

**PROBLEM I** Assume  $(\Omega, \mathcal{F}, \mathbb{P})$  is a probability space, and  $C \in \mathcal{F}$  satisfy  $\mathbb{P}(C) > 0$ . Let  $\mathbb{P}_C : \mathcal{F} \rightarrow \mathbb{R}$ ,  $\mathbb{P}_C(X) = \frac{\mathbb{P}(C \cap X)}{\mathbb{P}(C)}$ . Assume  $A, B \in \mathcal{F}$ , and  $\mathbb{P}(B \cap C) > 0$ , prove that  $\mathbb{P}_C(A | B) = \mathbb{P}(A | B \cap C)$ .

**SOLUTION**. Easily  $\mathbb{P}_C(B) = \frac{\mathbb{P}(B \cap C)}{\mathbb{P}(C)} > 0$ , so  $\mathbb{P}_C(A | B)$  is well-defined. Easily to get that

$$\mathbb{P}_C(A | B) = \frac{\mathbb{P}_C(A \cap B)}{\mathbb{P}_C(B)} = \frac{\frac{\mathbb{P}(A \cap B \cap C)}{\mathbb{P}(C)}}{\frac{\mathbb{P}(B \cap C)}{\mathbb{P}(C)}} = \frac{\mathbb{P}(A \cap B \cap C)}{\mathbb{P}(B \cap C)} = \mathbb{P}(A | B \cap C)$$

□

**PROBLEM II** Assume that  $(X_n : n \geq 0)$  is 1-dimentional simple symetry random walk, prove that  $(|X_n| : n \geq 0)$  is a Markov chain ranges in  $\mathbb{N}$ .

**SOLUTION**. Easy to know that  $(X_n : n \geq 0)$  is a Markov chain in  $\mathbb{Z}$ . Let  $\mathcal{F}_n := \sigma(X_1, \dots, X_n)$ ,  $\mathcal{G}_n := \sigma(|X_1|, \dots, |X_n|)$ , then easily  $\mathcal{G}_n \subset \mathcal{F}_n$ . Then we get that  $\mathbb{P}(|X_{n+1}| = i | \mathcal{F}_n) = \mathbb{P}(X_{n+1} = i | \mathcal{F}_n) + \mathbb{P}(X_{n+1} = -i | \mathcal{F}_n) = \mathbb{P}(X_{n+1} = i | X_n) + \mathbb{P}(X_{n+1} = -i | X_n) = \mathbb{P}(|X_{n+1}| = i | X_n) = \frac{1}{2} \mathbb{1}(|X_n - i| = 1)$ . Noting  $i \geq 0$ , we get that  $|X_n - i| = 1 \iff ||X_n| - i| = 1$ , so  $\mathbb{P}(|X_{n+1}| = i | \mathcal{F}_n)$  is measureable about  $\sigma(|X_n|)$ . Since  $\sigma(|X_n|) \subset \mathcal{G}_n \subset \mathcal{F}_n$ , so we finally get that  $\mathbb{P}(|X_{n+1}| = i | \mathcal{G}_n) = \mathbb{P}(|X_{n+1}| = i | |X_n|)$ . So  $(|X_n| : n \geq 0)$  is a Markov chain. □

**PROBLEM III** Assume  $(X_n : n \geq 0)$  is a Markov chain ranges in  $E$ . Assume  $\phi : E \rightarrow F$  is injection. Prove that  $(\phi(X_n) : n \geq 0)$  is a Markov chain ranges in  $\phi(E)$ .

**SOLUTION**. Without loss of generality assume  $F = \phi(E)$ , then  $\phi$  is bijection. Now let  $\mathcal{F}_n := \sigma(X_1, \dots, X_n)$ . Since  $\phi$  is bijection we easily get that  $\sigma(X_n) = \sigma(\phi(X_n))$ , so  $\mathcal{F}_n = \sigma(\phi(X_1), \dots, \phi(X_n))$ . Then  $\mathbb{P}(\phi(X_{n+1}) = i | \mathcal{F}_n) = \mathbb{P}(X_{n+1} = \phi^{-1}(i) | \mathcal{F}_n) = \mathbb{P}(X_{n+1} = \phi^{-1}(i) | X_{n+1}) = \mathbb{P}(\phi(X_{n+1}) = i | \phi(X_n))$ . So  $(\phi(X_n) : n \geq 0)$  is Markov chain. □

**PROBLEM IV** Assume  $(X_n : n \geq 0), (Y_n : n \geq 0)$  are two independent Markov chains on  $E, F$  respectively. Prove that  $((X_n, Y_n) : n \geq 0)$  is Markov chain on  $E \times F$ .

**SOLUTION**. Let  $\mathcal{F}_n = \sigma(X_0, \dots, X_n)$  and  $\mathcal{G}_n = \sigma(Y_0, \dots, Y_n)$ , Let  $\mathcal{H}_n = \sigma((X_0, Y_0), \dots, (X_n, Y_n))$ . Then easily  $\mathcal{H}_n = \sigma(\mathcal{F}_n, \mathcal{G}_n)$ . Easy to get that

$$\begin{aligned} \mathbb{P}(X_{n+1} = i, Y_{n+1} = j | \mathcal{H}_n) &= \mathbb{E}(\mathbb{P}(X_{n+1} = i, Y_{n+1} = j | \mathcal{H}_n, X_{n+1}) | \mathcal{H}_n) \\ &= \mathbb{E}(\mathbb{1}_i(X_{n+1}) \mathbb{P}(Y_{n+1} = j | \mathcal{F}_n, \mathcal{G}_n, X_{n+1}) | \mathcal{H}_n) \\ (Y_{n+1} \perp \mathcal{F}_n, X_{n+1}) &= \mathbb{E}(\mathbb{1}_i(X_{n+1}) \mathbb{P}(Y_{n+1} = j | Y_n) | \mathcal{H}_n) \\ (Y_n \in \mathcal{H}_n) &= \mathbb{P}(Y_{n+1} = j | Y_n) \mathbb{P}(X_{n+1} = i | \mathcal{H}_n) \\ &= \mathbb{P}(Y_{n+1} = j | Y_n) \mathbb{P}(X_{n+1} = i | X_n) \end{aligned}$$

So  $\mathbb{P}(X_{n+1} = i, Y_{n+1} = j | \mathcal{H}_n) \in \sigma(X_n, Y_n) \subset \mathcal{H}_n$ . So  $\mathbb{P}(X_{n+1} = i, Y_{n+1} = j | X_n, Y_n) = \mathbb{P}(X_{n+1} = i, Y_{n+1} = j | \mathcal{H}_n)$ . So  $((X_n, Y_n) : n \geq 0)$  is Markov chain. □

**PROBLEM V** Assume  $(X_n : n \geq 0), (Y_n : n \geq 0)$  are two independent Markov chains on  $E, F$  respectively. Let  $\mathcal{H}_n := \sigma((X_0, Y_0), \dots, (X_n, Y_n))$ . Prove that  $(X_n : n \geq 0)$  is Markov chain over  $(\mathcal{H}_n : n \geq 0)$ .

**SOLUTION**. Take  $\mathcal{F}_n, \mathcal{G}_n$  as above. Obviously  $X_n \in \mathcal{F}_n \subset \mathcal{H}_n$ . Easily  $\mathbb{P}(X_{n+1} = i \mid \mathcal{H}_n) = \mathbb{P}(X_{n+1} = i \mid \mathcal{F}_n, \mathcal{G}_n) = \mathbb{P}(X_{n+1} = i \mid \mathcal{F}_n) = \mathbb{P}(X_{n+1} \mid X_n)$ . So  $(X_n : n \geq 0)$  is Markov chain over  $(\mathcal{H}_n : n \geq 0)$ .  $\square$

**PROBLEM VI** Let  $\mu_0$  be a probability distribution on  $\mathbb{N}$ . For  $n \geq 1$ , let

$$\mu_n(0) = \mu_{n-1}^{*2}(0) + \mu_{n-1}^{*2}(1), \mu_n(j) = \mu_{n-1}^{*2}(j+1), \forall j \geq 1$$

Where  $\mu^{*2} = \mu * \mu$ . Let  $F_n$  be distribution function of  $\mu_n$ . Let  $F_{n-1}^{-1}(y) := \inf\{x \geq 0 : y \leq F_{n-1}(x)\}$  for  $y \in [0, 1]$ . Assume  $X_0 \sim \mu_0$ , and  $(U_n : n \geq 0)$  are i.i.d r.v. with distribution  $U(0, 1)$ . Let  $X_{n+1} := \max\{0, X_n + F_n^{-1}(U_n) - 1\}$ . Then  $(X_n : n \geq 0)$  is Markov chain.

**SOLUTION**. Let  $\mathcal{F} := \sigma(X_0, \dots, X_n)$ . For  $i > 0$ , we have

$$\begin{aligned} \mathbb{P}(X_{n+1} = i \mid \mathcal{F}_n) &= \mathbb{P}(X_n + F_n^{-1}(U_n) - 1 = i \mid \mathcal{F}_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{P}(X_n = k, F_n^{-1}(U_n) = i + 1 - k \mid \mathcal{F}_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{1}_k(X_n) \mathbb{P}(F_n^{-1}(U_n) = i + 1 - k) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{1}_k(X_n) \mathbb{P}(F_n^{-1}(U_n) = i + 1 - k \mid X_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{P}(X_n = k, F_n^{-1}(U_n) = i + 1 - k \mid X_n) \\ &= \mathbb{P}(X_n + F_n^{-1}(U_n) - 1 = i \mid X_n) = \mathbb{P}(X_{n+1} = i \mid X_n) \end{aligned}$$

For  $i = 0$ , we have

$$\begin{aligned} \mathbb{P}(X_{n+1} = 0 \mid \mathcal{F}_n) &= \mathbb{P}(X_n + F_n^{-1}(U_n) - 1 \leq 0 \mid \mathcal{F}_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{P}(X_n = k, F_n^{-1}(U_n) \leq 1 - k \mid \mathcal{F}_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{1}_k(X_n) \mathbb{P}(F_n^{-1}(U_n) \leq 1 - k) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{1}_k(X_n) \mathbb{P}(F_n^{-1}(U_n) \leq 1 - k \mid X_n) \\ &= \sum_{k \in \mathbb{Z}} \mathbb{P}(X_n = k, F_n^{-1}(U_n) \leq 1 - k \mid X_n) \\ &= \mathbb{P}(X_n + F_n^{-1}(U_n) - 1 \leq 0 \mid X_n) = \mathbb{P}(X_{n+1} = 0 \mid X_n) \end{aligned}$$

So  $(X_n : n \geq 0)$  is Markov chain.  $\square$