

$$\begin{aligned} P_j &= \lim_{t \rightarrow +\infty} P[X(t) = j] \\ \lim_{t \rightarrow +\infty} P[X(t) = j] &= \frac{E[\text{amount of time in } j \text{ during a cycle}]}{E[\text{time of a cycle}]} \\ \lim_{n \rightarrow +\infty} P_{ij}^n &= \frac{1}{\mu_{ii}} \\ \pi_j &= \frac{1}{\mu_{ii}} \end{aligned}$$

If we take  $j$  to equal  $i$ , then we obtain

$$\begin{aligned} \pi_j &= \frac{1}{\mu_{jj}} \\ \mu_{jj} &= \frac{1}{\pi_j} \quad \checkmark \end{aligned}$$

**Group 5 — Discrete time Markov chains**

9.5 points

1. The patients in a hospital are classified as belonging to the following states: (1) coronary care unit; (2) intensive care unit; (3) ambulatory unit; (4) extended care unit; (5) home or dead. As soon as a patient goes home or dies, a new patient is admitted to the coronary unit. The successive states the patient belongs to form a DTMC  $\{X_n : n \in \mathbb{N}_0\}$  with TPM

$$\mathbf{P} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0.1 & 0 & 0 & 0.9 & 0 \\ 0.1 & 0.1 & 0.1 & 0.5 & 0.2 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}.$$

(a) Draw the associated transition diagram, identify and classify the communicating classes of this DTMC. (1.5)

• **DTMC**

$\{X_n : n \in \mathbb{N}_0\}$

• **State space**

$\mathcal{S} = \{1, 2, 3, 4, 5\}$

• **TPM**

$$\mathbf{P} = \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ \frac{1}{10} & 0 & 0 & \frac{9}{10} & 0 \\ \frac{1}{10} & \frac{1}{10} & \frac{1}{10} & \frac{1}{2} & \frac{1}{5} \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

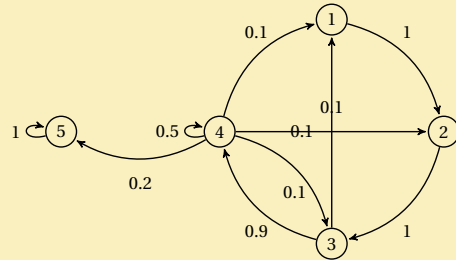
• **Communicating classes and classification of the states of the DTMC**

Judging by the transition diagram, the communicating classes of this Markov chain are  $\{1, 2, 3, 4\}$ , and  $\{5\}$ . Hence, the state space is not a single communicating class therefore we are dealing with a reducible DTMC.

Since state 5 is an absorbing state, that is,  $P_{55} = 1$ , no other state is accessible from it. Needless to say that  $\{5\}$  is a closed communicating class with a recurrent state.

Note that while state 5 is accessible from states 1, 2, 3, 4, the reverse is not true.  $\{1, 2, 3, 4\}$  is not a closed communicating class. Moreover, we can also add these 4 states are transient, because given that the DTMC starts in any of them, there is a non-zero probability that it will never return to them.

• **Transition diagram**



- (b) Admit that the initial state  $X_0$  has p.f.  $\underline{\alpha} = [0.4 \ 0.2 \ 0.2 \ 0.2 \ 0]$ . Compute  $P(X_1 = 1, X_2 = 2, X_4 = 4)$ . (1.5)

• **Initial state**

$X_0$   
 $\underline{\alpha} = [P(X_0 = i)]_{i \in \mathcal{S}} = [0.4 \ 0.2 \ 0.2 \ 0.2 \ 0]$

• **Requested probability**

By applying the multiplication rule and the Markov property, we get

$$P(X_1 = 1, X_2 = 2, X_4 = 4) = P(X_1 = 1) \times P(X_2 = 2 | X_1 = 1) \times P(X_4 = 4 | X_2 = 2)$$

$$\stackrel{\text{form.}}{=} (\underline{\alpha} \times \mathbf{P})_1 \times P_{12} \times P_{24}^2$$

where:

$$P(X_1 = 1) = \underline{\alpha} \times \text{1st. column of } \mathbf{P}$$

$$= [0.4 \ 0.2 \ 0.2 \ 0.2 \ 0] \times [0 \ 0 \ 0.1 \ 0.1 \ 0]^T$$

$$= 0.04;$$

$$P(X_2 = 2 | X_1 = 2) = P_{12}$$

$$= 1;$$

$$P(X_4 = 4 | X_2 = 2) = P_{24}^2$$

$$= \text{2nd. row of } \mathbf{P} \times \text{4th. column of } \mathbf{P}$$

$$= [0 \ 0 \ 1 \ 0 \ 0] \times [0 \ 0 \ 0.9 \ 0.5 \ 0]^T$$

$$= 0.9.$$

Thus,  $P(X_1 = 1, X_2 = 2, X_4 = 4) = 0.04 \times 1 \times 0.9 = 0.036$ .

- (c) Determine the expected number of transitions until a patient reaches state 5, given that he/she started in the coronary care unit ( $X_0 = 1$ ). (2.0)

**Hint:** Check the footnote.<sup>1</sup>

• **Initial state**

$X_0 = 1$

<sup>1</sup>The following results may come handy:

$$\begin{bmatrix} 1 & -\frac{5}{6} & 0 & 0 & 0 \\ 0 & 1 & -\frac{5}{6} & 0 & 0 \\ -\frac{1}{12} & 0 & 1 & -\frac{3}{4} & 0 \\ -\frac{1}{12} & -\frac{1}{12} & -\frac{1}{12} & \frac{7}{12} & -\frac{1}{6} \\ 0 & 0 & 0 & 0 & \frac{1}{6} \end{bmatrix}^{-1} = \begin{bmatrix} \frac{243}{203} & \frac{225}{203} & \frac{30}{29} & \frac{270}{203} & \frac{270}{203} \\ \frac{48}{203} & \frac{270}{203} & \frac{36}{29} & \frac{324}{203} & \frac{324}{203} \\ \frac{288}{1015} & \frac{402}{1015} & \frac{216}{145} & \frac{1944}{1015} & \frac{1944}{1015} \\ \frac{249}{1015} & \frac{411}{1015} & \frac{78}{145} & \frac{2442}{1015} & \frac{2442}{1015} \\ 0 & 0 & 0 & 0 & 6 \end{bmatrix}$$

• **Important**

We are dealing with an absorbing DTMC. The sub-stochastic matrix governing the transitions between its transient states ( $T = \{1, 2, 3, 4\}$ ) is given by

$$\mathbf{Q} = \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ \frac{1}{10} & 0 & 0 & \frac{9}{10} \\ \frac{1}{10} & \frac{1}{10} & \frac{1}{10} & \frac{1}{2} \end{bmatrix}$$

• **Requested expected value**

Let  $\tau = \inf\{n \in \mathbb{N}_0 : X_n \notin T\}$  be the number of transitions until a patient reaches the absorbing state 5.

The result in the footnote yields

$$[E(\tau | X_1 = i)]_{i \in T} \stackrel{\text{form.}}{=} (\mathbf{I} - \mathbf{Q})^{-1} \times \underline{1}$$

$$= \left( \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} - \begin{bmatrix} 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ \frac{1}{10} & 0 & 0 & \frac{9}{10} \\ \frac{1}{10} & \frac{1}{10} & \frac{1}{10} & \frac{1}{2} \end{bmatrix} \right)^{-1} \times \underline{1}$$

$$= \begin{bmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ -\frac{1}{10} & 0 & 1 & -\frac{9}{10} \\ -\frac{1}{10} & -\frac{1}{10} & -\frac{1}{10} & \frac{1}{2} \end{bmatrix}^{-1} \times \underline{1}$$

$$= \begin{bmatrix} \frac{16}{9} & \frac{41}{18} & \frac{25}{9} & 5 \\ \frac{7}{9} & \frac{41}{18} & \frac{25}{9} & 5 \\ \frac{7}{9} & \frac{23}{18} & \frac{25}{9} & 5 \\ \frac{9}{5} & \frac{18}{5} & \frac{7}{6} & 5 \end{bmatrix} \times \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$$

$$= \begin{bmatrix} \frac{71}{6} \\ \frac{65}{6} \\ \frac{59}{6} \\ \frac{17}{2} \end{bmatrix}.$$

Thus, the requested expected value is  $E(\tau | X_1 = 1) = \frac{16}{9} + \frac{41}{18} + \frac{25}{9} + 5 = \frac{71}{6} = 11.8(3)$ .

- (d) Suppose being in states 1, 2, 3, 4, 5 cost 250, 300, 200, 150, 100 euro (respectively). (2.0)

Assume a discount factor  $\alpha = \frac{5}{6}$  and compute the expected total discounted expenditure of a patient starting in the coronary care unit.

**Hint:** Check the footnote.

• **Vector of costs**

$$\underline{c} = [c(j)]_{j \in \mathcal{S}}$$

$$= \begin{bmatrix} 250 \\ 300 \\ 200 \\ 150 \\ 100 \end{bmatrix}$$

• **Discount factor**

$\alpha = \frac{5}{6}$

• **Vector of the expected total discounted expenditures**

The expected total discounted cost incurred, starting at state  $i$ , is equal to  $\phi(i) = E[\sum_{n=0}^{+\infty} \alpha^n c(X_n) | X_0 = i]$ . Equivalently,

$$\underline{\phi} = [\phi(i)]_{i \in \mathcal{S}} = (\mathbf{I} - \alpha \mathbf{P})^{-1} \times \underline{c}.$$

We get

$$\begin{aligned} \underline{\phi} &= (\mathbf{I} - \alpha \mathbf{P})^{-1} \times \underline{c} \\ &= \left( \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} - \frac{5}{6} \times \begin{bmatrix} 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ \frac{1}{10} & 0 & 0 & \frac{9}{10} & 0 \\ \frac{1}{10} & \frac{1}{10} & \frac{1}{10} & \frac{1}{2} & \frac{1}{5} \\ 0 & 0 & 0 & 0 & 0 \end{bmatrix} \right)^{-1} \times \underline{c} \\ &= \begin{bmatrix} 1 & -\frac{5}{6} & 0 & 0 & 0 \\ 0 & 1 & -\frac{5}{6} & 0 & 0 \\ -\frac{1}{12} & 0 & 1 & -\frac{3}{4} & 0 \\ -\frac{1}{12} & -\frac{1}{12} & -\frac{1}{12} & \frac{7}{4} & -\frac{1}{6} \\ 0 & 0 & 0 & 0 & \frac{1}{6} \end{bmatrix}^{-1} \times \underline{c} \\ &= \begin{bmatrix} \frac{243}{203} & \frac{225}{203} & \frac{30}{29} & \frac{270}{203} & \frac{270}{203} \\ \frac{48}{203} & \frac{270}{203} & \frac{36}{29} & \frac{324}{203} & \frac{324}{203} \\ \frac{288}{1015} & \frac{402}{1015} & \frac{216}{145} & \frac{1944}{1015} & \frac{1944}{1015} \\ \frac{249}{1015} & \frac{411}{1015} & \frac{78}{145} & \frac{2442}{1015} & \frac{2442}{1015} \\ 0 & 0 & 0 & 0 & 6 \end{bmatrix} \times \begin{bmatrix} 250 \\ 300 \\ 200 \\ 150 \\ 100 \end{bmatrix} \end{aligned}$$

• Requested expected total discounted expenditure (customer starting with brand 1)

$$\begin{aligned} \phi(1) &= \text{1st. row of } (\mathbf{I} - \alpha \mathbf{P})^{-1} \times \underline{c} \\ &= \begin{bmatrix} \frac{243}{203} & \frac{225}{203} & \frac{30}{29} & \frac{270}{203} & \frac{270}{203} \end{bmatrix} \times \begin{bmatrix} 250 \\ 300 \\ 200 \\ 150 \\ 100 \end{bmatrix} \\ &= \frac{237750}{203} \\ &\approx 1171.18. \end{aligned}$$

2. Let  $\{X_n : n \in \mathbb{N}_0\}$  be a branching process such that: the number of offspring per individual has a binomial(2,  $p$ ) distribution, where  $p \in (\frac{1}{2}, 1]$ ; the number of initial individuals is a r.v.  $X_0 \sim \text{geometric}(p)$ . Derive the extinction probability of this branching process. (2.5)

• Simple branching process

$$\{X_n : n \in \mathbb{N}_0\}$$

$$X_0 = 1$$

$$X_n = \text{size of generation } n$$

$$X_n = \sum_{i=1}^{X_{n-1}} Z_i, \quad n \in \mathbb{N}$$

• Number of offspring per individual and its p.g.f.

$$Z_l \equiv Z_{l,n} = \text{number of offspring of the } l^{\text{th}} \text{ individual of generation } n$$

$$Z_l \stackrel{i.i.d.}{Z} \sim \text{binomial}(2, p), \quad l \in \mathbb{N}$$

$$E(Z) \stackrel{form.}{=} 2p > 1, \text{ when } p \in (\frac{1}{2}, 1]$$

$$P_Z(s) = E(s^Z) \stackrel{form.}{=} (1 - p + ps)^2$$

• Probability of extinction (with one single initial individual)

$$E(Z) > 1, \text{ therefore the probability of extinction, } \pi \stackrel{form.}{=} \lim_{n \rightarrow +\infty} P(X_n = 0), \text{ is the smallest positive number satisfying}$$

$$\begin{aligned} s &\stackrel{form.}{=} P_Z(s) \\ s &= (1 - p + ps)^2 \\ p^2 s^2 - 2p^2 s + p^2 + 2ps - 2p + 1 - s &= 0 \\ p^2 s^2 - (2p^2 - 2p + 1)s + (p^2 - 2p + 1) &= 0 \\ s &= \frac{(2p^2 - 2p + 1) \pm \sqrt{(2p^2 - 2p + 1)^2 - 4p^2(1 - p)^2}}{p^2} \\ s &= \dots \\ s &= \frac{(2p^2 - 2p + 1) \pm \sqrt{1 - 4p - 4p^2}}{p^2} \\ s &= \frac{(2p^2 - 2p + 1) \pm \sqrt{(1 - 2p)^2}}{p^2} \\ s &= \frac{(2p^2 - 2p + 1) \pm (1 - 2p)}{p^2} \\ s &= 1 \text{ or } \frac{(1 - p)^2}{p^2}. \end{aligned}$$

Since  $p \in (\frac{1}{2}, 1]$ , the smallest positive root is  $\pi = \frac{(1-p)^2}{p^2}$ .

• New initial state

$$X_0 \sim \text{geometric}(p), \quad p \in (\frac{1}{2}, 1]$$

$$P_{X_0}(s) = E(s^{X_0}) \stackrel{form.}{=} \frac{ps}{1 - (1-p)s}$$

• New probability of extinction

Using the total probability law, the fact that the offspring are produced independently and  $X_0 \sim Z$ , we get

$$\begin{aligned} P(\text{extinction}) &= \sum_{j=0}^{+\infty} P(\text{extinction} \mid X_0 = j) \times P(X_0 = j) \\ &= \sum_{j=0}^{+\infty} \pi^j \times P(X_0 = j) \\ &= P_{X_0}(\pi) \\ &= \frac{p\pi}{1 - (1-p)\pi} \\ &= \frac{p \frac{(1-p)^2}{p^2}}{1 - (1-p) \frac{(1-p)^2}{p^2}} \\ &= \frac{(p-1)^2 p}{p^3 - 2p^2 + 3p - 1}. \end{aligned}$$

Group 6 — Continuous time Markov chains

8.0 points

1. Admit that a company uses  $m$  ( $m \in \mathbb{N}$ ) robots. Furthermore: each robot breaks down after an exponentially distributed time with parameter  $\lambda$ ; the company has  $m$  repair people to do service when robots fail (one repair person per robot); the repair time for each robot is exponentially distributed with mean  $\mu^{-1}$ .

Let  $X(t)$  represent the number of robots being repaired at time  $t$  and admit that  $\{X(t) : t \geq 0\}$  is a CTMC.

- (a) Identify the state space and the infinitesimal generator  $\mathbf{R}$  of this birth and death process. (1.5)  
Draw the associated rate diagram.

- **CTMC**

$$\{X(t) : t \geq 0\}$$

$X(t)$  = number of robots being repaired at time  $t$

- **State space**

$$\mathcal{S} = \{0, 1, 2, \dots, m\}$$

- **Birth/death rates**

We can interpret a break down of a robot as an arrival (to the repair shop) or a *birth*. Also note that as soon as a reparation of a robot is concluded, we can say that a departure (from the repair shop) or *death* has occurred. Moreover, there are  $m$  repair people or servers.

[We are dealing with an  $M/M/m$  queueing system with a finite customer population, namely  $m$ .]

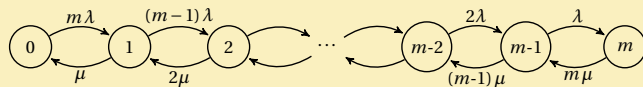
Thus,

$$\lambda_i = (m-i)\lambda, \quad i = 0, 1, \dots, m-1$$

$$\mu_i = i\mu, \quad i = 1, 2, \dots, m.$$

- **Rate diagram**

[Recall that the rate diagram of a CTMC is a directed graph — with no loops — in which each state is represented by a node and there is an arc going from node  $i$  to node  $j$  (if  $q_{ij} > 0$ ) with  $q_{ij}$  written on it. These rates coincide with the birth and death rates...]



- **Infinitesimal generator**

This matrix has entries

$$r_{ij} = \begin{cases} q_{ij}, & i \neq j \\ -v_i = -\sum_{m \in \mathcal{S}} q_{im}, & j = i \end{cases}$$

and in this case  $\mathbf{R} = [r_{ij}]_{i,j \in \mathcal{S}}$  is equal to

$$\begin{bmatrix} -m\lambda & m\lambda & 0 & 0 & 0 & 0 & 0 \\ \mu & -[(m-1)\lambda + \mu] & (m-1)\lambda & 0 & 0 & 0 & 0 \\ 0 & 2\mu & -[(m-1)\lambda + 2\mu] & (m-2)\lambda & 0 & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & (m-2)\mu & -[2\lambda + (m-2)\mu] & 2\lambda & 0 \\ 0 & 0 & 0 & 0 & (m-1)\mu & -[\lambda + (m-1)\mu] & \lambda \\ 0 & 0 & 0 & 0 & 0 & m\mu & -m\mu \end{bmatrix}$$

(b) Obtain the equilibrium probabilities  $P_j = \lim_{t \rightarrow +\infty} P[X(t) = j | X(0) = 0]$ . (2.5)

Show that  $L_s$ , the number of robots being repaired upon arrival to the repair shop in the long-run, has a binomial distribution and compute its expected value.

**Hint:**  $(a+b)^m = \sum_{n=0}^m \binom{m}{n} a^n b^{m-n}$ , for  $m \in \mathbb{N}$ .

- **Performance measure**

$L_s$  = number of robots being repaired upon arrival to the repair shop in the long-run

- **Equilibrium probabilities**

Since this CTMC has a finite state space, we only need to deal with  $\rho = \frac{\lambda}{\mu} < +\infty$  to guarantee the existence of equilibrium probabilities  $P_j = \lim_{t \rightarrow +\infty} P_j(t) = P(L_s = j)$ .

$$P_0 = \left[ 1 + \sum_{n=1}^{+\infty} \left( \prod_{i=0}^{n-1} \frac{\lambda_i}{\mu_{i+1}} \right) \right]^{-1}$$

$$\begin{aligned} P_0 &= \left[ 1 + \sum_{n=1}^{+\infty} \left( \prod_{i=0}^{n-1} \frac{(m-i)\lambda_i}{(i+1)\mu} \right) \right]^{-1} \\ &= \left[ 1 + \sum_{n=1}^m \frac{m \times (m-1) \times \dots \times 2 \times 1}{1 \times 2 \times \dots \times (m-1) \times m} \left( \frac{\lambda}{\mu} \right)^n \right]^{-1} \\ &= \left[ 1 + \sum_{n=1}^m \binom{m}{n} \left( \frac{\lambda}{\mu} \right)^n \right]^{-1} \\ &= \left[ \sum_{n=0}^m \binom{m}{n} \left( \frac{\lambda}{\mu} \right)^n \times 1^{m-n} \right]^{-1} \\ &= \frac{1}{\left( 1 + \frac{\lambda}{\mu} \right)^m} \\ &= \left( \frac{\mu}{\lambda + \mu} \right)^m \end{aligned}$$

$$\begin{aligned} P_j &= P_0 \times \prod_{i=0}^{j-1} \frac{\lambda_i}{\mu_{i+1}} \\ &= P_0 \times \prod_{i=0}^{j-1} \frac{(m-i)\lambda_i}{(i+1)\mu} \\ &= \frac{\binom{m}{j} \left( \frac{\lambda}{\mu} \right)^j}{\left( 1 + \frac{\lambda}{\mu} \right)^m} \\ &= \binom{m}{j} \left( \frac{\frac{\lambda}{\mu}}{1 + \frac{\lambda}{\mu}} \right)^j \left( \frac{1}{1 + \frac{\lambda}{\mu}} \right)^{m-j} \\ &= \binom{m}{j} \left( \frac{\lambda}{\lambda + \mu} \right)^j \left( \frac{\mu}{\lambda + \mu} \right)^{m-j}, \quad j = 1, 2, \dots, m, \end{aligned}$$

i.e.,  $P(L_s = j) = \binom{m}{j} \left( \frac{\lambda}{\lambda + \mu} \right)^j \left( \frac{\mu}{\lambda + \mu} \right)^{m-j}$ ,  $j = 0, 1, 2, \dots, m$ , hence

$$L_s \sim \text{binomial} \left( m, \frac{\lambda}{\lambda + \mu} \right).$$

- **Requested expected value**

Since  $L_s \sim \text{binomial} \left( m, \frac{\lambda}{\lambda + \mu} \right)$ , we get

$$E(L_s) \stackrel{\text{form.}}{=} m \times \frac{\lambda}{\lambda + \mu}.$$

(c) This repair shop can be modelled as a  $M/M/m$  queueing system with a finite customer population (1.0)

of size  $m$ . Let  $\lambda_e = \sum_{j=0}^m \lambda_j P_j$  be the rate of entering customers in the repair shop, in the long-run.

Use Little's law to confirm that the expected value of the time a machine spends in the repair shop is equal to  $\frac{1}{\mu}$ , in the long-run.

- **Important**

Let  $W_s$  be the time a machine spends in the repair shop, in the long-run.

According to Little's law,  $E(L_s) = \lambda_e \times E(W_s)$ .

- **Rate of entering customers**

$$\begin{aligned} \lambda_e &= \sum_{j=0}^m \lambda_j \times P_j \\ &= \sum_{j=0}^m (m-j)\lambda \times P_j \\ &= m\lambda - \lambda E(L_s) \end{aligned}$$



$$\lambda_e \stackrel{(b)}{=} m\lambda - \lambda \frac{m\lambda}{\lambda + \mu}$$

$$= \frac{m\lambda\mu}{\lambda + \mu}$$

• Requested expected value

$$E(W_s) = \frac{E(L_s)}{\lambda_e}$$

$$= \frac{\frac{m\lambda}{\lambda + \mu}}{\frac{m\lambda\mu}{\lambda + \mu}}$$

$$= \frac{1}{\mu} \quad \checkmark$$

2. Customers arrive to a laundromat according to a Poisson process with rate equal to 8 customers per hour. Assume the laundromat has five machines and customers are only allowed to use one machine at a time. The time a customer uses a machine is exponentially distributed with a mean of one hour. Assume that customers do not enter the laundromat if all five machines are taken.

(a) After having chosen a suitable queueing system, calculate the fraction of arriving customers who find the laundromat full and are lost, in the long-run. (2.0)

• Birth-death queueing system

$M/M/m/m$   
 $m = 5$

• State space

$\mathcal{S} = \{0, 1, \dots, m\}$

• Birth/death rates

$\lambda_k = \lambda = 8, \quad k \in \{0, 1, \dots, m-1\}$   
 $\mu_k = k\mu = k, \quad k \in \{1, 2, \dots, m\}$

• Traffic intensity/ergodicity condition

$\rho = \frac{\lambda}{m\mu} = \frac{8}{5 \times 1} = 8/5 < +\infty$

• Performance measure (in the long-run)

$L_s$  = number of customers an arriving customer sees to the laundromat

$$P(L_s = k) = \begin{cases} \frac{\frac{(m\rho)^k}{k!}}{\sum_{j=0}^m \frac{(m\rho)^j}{j!}}, & k = 0, 1, \dots, m \\ 0, & k = m+1, m+2, \dots \end{cases}$$

• Requested fraction

$$P(L_s = m) = \frac{B(m, m\rho)}{\sum_{j=0}^m \frac{(m\rho)^j}{j!}}$$

$$= \frac{\frac{(m\rho)^m}{m!}}{\sum_{j=0}^m \frac{(m\rho)^j}{j!}}$$

$$\stackrel{m=5, \lambda=8, \mu=1}{=} \frac{\frac{8^5}{5!}}{\sum_{j=0}^5 \frac{8^j}{j!}}$$

$$= \frac{8^5/5!}{1 + 8 + 8^2/2! + 8^3/3! + 8^4/4! + 8^5/5!}$$

$$= \frac{\frac{32768}{120}}{1 + 8 + 32 + \frac{512}{6} + \frac{4096}{24} + \frac{32768}{120}}$$

$$= \frac{4096}{8551}$$

$$\approx 0.479008.$$

(b) On average, how many servers are idle, in the long-run?

(1.0)

• Requested average

Since 5, 4, 3, 2, 1, 0 machines are idle if and only if  $L_s = 0, 1, 2, 4, 5$ , respectively.

$$\sum_{j=0}^m (m-j) \times P_j = m \times \sum_{j=0}^m P_j - \sum_{j=0}^m j \times P_j$$

$$= m \times 1 - \sum_{j=0}^m j \times P_j$$

$$= m - E(L_s)$$

$$\stackrel{form.}{=} m - m\rho [1 - B(m, m\rho)]$$

$$\stackrel{(a)}{\approx} 5 - \frac{8}{1} \times (1 - 0.479008)$$

$$\approx 0.832064.$$