## under Graduate Homework In Mathematics

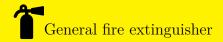
## **MarkovProcess**

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

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

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

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

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

- 近明. 1. Let  $\mathscr{C}$  be all the open set of  $(E, \rho)$ ,  $\mathscr{B}(E)$  be the Borel algebra of  $(E, \rho)$ . Then  $\mathscr{B}(E) = \sigma(\mathscr{C})$ , where  $\sigma(\mathscr{C})$  means the  $\sigma$  algebra generated from  $\mathscr{C}$ . Then  $\mathscr{B}(E^2) = \sigma(\{A \times B : A, B \in \mathscr{C}\}) \supset \sigma(\{A \times E : A \in \mathscr{C}\}) = \sigma(\mathscr{C}) \times E = \mathscr{B}(E) \times E$ . By the same way, we can get that  $E \times \mathscr{B}(E) \subset \mathscr{B}(E^2)$ .  $\forall A, B \in \mathscr{B}(E), A \times B$ , then  $A \times B = (A \times E) \cap (E \times B) \in \mathscr{B}(E) \times E \cap E \times \mathscr{B}(E) \subset \mathscr{B}(E^2)$ . Therefore,  $\mathscr{B}(E)^2 = \sigma(\{A \times B : A, B \in \mathscr{B}(E)\}) \subset \mathscr{B}(E^2)$ . Since  $(E, \rho)$  is separable, then  $\exists \mathscr{D} \subset \mathscr{C}$ , which is a countable topology base of  $(E, \rho)$ . Then  $\forall A, B \in \mathscr{C}, A \times B \subset \sigma(\mathscr{D}^2)$ , and  $\sigma(\mathscr{D}^2) \subset \mathscr{B}(E^2)$ , so  $\mathscr{B}(E^2) = \sigma(D^2)$ . Besides, oboviously  $\sigma(\mathscr{D}^2) \subset \mathscr{B}(E)^2$ . Therefore,  $\mathscr{B}(E^2) \subset \mathscr{B}(E)^2$ . Then  $\mathscr{B}(E^2) = \mathscr{B}(E)^2$ .
  - 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)  $\mathscr{H}$  is a  $\sigma$ -algebra: obviously,  $E^2 \in \mathscr{H}$ . If  $B \in \mathscr{H}$ , then  $\{(X_s, X_t) \in B\} \in \mathscr{F}$ . So  $\{(X_s, X_t) \in B^c\} = \{(X_s, X_t) \in B\}^c \in \mathscr{F}$ . Thus,  $B^c \in \mathscr{F}$ . If  $(B_n \in \mathscr{F} : n \in \mathbb{N}^+)$ , then  $\{(X_s, X_t) \in B_n\} \in \mathscr{F}$ , then  $\{(X_s, X_t) \in \bigcup_{n \in \mathbb{N}^+} B_n\} = \bigcup_{n \in \mathbb{N}^+} \{(X_s, X_t) \in B_n\} \in \mathscr{F}$ .
    - (b)  $\mathscr{H} \supset \{A_1 \times A_2 : A_1, A_2 \in \mathscr{B}(E)\}:$ Since  $\{(X_s, X_t) \in A_1 \times A_2\} = \{X_s \in A_1\} \cap \{X_t \in A_2\} \in \mathscr{F}$ , then  $A_1 \times A_2 \in \mathscr{F}$ .

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

SOLION. 1. First way to solve the problem:

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

2. Second way to solve the problem:

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

ROBEM III Let  $\mathscr{D}_X := \{\mu_J^X : J \in S(I)\}$  be the family of finite-dimentional 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  $A_1, A_2 \in \mathscr{E}, B \in \mathscr{E}^n$ , 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

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

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

$$\mu_{K_{1}}^{X}(A_{1} \times A_{2} \times B)$$

$$= \mathbb{P}(\{(X_{s_{1}}, X_{s_{2}}, X_{t_{1}}, \cdots, X_{t_{n}}) \in A_{1} \times A_{2} \times B\})$$

$$= \mathbb{P}(\{X_{s_{1}} \in A_{1}, X_{s_{2}} \in A_{2}, (X_{t_{1}}, \cdots, X_{t_{n}}) \in B\})$$

$$= \mathbb{P}(\{(X_{s_{2}}, X_{s_{1}}, X_{t_{1}}, \cdots, X_{t_{n}}) \in A_{1} \times A_{2} \times B\})$$

$$= \mu_{K_{2}}^{X}(A_{1} \times A_{2} \times B)$$

$$(1)$$

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

$$\mu_{K_{1}}^{X}(E \times E \times B) 
= \mathbb{P}(\{X_{s_{1}} \in E, X_{s_{2}} \in E, (X_{t_{1}}, \dots, X_{t_{n}}) \in B\}) 
= \mathbb{P}(\{X_{s_{1}} \in E\}) \mathbb{P}(\{X_{s_{2}} \in E\}) \mathbb{P}(\{(X_{t_{1}}, \dots, X_{t_{n}}) \in B\}) 
= \mathbb{P}(\{(X_{t_{1}}, \dots, X_{t_{n}}) \in B\}) 
= \mu_{I}^{X}(B)$$
(2)

ROBEM 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:

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

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

SOUTHON. Since  $S_n = \sum_{k=1}^n \tau_k$ , where  $(\tau_k : k \in \mathbb{N}^+)$  is i.i.d., then by SLLN  $\frac{S_n}{n} \to \mathbb{E}(\tau_1) < \infty$ . So  $S_n \to \infty$ , 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.

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

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

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

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

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

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

$$\mu_{X}(A)$$

$$=\mathbb{P}(X \in A_{J} \times \prod_{i \in J^{c}} E_{i})$$

$$=\mathbb{P}((X_{t_{1}}, \dots, X_{t_{n}}) \in A_{J}, X_{i} \in E_{i}, i \in J^{c})$$

$$=\mathbb{P}((X_{t_{1}}, \dots, X_{t_{n}}) \in A_{J})$$

$$=\mu_{J}^{X}(A)$$

$$=\mu_{J}^{Y}(A)$$

$$=\mu_{Y}(A)$$
(3)

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

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

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