

Problem 1. [Renewal equation]

Let X_1, X_2, \dots be a sequence of independent $\text{Uniform}(0, 1)$ random variables, $t \in (0, 1]$, and

$$N(t) := \min \left\{ n \in \mathbb{N} : \sum_{k=1}^n X_k > t \right\}$$

be the smallest number of X 's that need to be added so that their sum exceeds t , as shown in Figure 1 (where $N(t) = 4$). Define $M(t) := \mathbb{E}[N(t)]$. From the definition, it is clear that $N(t)$ is a random variable taking values in \mathbb{N} .

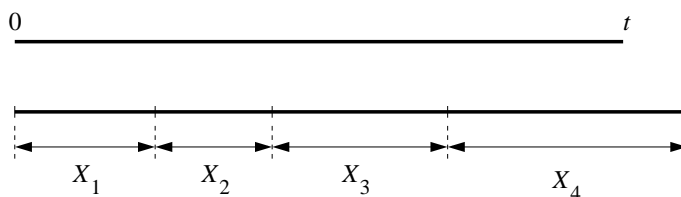


Figure 1: Covering the interval $[0, t]$ with lines with random lengths X_m .

- (a) Explain why, for any $x_1 \in (0, 1]$,

$$\mathbb{E}[N(t)|X_1 = x_1] = \begin{cases} 1 & \text{for } t < x_1, \\ 1 + M(t - x_1) & \text{for } x_1 \leq t. \end{cases}$$

Hint: Condition on X_1 . Please explain your reasoning (no calculations are needed)!

- (b) Show that $M(t)$ satisfies the integral equation

$$M(t) = 1 + \int_0^t M(t - x_1) dx_1. \quad (1)$$

Hint: Use the Tower Rule by conditioning on X_1 . Do not forget that $t \in (0, 1]$.

- (c) Solve the integral equation for $M(t)$ by first converting it to a differential equation. First change variables in the integral to rewrite (1) in the form $M(t) = 1 + \int_0^t M(y) dy$ (write explicitly the change of variables). Differentiate both sides of this equation to derive a *differential* equation for $M(t)$. What is the initial condition for $M(0)$ that the function $M(t)$ should satisfy (justify your answer with a couple of sentences). Solve the initial value problem for $M(t)$ you have obtained.
- (d) Now solve the integral equation for $M(t)$ by using Laplace transform. You can use the following facts about Laplace transform, without proving them:

- if $h(x) = x^n$, for $n \in \mathbb{N}$, then $\mathcal{L}[h](\xi) = \frac{n!}{\xi^{n+1}}$, $\xi > 0$;
- if $h(x) = e^{ax}$ (for any $a \in \mathbb{R}$), then $\mathcal{L}[h](\xi) = \frac{1}{\xi - a}$, $\xi > a$;
- if $A(t) = \int_0^t M(t - x_1) f(x_1) dx_1$, then $\mathcal{L}[A](\xi) = \mathcal{L}[M](\xi) \mathcal{L}[f](\xi)$.

Problem 2. [Autocorrelation and autocovariance functions of a Poisson process]

Let $N = \{N_t : t \geq 0\}$ be a Poisson process with a constant rate λ .

- (a) Compute the expected value $m_N(t) = \mathbb{E}[N_t]$.

Hint: Use that N_t is a Poisson distributed random variable with parameter λt . Recall that if $X \sim \text{Poisson}(\alpha)$, then $\mathbb{E}[X] = \alpha$ and $\text{Var } X = \alpha$.

- (b) Use your result in (a) and the basic properties of Poisson processes to find the autocorrelation function, $R_N(s, t) = \mathbb{E}[N_s N_t]$, of the Poisson process N .

Hint: First assume that $0 \leq s \leq t$ and compute $R_N(s, t)$; then think how your result will change if $0 \leq t \leq s$, and write your result for $R_N(s, t)$ in the general case.

- (c) Compute the autocovariance function,

$$C_N(s, t) = \mathbb{E}[(N_s - \mathbb{E}N_s)(N_t - \mathbb{E}N_t)] = R_N(s, t) - m_N(s)m_N(t) .$$

Look at $C_N(t, t)$ and at the hint to part (a) – are the results for $C_N(t, t)$ and $\text{Var } X_t$ consistent?

- (d) Compute the correlation coefficient, $\rho_N(s, t) = \frac{C_N(s, t)}{\sqrt{C_N(s, s)C_N(t, t)}}$. From your result, find $\rho_N(s, t)$ for $t \geq s$ and find $\lim_{t \rightarrow s^+} \rho_N(s, t)$. Is this what you expected? Why?

- (e) Is N a wide-sense stationary process? Explain briefly.

- (f) Is N a strong-sense stationary process? Why?

Problem 3. [P.m.f. of order k and conditional transition function of a Markov process]

Let $X = \{X_t : t \geq 0\}$ be a continuous-time Markovian process with state space $\mathbb{Z}_+ = \{0, 1, 2, 3, \dots\}$. Let $N = \{N_t : t \geq 0\}$ be a Poisson process with a constant rate λ .

Recall that the *probability mass function (p.m.f.) of order k* for the process X is defined as the joint p.m.f. of the random vector $(X_{t_1}, \dots, X_{t_k})$

$$p^{(k)}(i_1, \dots, i_k; t_1, \dots, t_k) := p_{(X_{t_1}, \dots, X_{t_k})}(i_1, \dots, i_k) = \mathbb{P}(X_{t_1} = i_1, \dots, X_{t_k} = i_k) ,$$

where $t_m \geq 0$ and $i_m \in \mathbb{Z}_+$ for $m = 1, \dots, k$ (see Definition 2.1.4 on page 49 of Lefebvre's book, where the notation is slightly different).

The *conditional transition function* of the process X is defined as the conditional p.m.f. of X at a moment t conditioned on an earlier moment t_0 :

$$p(i, i_0; t, t_0) := p_{X_t|X_{t_0}}(i|i_0) = \mathbb{P}(X_t = i | X_{t_0} = i_0) = \frac{\mathbb{P}(X_t = i, X_{t_0} = i_0)}{\mathbb{P}(X_{t_0} = i_0)}, \quad t_0 < t$$

(compare this with its continuous analogue, defined in Eqn. (2.49) on page 62 of Lefebvre's book). Throughout this problem, t_0, t_1 , and t_2 are times satisfying

$$0 \leq t_0 \leq t_1 \leq t_2.$$

- (a) Use some basic property of probability to explain why

$$\sum_{i_1} \mathbb{P}(X_{t_2} = i_2, X_{t_1} = i_1, X_{t_0} = i_0) = \mathbb{P}(X_{t_2} = i_2, X_{t_0} = i_0).$$

- (b) Use your result from part (a) to show that the conditional transition function of the process X satisfies

$$\sum_{i_1} p(i_2, i_1; t_2, t_1) p(i_1, i_0; t_1, t_0) = p(i_2, i_0; t_2, t_0).$$

This equation has a name – what is it?

Hint: Write

$$\sum_{i_1} p(i_2, i_1; t_2, t_1) p(i_1, i_0; t_1, t_0) = \mathbb{P}(X_{t_2} = i_2 | X_{t_1} = i_1) \mathbb{P}(X_{t_1} = i_1 | X_{t_0} = i_0),$$

then explain why $\mathbb{P}(X_{t_2} = i_2 | X_{t_1} = i_1) = \mathbb{P}(X_{t_2} = i_2 | X_{t_1} = i_1, X_{t_0} = i_0)$ (it relies on one property of the process X – which one?), and use the definition of conditional probability to derive the desired equality. The calculations are really simple (look for a nice cancellation).

- (c) Prove that

$$\sum_{i_0} p(i_1, i_0; t_1, t_0) p^{(1)}(i_0; t_0) = p^{(1)}(i_1; t_1).$$

- (d) Explain in words the intuitive meaning of the identity

$$\sum_{i_0} p^{(2)}(i_0, i_1; t_0, t_1) = p^{(1)}(i_1; t_1), \quad t_0 < t_1.$$

- (e) Explain in words the intuitive meaning of the identity

$$\sum_{i_1} p^{(3)}(i_0, i_1, i_2; t_0, t_1, t_2) = p^{(2)}(i_0, i_2; t_0, t_2), \quad t_0 < t_1 < t_2.$$

- (f) In the rest of the problem you will consider the particular case of the Poisson process N with rate $\lambda = \text{const}$. What is $p^{(1)}(i_0; t_0) = \mathbb{P}(N_{t_0} = i_0)$? You can write the answer very simply; use whatever you know about the Poisson process.

- (g) What is $p^{(2)}(i_0, i_1; t_0, t_1)$ for $0 \leq t_0 < t_1$? Clearly, since the Poisson process is non-decreasing, $p^{(2)}(i_0, i_1; t_0, t_1)$ will be non-zero only for $0 \leq i_0 \leq i_1$.

Hint: The independence of increments of the Poisson process will allow you to answer this question with practically no calculations.

- (h) Use reasoning similar to part (g) to write $p^{(3)}(i_0, i_1, i_2; t_0, t_1, t_2)$ for $0 \leq t_0 < t_1 < t_2$ for the Poisson process; the result will be non-zero only for $0 \leq i_0 \leq i_1 \leq i_2$.
- (i) Use the concrete expressions for $p^{(1)}(i_0; t_0)$ and $p^{(2)}(i_0, i_1; t_0, t_1)$ to prove the identity

$$\sum_{i_0} p^{(2)}(i_0, i_1; t_0, t_1) = p^{(1)}(i_1; t_1), \quad t_0 < t_1$$

for the particular case of a Poisson process N with constant rate λ . What should be the range for the summation index i_0 , and why? You will need the binomial formula.

Problem 4. [Autocorrelation and autocovariance functions of a flip-flop process]

Consider again the flip-flop process $X = \{X_t : t \geq 0\}$ from Problem 3 of Homework 6. Namely, assume that $N = \{N_t : t \geq 0\}$ is a Poisson process with intensity λ , and define the flip-flop process $X = \{X_t : t \geq 0\}$ with state space $S = \{0, 1\}$ by

$$X_t = \frac{1}{2} + (-1)^{N_t} \left(X_0 - \frac{1}{2} \right),$$

where X_0 is a random variable with values in S that is independent of the process N . This complicated formula simply means that the process X switches between 0 and 1 at each event of N . Since N is a time-homogeneous Markov chain, X is also a time-homogeneous Markov chain. Assume that the initial state X_0 of the Markov chain X is a random variable with distribution $p_j(0) = \mathbb{P}(X_0 = j) = \frac{1}{2}$ for $j = 0, 1$. Note that $\mathbf{p}(0) = \left(\frac{1}{2} \ \frac{1}{2}\right)$ is the stationary distribution of the chain X , as you found in Homework 6. Recall that in Homework 6 you have found (in several ways) the stochastic semigroup \mathbf{P}_t of the continuous-time Markov chain X :

$$\mathbf{P}_t = \begin{pmatrix} p_{00}(t) & p_{01}(t) \\ p_{10}(t) & p_{11}(t) \end{pmatrix} = \begin{pmatrix} \frac{1}{2}(1 + e^{-2\lambda t}) & \frac{1}{2}(1 - e^{-2\lambda t}) \\ \frac{1}{2}(1 - e^{-2\lambda t}) & \frac{1}{2}(1 + e^{-2\lambda t}) \end{pmatrix}.$$

- (a) Show the distribution of the random variable X_t is $\mathbf{p}(t) = (p_0(t) \ p_1(t)) = \left(\frac{1}{2} \ \frac{1}{2}\right)$ (where $p_i(t) = \mathbb{P}(X_t = i)$). You should do this in two ways: (1) by a calculation using the explicit expressions for \mathbf{P}_t and $\mathbf{p}(0)$, and (2) by giving a simple reason why this is the answer, without any calculations.
- (b) For $s \geq 0, t \geq 0$, show that the expectation of X_t is $m_X(t) := \mathbb{E}[X_t] = \frac{1}{2}$, and the autocorrelation function is $R_X(s, s+t) := \mathbb{E}[X_s X_{s+t}] = \frac{1}{4} (1 + e^{-2\lambda t})$.

Hint: To compute the expected value of the product $X_s X_{s+t}$, use the Tower Rule by conditioning on the state of the system at the earlier time, namely, $\mathbb{E}[X_s X_{s+t}] = \mathbb{E}[\mathbb{E}[X_s X_{s+t} | X_s]]$. You can first compute

$$\mathbb{E}[X_s X_{s+t} | X_s = j] = j \mathbb{E}[X_{s+t} | X_s = j] = j \sum_{k=0}^1 k p_{jk}(t) = \dots,$$

and then apply the Tower Rule,

$$\mathbb{E}[X_s X_{s+t}] = \mathbb{E}[\mathbb{E}[X_s X_{s+t} | X_s]] = \sum_{j=0}^1 \mathbb{E}[X_s X_{s+t} | X_s = j] \mathbb{P}(X_s = j) = \cdots .$$

(c) Is the process X weakly stationary? Explain briefly.

(d) Compute the autocovariance function

$$C_X(s, s+t) := \text{Cov}(X_s, X_{s+t}) = \mathbb{E}[(X_s - \mathbb{E}X_s)(X_{s+t} - \mathbb{E}X_{s+t})] = R_X(s, s+t) - m_X(s) m_X(s+t) ,$$

and show that the autocorrelation function is equal to $e^{-2\lambda t}$.

Food for Thought Problem 1. [Riemann-Stieltjes integral]

Let the function $F : \mathbb{R} \rightarrow [0, 1]$ shown in Figure 2 be defined by $F(x) = 0$ for $x \leq 0$, $F(x) = 1$ for $x \geq 1$, and

$$F(x) = \frac{1}{2^{j-1}} \quad \text{for} \quad \frac{1}{2^j} \leq x < \frac{1}{2^{j-1}} , \quad j \in \{1, 2, 3, \dots\}$$

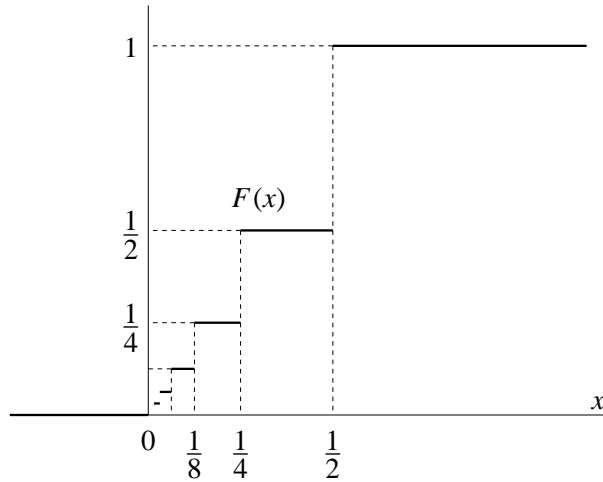


Figure 2: Graph of the function F .

Show that $\int_{\mathbb{R}} dF(x) = 1$, $\int_{\mathbb{R}} x dF(x) = \frac{1}{3}$, $\int_{\mathbb{R}} \ln x dF(x) = -2 \ln 2$.

Hint: You will need that, for $|q| < 1$, $\sum_{k=1}^{\infty} k q^{k-1} = \frac{d}{dq} \sum_{k=0}^{\infty} q^k = \frac{d}{dq} \frac{1}{1-q} = \frac{1}{(1-q)^2}$.

Food for Thought Problem 2. [Delta function and its derivatives]

Use the definition of Dirac δ -function and its derivatives to show that

$$\int_{\mathbb{R}} x^2 \delta_3(x) dx = 9 , \quad \int_{\mathbb{R}} e^{x^2} \delta'_3(x) dx = -6e^9 , \quad \int_{\mathbb{R}} e^{x^2} \delta''_3(x) dx = 38e^9 .$$