

# under Graduate Homework In Mathematics

**MarkovProcess**

王胤雅

201911010205

201911010205@mail.bnu.edu.cn

Beijing Normal University



General fire extinguisher

**PROBLEM I** Assume  $(\mathcal{F}_t : t \geq 0, t \in \mathbb{R})$  is a filtration. For  $t \geq 0$  we let  $\mathcal{F}_{t+} := \bigcap_{s>t} \mathcal{F}_s$ . Prove that  $\mathcal{F}_t \subset \mathcal{F}_{t+}$  and  $(\mathcal{F}_{t+} : t \geq 0)$  is a filtration.

1. Since  $(\mathcal{F}_t : t \geq 0, t \in \mathbb{R})$  is a filtration,  $\forall s > t, \mathcal{F}_s \supset \mathcal{F}_t$ , then by the definition of  $\mathcal{F}_{t+}$ ,  $\forall s > t, \forall x \in \mathcal{F}_t, x \in \mathcal{F}_s$ , so  $x \in \mathcal{F}_{t+}$ . Therefore,  $\mathcal{F}_t \subset \mathcal{F}_{t+}$ .
2. Since  $\forall s > t, \mathcal{F}_s$  is a  $\sigma$ -algebra, so it is obvious that  $\mathcal{F}_{t+}$  is a  $\sigma$ -algebra.  $\forall r > t, \forall s > r, \mathcal{F}_s \supset \mathcal{F}_r \supset \mathcal{F}_t$ , then  $\bigcap_{s>r} \mathcal{F}_s \supset \bigcap_{s>t} \mathcal{F}_s$ , that is  $\mathcal{F}_{r+} \supset \mathcal{F}_{t+}$ .

**PROBLEM II** Assume  $(X_t : t \geq 0, t \in \mathbb{R})$  is a stochastic process on probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . Prove that  $\forall s, t \geq 0, \varepsilon > 0, \{\rho(X_s, X_t) \geq \varepsilon\} \in \mathcal{F}$ .

*Lemma 1.*  $\{\rho(X_s, X_t) < \varepsilon\} = \bigcup_{q \in D} \{\rho(X_s, q) + \rho(X_t, q) < \varepsilon\}$ , where  $D := E \cap Q^d, E = R^d$ .

- 证明.*
1. Since  $\rho(X_s, X_t) < \rho(X_s, q) + \rho(X_t, q) < \varepsilon$ , then  $\{\rho(X_s, X_t) < \varepsilon\} \subset \bigcup_{q \in D} \{\rho(X_s, q) + \rho(X_t, q) < \varepsilon\}$ ,
  2. Only need to prove that if  $\rho(X_s, X_t) < \varepsilon$ , then  $\exists q \in D, \rho(X_s, q) + \rho(X_t, q) < \varepsilon$ . Since  $D$  is dense in  $E$ ,  $\exists q \in D$  s.t.  $\rho(X_s, q) \leq \frac{\varepsilon - \rho(X_s, X_t)}{4}$ , so  $\rho(X_t, q) + \rho(X_s, q) \leq \rho(X_t, X_s) + 2\rho(X_s, q) \leq \rho(X_t, X_s) + \frac{\varepsilon - \rho(X_s, X_t)}{2} < \varepsilon$ .

□

*Lemma 2.*  $(\Omega, \mathcal{F}, \mathbb{P})$  is a probability space,  $(E, \mathcal{E})$  is a measurable space,  $(E, \rho)$  is a distance space.  $(X_t : t \geq 0, t \in \mathbb{R})$  is a stochastic process from  $(\Omega, \mathcal{F}, \mathbb{P})$  to  $(E, \mathcal{E})$ . If  $(E, \rho)$  is separable, then  $\mathcal{B}(E)^2 = \mathcal{B}(E^2)$ . Moreover, if  $\mathcal{E} = \mathcal{B}(E)$ , then  $\forall \varepsilon > 0, s, t \geq 0, \{\rho(X_s, X_t) \geq \varepsilon\} \in \mathcal{F}$ .

- 证明.*
1. Let  $\mathcal{C}$  be all the open set of  $(E, \rho)$ ,  $\mathcal{B}(E)$  be the Borel algebra of  $(E, \rho)$ . Then  $\mathcal{B}(E) = \sigma(\mathcal{C})$ , where  $\sigma(\mathcal{C})$  means the  $\sigma$  algebra generated from  $\mathcal{C}$ . Then  $\mathcal{B}(E^2) = \sigma(\{A \times B : A, B \in \mathcal{C}\}) \supset \sigma(\{A \times E : A \in \mathcal{C}\}) = \sigma(\mathcal{C}) \times E = \mathcal{B}(E) \times E$ . By the same way, we can get that  $E \times \mathcal{B}(E) \subset \mathcal{B}(E^2)$ .  $\forall A, B \in \mathcal{B}(E), A \times B$ , then  $A \times B = (A \times E) \cap (E \times B) \in \mathcal{B}(E) \times E \cap E \times \mathcal{B}(E) \subset \mathcal{B}(E^2)$ . Therefore,  $\mathcal{B}(E)^2 = \sigma(\{A \times B : A, B \in \mathcal{B}(E)\}) \subset \mathcal{B}(E^2)$ . Since  $(E, \rho)$  is separable, then  $\exists \mathcal{D} \subset \mathcal{C}$ , which is a countable topology base of  $(E, \rho)$ . Then  $\forall A, B \in \mathcal{C}, A \times B \subset \sigma(\mathcal{D}^2)$ , and  $\sigma(\mathcal{D}^2) \subset \mathcal{B}(E^2)$ , so  $\mathcal{B}(E^2) = \sigma(\mathcal{D}^2)$ . Besides, obviously  $\sigma(\mathcal{D}^2) \subset \mathcal{B}(E)^2$ . Therefore,  $\mathcal{B}(E^2) \subset \mathcal{B}(E)^2$ . Then  $\mathcal{B}(E^2) = \mathcal{B}(E)^2$ .
  2. Since  $\{(x, y) \in E^2 : \rho(x, y) \geq \varepsilon\} \in \mathcal{B}(E^2) = \mathcal{B}(E)^2$ , then  $\exists A \in \mathcal{B}(E)^2$  s.t.  $\{(x, y) \in E^2 : \rho(x, y) \geq \varepsilon\} = A$ . Let  $\mathcal{H} := \{B \in \mathcal{B}(E)^2 : \{(X_s, X_t) \in B\} \in \mathcal{F}\}$ . Next, we will prove  $\mathcal{H} = \mathcal{B}(E)^2$ .

- (a)  $\mathcal{H}$  is a  $\sigma$ -algebra: obviously,  $E^2 \in \mathcal{H}$ . If  $B \in \mathcal{H}$ , then  $\{(X_s, X_t) \in B\} \in \mathcal{F}$ . So  $\{(X_s, X_t) \in B^c\} = \{(X_s, X_t) \in B\}^c \in \mathcal{F}$ . Thus,  $B^c \in \mathcal{H}$ . If  $(B_n \in \mathcal{H} : n \in \mathbb{N}^+)$ , then  $\{(X_s, X_t) \in B_n\} \in \mathcal{F}$ , then  $\{(X_s, X_t) \in \bigcup_{n \in \mathbb{N}^+} B_n\} = \bigcup_{n \in \mathbb{N}^+} \{(X_s, X_t) \in B_n\} \in \mathcal{F}$ .
- (b)  $\mathcal{H} \supset \{A_1 \times A_2 : A_1, A_2 \in \mathcal{B}(E)\}$ :  
Since  $\{(X_s, X_t) \in A_1 \times A_2\} = \{X_s \in A_1\} \cap \{X_t \in A_2\} \in \mathcal{F}$ , then  $A_1 \times A_2 \in \mathcal{H}$ .

Then,  $\{\rho(X_s, X_t) \geq \varepsilon\} = \{(X_s, X_t) \in A\} \in \mathcal{F}$ .

□

**SOLUTION.** 1. First way to solve the problem:

Since  $\mathcal{F}$  is a  $\sigma$ -algebra, then it is equal to prove that  $\forall s, t \geq 0, \varepsilon > 0, \{\rho(X_s, X_t) < \varepsilon\} \in \mathcal{F}$ . By Lemma 1 and  $D$  is countable, only need to prove that  $\forall q \in D, \{\rho(X_s, q) + \rho(X_t, q) < \varepsilon\} \in \mathcal{F}$ . And obviously,  $\{\rho(X_s, q) + \rho(X_t, q) < \varepsilon\} = \bigcup_{p \in D \cap [0, \varepsilon]} \{\rho(X_s, q) < p, \rho(X_t, q) < \varepsilon - p\}$ . So only need to prove that  $\{\rho(X_s, q) < p, \rho(X_t, q) < \varepsilon - p\} \in \mathcal{F}$ . Since  $\{\rho(X_s, q) < p, \rho(X_t, q) < \varepsilon - p\} = \{\rho(X_s, q) < p\} \cap \{\rho(X_t, q) < \varepsilon - p\}$ , and  $(X_t : t \geq 0, t \in \mathbb{R})$  is a stochastic process, then  $\{\rho(X_s, q) < p\}, \{\rho(X_t, q) < \varepsilon - p\} \in \mathcal{F}$ .

2. Second way to solve the problem:

Since  $E \subset \mathbb{R}^d, \mathcal{E} = E \cap \mathcal{B}^d$ , so  $(E, \rho)$  can be a separable distance space, where  $\rho$  is the distance in  $\mathbb{R}^d$ . By Lemma 2, we get  $\{\rho(X_s, X_t) \geq \varepsilon\} \in \mathcal{F}$ .

□

**PROBLEM III** Let  $\mathcal{D}_X := \{\mu_J^X : J \in S(I)\}$  be the family of finite-dimensional distributions of a stochastic process  $(X_t : t \geq 0, t \in \mathbb{R})$ .  $\forall (s_1, s_2) \in S(I)$  and  $J = (t_1, \dots, t_n) \in S(I)$ , write  $K_1 := (s_1, s_2, t_1, \dots, t_n) \in S(I), K_2 := (s_2, s_1, t_1, \dots, t_n) \in S(I)$ . Take  $A_1, A_2 \in \mathcal{E}, B \in \mathcal{E}^n$ , prove that

$$\mu_{K_1}^X(A_1 \times A_2 \times B) = \mu_{K_2}^X(A_2 \times A_1 \times B)$$

and

$$\mu_{K_1}^X(E \times E \times B) = \mu_{K_2}^X(E \times E \times B) = \mu_J^X(B)$$

**SOLUTION.** By the definition of  $\mu_J^X(H) := \mathbb{P}\{(X_{t_1}, \dots, X_{t_n}) \in H\}$ , where  $J = (t_1, \dots, t_n) \in S(I)$ ,  $H \in \mathcal{F}$ . Then

$$\begin{aligned} & \mu_{K_1}^X(A_1 \times A_2 \times B) \\ &= \mathbb{P}(\{(X_{s_1}, X_{s_2}, X_{t_1}, \dots, X_{t_n}) \in A_1 \times A_2 \times B\}) \\ &= \mathbb{P}(\{X_{s_1} \in A_1, X_{s_2} \in A_2, (X_{t_1}, \dots, X_{t_n}) \in B\}) \\ &= \mathbb{P}(\{(X_{s_2}, X_{s_1}, X_{t_1}, \dots, X_{t_n}) \in A_1 \times A_2 \times B\}) \\ &= \mu_{K_2}^X(A_1 \times A_2 \times B) \end{aligned} \tag{1}$$

Especially, when  $A_1 = A_2 = E$ , the equation is true as well. So only need to prove:  $\mu_{K_1}^X(E \times E \times B) = \mu_J^X(B)$ . And

$$\begin{aligned} & \mu_{K_1}^X(E \times E \times B) \\ &= \mathbb{P}(\{X_{s_1} \in E, X_{s_2} \in E, (X_{t_1}, \dots, X_{t_n}) \in B\}) \\ &= \mathbb{P}(\{X_{s_1} \in E\})\mathbb{P}(\{X_{s_2} \in E\})\mathbb{P}(\{(X_{t_1}, \dots, X_{t_n}) \in B\}) \\ &= \mathbb{P}(\{(X_{t_1}, \dots, X_{t_n}) \in B\}) \\ &= \mu_J^X(B) \end{aligned} \tag{2}$$

□

**PROBLEM IV** Assume  $(\tau_k : k \in \mathbb{N}^+)$  is an i.i.d sequence of r.v. with exponential distribution with parameter  $\alpha > 0$ . Let  $S_n := \sum_{k=1}^n \tau_k$ . For  $t \geq 0, t \in \mathbb{R}$ , let:

$$N_t := \sum_{n=1}^{\infty} \mathbb{1}_{\{S_n \leq t\}}, X_t := \sum_{n=1}^{\infty} \mathbb{1}_{\{S_n < t\}}$$

Prove that  $N$  and  $X$  are modifications of each other, but they are not indistinguishable.

**SOLUTION**. Since  $S_n = \sum_{k=1}^n \tau_k$ , where  $(\tau_k : k \in \mathbb{N}^+)$  is i.i.d., then by SLLN  $\frac{S_n}{n} \rightarrow \mathbb{E}(\tau_1) < \infty$ . So  $S_n \rightarrow \infty$ , then  $N_t, X_t$  are all well-defined r.v.. Since  $\forall t \geq 0, t \in \mathbb{R}$ ,  $\mathbb{P}(\{N_t \neq X_t\}) = \mathbb{P}(\{S_n = t, \exists n \in \mathbb{N}^+\})$ . Since  $(\tau_k, k \in \mathbb{N}^+)$  are all continuous i.i.d. r.v., then  $(S_n, n \in \mathbb{N}^+)$  are all continuous r.v.. Therefore,  $\mathbb{P}(\{S_n = t, \exists n \in \mathbb{N}^+\}) = \sum_{n=1}^{\infty} \mathbb{P}(\{S_n = t\}) = 0$ . So  $N$  and  $X$  are modifications. And  $\forall \omega \in \Omega$ ,  $t := \tau_1(\omega)$ ,  $N_t = 1 \neq X_t = 0$ , then  $\{N_t = X_t, \forall t \geq 0, t \in \mathbb{R}\} = \emptyset$ . Thus,  $N$  and  $X$  are not indistinguishable.  $\square$

**PROBLEM V** Assume  $T$  is non-negative r.v. with distribution function  $F$  continuous on  $\mathbb{R}$ . Let  $X_t = \mathbb{1}_{\{T \leq t\}}$ . Prove that  $X$  is stochastically continuous.

**SOLUTION**.  $\forall \varepsilon > 0, t \geq 0, t \in \mathbb{R}$ ,  $s \rightarrow t^+$ ,  $\mathbb{P}(|X_s - X_t| \geq \varepsilon) = \mathbb{P}(\{X_s - X_t \geq \varepsilon\}) = \mathbb{P}(\{X_s = 1, X_t = 0\}) = \mathbb{P}(\{t < T \leq s\}) = F(s) - F(t) \rightarrow 0$ . In the same way, we can get  $s \rightarrow t^-$ ,  $\mathbb{P}(|X_s - X_t| \geq \varepsilon) \rightarrow 0$ . Therefore,  $X$  is stochastically continuous.  $\square$

**PROBLEM VI** Assume  $I = \mathbb{Z}^+$ , then the stochastic process  $X = (X_0, X_1, \dots)$  is a r.v. from  $\Omega$  to  $E^\infty$ . Define the distribution of  $X$ ,  $\mu_X$ , as follows:

$$\mu_X(A) = \mathbb{P}(X \in A), A \in \mathcal{E}^\infty$$

Then stochastic process  $X, Y$  are equivalent  $\iff \mu_X = \mu_Y$ .

**SOLUTION**. 1. If  $X, Y$  are equivalent, so  $\forall J \in S(\mathbb{N}), \forall A \in E^{|J|}, \mu_J^X(A) = \mathbb{P}(\{(X_{t_1}, \dots, X_{t_n}) \in A, (t_1, \dots, t_n) = J\}) = \mathbb{P}(\{(Y_{t_1}, \dots, Y_{t_n}) \in A\})$ . Since  $\mathcal{E}^\infty = \sigma(\mathcal{C})$ , where  $\mathcal{C} := \{A_J \times \prod_{i \in J^c} E_i : J \in S(\mathbb{N})\}$  is a semialgebraic set. Then  $\forall A_J \times \prod_{i \in J^c} E_i \in \mathcal{C}$ ,

$$\begin{aligned} & \mu_X(A) \\ &= \mathbb{P}(X \in A_J \times \prod_{i \in J^c} E_i) \\ &= \mathbb{P}((X_{t_1}, \dots, X_{t_n}) \in A_J, X_i \in E_i, i \in J^c) \\ &= \mathbb{P}((X_{t_1}, \dots, X_{t_n}) \in A_J) \\ &= \mu_J^X(A) \\ &= \mu_J^Y(A) \\ &= \mu_Y(A) \end{aligned} \tag{3}$$

By the measure extension theorem,  $\mu_X(A) = \mu_Y(A), \forall A \in \mathcal{E}^\infty$ .

2. If  $\mu_X(A) = \mu_Y(A), \forall A \in \mathcal{E}^\infty$ , then by the discussion above, we get easily  $X, Y$  are equivalent.  $\square$