

Problem 1.

(a) Let $U_1 = T_1, U_2 = T_2 - T_1, U_3 = T_3 - T_2$. Since $U_i \sim \exp(2)$,

$$\begin{aligned} P(T_1 + T_2 < T_3) &= P(2U_1 + U_2 < U_1 + U_2 + U_3) = P(U_1 < U_3) \\ &= \int_0^\infty \int_0^t 2e^{-2u} du 2e^{-2t} dt = \int_0^\infty \left(-e^{-2u} \Big|_0^t \right) 2e^{-2t} dt \\ &= 2 \int_0^\infty (e^{-2t} - e^{-4t}) dt = \left(-\frac{1}{2}e^{-2t} + \frac{1}{4}e^{-4t} \right) \Big|_0^\infty = \frac{1}{2} \end{aligned}$$

(b) $E N^2(S) = E [E N^2(S) | S] = E [2S + 4S^2] = 2 \cdot \frac{1}{2} + 4 \cdot \frac{1}{3} = \frac{7}{3}$

(c) $P(X_{n+1} = j | X_n = i, X_{n-1} = i_{n-1}, \dots, X_0 = i_0) = P(N((n+1)^2) - N(n^2))$

$$= P(X_{n+1} = j | X_n = i) = \frac{2(2n+1)^{j-i}}{(j-i)!} e^{-2(2n+1)} \mathbb{1}_{(j \geq i \geq i_{n-1} \geq \dots \geq i_0)}$$

The RVs $(X_n)_n$ form a first-order homogeneous Markov chain.

Problem 2.

(a) By the Cesàro means theorem, for any $\omega \in \Omega$ such that

$$\lim_{n \rightarrow \infty} X_n = 0, \text{ also } \lim_{n \rightarrow \infty} \frac{1}{n} S_n = 0; \text{ thus } P\left(\frac{1}{n} S_n \rightarrow 0\right) = 1.$$

(b) Consider a sequence $X_n(\omega) \rightarrow 0$; suppose that $|X_n(\omega)| < \epsilon$ for all $n \geq N = N(\omega)$. Then, using the estimate of partial sums of the harmonic series,

$$\left| \frac{1}{\log n} \sum_{k=1}^n \frac{X_k(\omega)}{k} \right| \leq \epsilon \left| \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \right| \leq \epsilon \frac{\log n + 1}{\log n} \leq 2\epsilon$$

and thus, $\left| \frac{1}{\log n} \sum_{k=1}^n \frac{X_k(\omega)}{k} \right| \rightarrow 0$.

(c) As above in part (b)

$$\sum_{k=2}^n \frac{1}{k \log k} \leq \int_3^{n+1} \frac{dx}{x \log x} = \int_{\log 3}^{\log(n+1)} \frac{du}{u} = \log \log(n+1)$$

The remainder of the proof is the same as in Part (b).

Note: In the problem statement the sum starts at $k=1$, but for $k=1$ the summation term is not well defined. Take instead $k=2$.

(d) To show this take $X_n \sim \text{Bernoulli}$

$$P(X_n=0) = 1 - \frac{1}{n}; \quad P(X_n=2^n) = \frac{1}{n}, \quad n \geq 1$$

Then $X_n \xrightarrow{p} 0$, but

$$P\left(\frac{1}{n} \sum_{k=1}^n X_k < \varepsilon\right) \geq P\left(X_{\lfloor \frac{n}{2} \rfloor} = X_{\lfloor \frac{n}{2} \rfloor + 1} = \dots = X_n = 0\right) \geq \left(1 - \frac{1}{n}\right)^{n/2} \not\rightarrow 0$$

(e) Assuming that $\text{Var}(X) < \infty$, we use SLLN to claim that

$$\frac{X_1 + \dots + X_n}{n} \xrightarrow{\text{a.s.}} EX; \quad \frac{X_1^2 + \dots + X_n^2}{n} \xrightarrow{\text{a.s.}} EX^2$$

Since the intersection of two events of prob. 1 has probability 1,

$$\frac{X_1 + \dots + X_n}{X_1^2 + \dots + X_n^2} \xrightarrow{\text{a.s.}} \frac{EX}{EX^2}$$

Problem 3.

$$(a) \quad E|X_n| = EX_n = (EY)^n = 1.$$

$$E(X_{n+1} | X_1^n) = E(X_n Y_{n+1} | X_1^n) = X_n$$

which proves that $(X_n)_n$ form a martingale with respect to the natural filtration.

(b) $(X_n, \mathcal{F}_n)_n$ forms an L_1 -bounded martingale; thus there is an RV X_∞ on $(\Omega, \mathcal{F}_\infty)$ s.t. $X_n \xrightarrow{a.s.} X_\infty$.

$$\text{To find the limit, consider } \log X_n = \sum_{k=1}^n \log Y_k$$

$$\text{We have } \frac{\log X_n}{n} = \frac{\sum \log Y_k}{n} \xrightarrow{a.s.} E \log Y_k = \frac{1}{2} \left(\log \frac{1}{4} \cdot \frac{3}{4} \right) = \frac{1}{2} \log \frac{3}{16} < 0.$$

$$\text{Thus } P \left(\lim_{n \rightarrow \infty} \frac{\log X_n}{n} = \log \frac{\sqrt{3}}{4} \right) = 1$$

$$\text{or } \log X_n \xrightarrow{a.s.} -\infty, \text{ or } X_n \xrightarrow{a.s.} X_\infty = 0$$

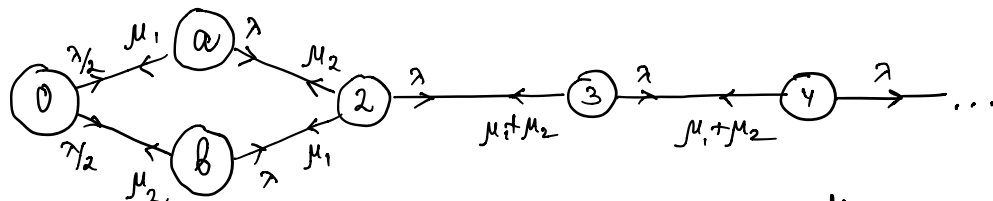
At the same time, $X_n \not\xrightarrow{L_1} X_\infty$, i.e. $E|X_n - 0| \not\rightarrow 0$

so the sequence does not converge in L_1 .

A necessary condition for L_1 convergence is $E(X_\infty | X_1^n) = X_n$, but it is clearly not satisfied.

Problem 4.

(a) The Markov chain can be graphically represented as follows:



where on the edges we write exponential transition rates.

The generator matrix Q has the form

$$Q = \begin{matrix} & \begin{matrix} 0 & a & b & 2 & 3 & 4 & 5 \end{matrix} \\ \begin{matrix} 0 \\ a \\ b \\ 2 \\ 3 \\ 4 \end{matrix} & \begin{bmatrix} -\lambda & \gamma/2 & \gamma/2 & 0 & 0 & 0 & 0 & \dots \\ \mu_1 & -\mu_1 - \lambda & 0 & \lambda & 0 & 0 & 0 & \dots \\ \mu_2 & 0 & -\mu_2 - \lambda & \lambda & 0 & 0 & 0 & \dots \\ 0 & \mu_2 & \mu_1 & -\lambda - \mu_1 - \mu_2 & \lambda & 0 & 0 & \dots \\ 0 & 0 & 0 & \mu_1 + \mu_2 & -\lambda - \mu_1 - \mu_2 & \lambda & 0 & \dots \\ 0 & 0 & 0 & 0 & \mu_1 + \mu_2 & -\lambda - \mu_1 - \mu_2 & \lambda & \dots \\ \dots & \dots & \dots & \dots & \dots & \dots & \dots & \dots \end{bmatrix} \end{matrix}$$

where q_{ij} represents the exponential rate of moving $i \rightarrow j$.

Write $\pi Q = 0$ in detail. The 3 first relations are:

$$(1) \begin{cases} \pi_0 \lambda = \mu_1 \pi_a + \mu_2 \pi_b \\ \pi_0 \frac{\lambda}{2} - (\mu_1 + \lambda) \pi_a + \mu_2 \pi_2 = 0 \\ \pi_0 \frac{\lambda}{2} - (\mu_2 + \lambda) \pi_b + \mu_1 \pi_2 = 0 \end{cases}$$

Eliminating π_b and π_2 , we obtain

$$0 = \lambda \pi_0 - \lambda \pi_a - \mu_1 \pi_a + \mu_2 \pi_2 - \mu_2 \pi_b - \lambda \pi_b + \mu_1 \pi_2$$
$$= -\lambda(\pi_a + \pi_b) + (\mu_1 + \mu_2)\pi_2$$

$$\text{i.e., } \pi_2 = \frac{\lambda}{\mu_1 + \mu_2}(\pi_a + \pi_b) = \frac{\lambda}{\mu_1 + \mu_2} \left(\pi_a + \frac{\lambda \pi_0 - \mu_1 \pi_a}{\mu_2} \right)$$

Substitute this into the second equation in (1). Simplifying we obtain

$$(2) \quad \pi_0 \frac{\lambda}{2} = \pi_a \mu_1, \quad \text{i.e., } \pi_0 Q_{0a} = \pi_a Q_{a0}$$

Now substitute $\pi_a = \pi_0 \frac{\lambda}{2\mu_1}$ into the first equation in (1):

$$(3) \quad \pi_0 \lambda = \pi_0 \frac{\lambda}{2} + \mu_2 \pi_b, \quad \text{i.e., } \pi_0 Q_{0b} = \pi_b Q_{b0}$$

Eliminating π_0 from the second and third equations in (1),

we obtain

$$(4) \quad \begin{cases} \pi_a Q_{a2} = \pi_2 Q_{2a} \\ \pi_b Q_{b2} = \pi_2 Q_{2b} \end{cases}$$

Finally let us states $i \geq 2$. The i^{th} relation gives

$$\pi_a \lambda + \pi_b \lambda - \pi_2 (\lambda + \mu_1 + \mu_2) + \pi_3 (\mu_1 + \mu_2) = 0;$$

substituting π_a and π_b from (2)-(3), we obtain

$$\pi_2 \lambda = \pi_3 (\mu_1 + \mu_2)$$

and generally

$$(5) \quad \pi_i Q_{i,i+1} = \pi_{i+1} Q_{i+1,i}, \quad i \geq 2$$

This shows that the chain is reversible.

(b) The chain is irreducible, so its stationary distribution π is at the same time the limiting distribution.

The stationary distribution is easily computed from Eqs. (2)-(5) in part (a). Specifically,

$$(b) \begin{cases} \pi_a = \frac{\lambda}{2\mu_1} \pi_0 ; \pi_b = \frac{\lambda}{2\mu_2} \pi_0 \\ \pi_2 = \frac{\lambda^2}{2(\mu_1 + \mu_2)} \pi_0 \\ \frac{\pi_i}{\pi_{i+1}} = \frac{\lambda}{\mu_1 + \mu_2}, \quad i \geq 2 \end{cases}$$

Denote $x := \frac{\lambda}{\mu_1 + \mu_2}$ and assume that $x < 1$ (if not, then the queue increases without limit, and there is no stationarity).

Normalization condition gives

$$\begin{aligned} \pi_0 \left(1 + \frac{\lambda}{2\mu_1} + \frac{\lambda}{2\mu_2} + \frac{\lambda^2}{2(\mu_1 + \mu_2)} \right) + \pi_2 \sum_{i \geq 1} x^i \\ = \pi_0 \left(1 + \frac{\lambda}{2\mu_1} + \frac{\lambda}{2\mu_2} + \frac{\lambda^2}{2(\mu_1 + \mu_2)} \frac{1}{1-x} \right) = \pi_0 \left(1 + \frac{\lambda}{2\mu_1} + \frac{\lambda}{2\mu_2} + \frac{\lambda^2}{2(\mu_1 + \mu_2 - \lambda)} \right) \\ = 1 \end{aligned}$$

This yields the value of π_0 , and the remaining components of the distribution are found from (6) above.

Problem 5.

(a) Let us compute $EX(t)$:

$$EX(t) = EY^3 t = t \int_{-1}^1 y^3 \frac{dy}{2} = 0.$$

The auto-covariance function is found as

$$E(X(t)X(t+s)) = t(t+s) EY^6 = \frac{1}{7} t(t+s).$$

This depends on t , so $X(t)$ is not even WSS, let alone stationary.

(b) Take t_1, t_2, t_3, t_4 to satisfy $0 < t_1 < t_2 < t_3 < t_4$, and let

$$Z_1 = X(t_2) - X(t_1)$$

$$Z_2 = X(t_4) - X(t_3)$$

$$EZ_1 = Ee^{\gamma t_2} - Ee^{\gamma t_1} = (t_2 - t_1) \int_0^1 e^y dy = (t_2 - t_1)(e - 1).$$

$$EZ_2 = (t_4 - t_3)(e - 1)$$

$$E(Z_1 Z_2) = E((t_2 - t_1)(t_4 - t_3)e^{2\gamma}) = (t_2 - t_1)(t_4 - t_3) \frac{e^2 - 1}{2}$$

$$\neq EZ_1 EZ_2$$

and thus the increments are correlated, so not independent.

Now consider the distribution of $W(t_1, t_2, s) = X(t_2 + s) - X(t_1 + s)$:

$$F_w(x) = P(X(t_2 + s) - X(t_1 + s) < x) = P(e^{\gamma}(t_2 - t_1) < x)$$

This probability does not depend on s , so the increments are stationary.