

**PROBLEM I** Prove that if  $(X_n : n \geq 0)$  is a simple random walk, then so is  $(-X_n : n \geq 0)$ .

**SOLUTION.** Let  $\xi_n := X_n - X_{n-1}$  for  $n \in \mathbb{N}^+$ . Then Since  $(X_n : n \in \mathbb{N})$  is simple random walk we have  $X_0, \xi_1, \xi_2, \dots$  are independent r.v. ranges in  $\mathbb{Z}$ , and  $\xi_i, i = 1, 2, \dots$  are i.i.d., and  $\mathbb{P}(|\xi_i| = 1) = 1$ . So we easily get  $-X_0, -\xi_1, -\xi_2, \dots$  are independent r.v. ranges in  $\mathbb{Z}$ , and  $-\xi_i, i = 1, 2, \dots$  are i.i.d., and  $\mathbb{P}(|-\xi_i| = 1) = 1$ . Since  $-X_n = X_0 + \sum_{k=1}^n \xi_k$ , by the definition of simple random walk we obtain  $(-X_n : n \in \mathbb{N})$  is a simple random walk.  $\square$

**PROBLEM II** Let  $(X_n : n \geq 0)$  be a  $d$ -dimensional random walk with  $\mathbb{P}(|\xi_i| \geq 1) > 0$ , prove that  $\mathbb{P}(\sup_n |X_n| = \infty) = 1$ .

**SOLUTION.** Let  $t \in \mathbb{Z}^d, t \neq 0$  and  $\mathbb{P}(\xi_i = t) > 0$ . Since  $\mathbb{P}(\sup_n |X_n| = \infty) = \inf_{k \in \mathbb{N}} \mathbb{P}(\sup_n |X_n| \geq k)$ , we only need to prove  $\mathbb{P}(\sup_n |X_n| \geq k) = 1$  for every  $k \in \mathbb{N}$ . Take  $K > 3k, K \in \mathbb{N}$ . Let  $A_s := \{\xi_i = t : i = sK + 1, sK + 2, \dots, sK + K - 1\}$ . Then for  $\omega \in A_s$ , we have  $|X_{sK+K} - X_{sK}| = |\sum_{u=1}^{K-1} t| = K|t| \geq K \geq 3k$ . Then  $\sup_n |X_n| \geq \max\{|X_{sK+K}|, |X_{sK}|\} \geq \frac{1}{2}|X_{sK+K} - X_{sK}| \geq k$ . So we get  $\forall s, A_s \subset \{\sup_n |X_n| \geq k\}$ . Since  $\xi_i$  are independent, easily get  $A_s, s = 1, 2, \dots$  are independent. Noting  $\mathbb{P}(A_s) = \mathbb{P}(\xi_i = t)^K > 0$ , we get  $\sum_{s \in \mathbb{N}} \mathbb{P}(A_s) = \infty$ . So from BC-theorem we get  $\mathbb{P}(A_s, i.o.) = 1$ , thus  $\mathbb{P}(\bigcup_{s \in \mathbb{N}} A_s) = 1$ . Thus,  $\mathbb{P}(\sup_n |X_n| \geq k) = 1$ , for every  $k \in \mathbb{N}$ . Thus,  $\mathbb{P}(\sup_n |X_n| = \infty) = \mathbb{P}(\bigcap_{k \in \mathbb{N}} \{\sup_n |X_n| \geq k\}) = 1$ .  $\square$

**PROBLEM III** Let  $(S_n : n \geq 0)$  be a symmetry simple random walk with  $S_0 = 0$ , for  $d = 2$ , prove that

$$\mathbb{P}(S_{2n} = 0) = \frac{1}{4^{2n}} \left( \frac{(2n)!}{(n!)^2} \right)^2$$

For  $d = 3$ , prove that

$$\mathbb{P}(S_{2n} = 0) = \frac{1}{2^{2n}} \frac{(2n)!}{(n!)^2} \sum_{i+j+k=n} \left( \frac{1}{3^n} \frac{n!}{i!j!k!} \right)^2$$

**SOLUTION.** First we consider  $d = 2$ . Write  $\xi_i = S_i - S_{i-1}$ . Then we know  $S_{2n}$  occur  $\iff$  the number of  $(1, 0)$  and  $(-1, 0)$  in  $\{\xi_i : i = 1, \dots, 2n\}$ , and the number of  $(0, 1)$  and  $(0, -1)$  in  $\{\xi_i : i = 1, \dots, 2n\}$ . We assume there is  $k$  pairs of  $(1, 0), (-1, 0)$ , then easily there is  $n - k$  pairs of  $(0, 1), (0, -1)$ . The probability is  $\binom{2n}{k} \binom{2n-k}{n-k} \binom{2n-2k}{n-2k} \frac{1}{4^{2n}}$ . So the total probability is  $\mathbb{P}(S_{2n} = 0) = \sum_{k=0}^n \binom{2n}{k} \binom{2n-k}{n-k} \binom{2n-2k}{n-2k} \frac{1}{4^{2n}} = \sum_{k=0}^n \frac{(2n)!}{k!k!(n-k)!(n-k)!4^{2n}}$ . Noting that  $\sum_{k=0}^n \frac{(n!)^2}{k!k!(n-k)!(n-k)!} = \sum_{k=0}^n \binom{n}{k} \binom{n}{n-k} = \binom{2n}{n} = \frac{(2n)!}{n!n!}$ , we finally get  $\mathbb{P}(S_{2n} = 0) = \frac{1}{4^{2n}} \left( \frac{(2n)!}{n!n!} \right)^2$ .

Use the same method, consider  $d = 3$ , we have  $\mathbb{P}(S_{2n} = 0) = \sum_{i+j+k=n} \binom{2n}{i} \binom{2n-i}{i} \binom{2n-2i}{j} \binom{2n-2i-j}{j} \binom{2n-2i-2j}{k} \frac{1}{3^{2n}}$ . So easily to get  $\mathbb{P}(S_{2n} = 0) = \frac{1}{2^{2n}} \frac{(2n)!}{(n!)^2} \sum_{i+j+k=n} \left( \frac{1}{3^n} \frac{n!}{i!j!k!} \right)^2$ .  $\square$

**PROBLEM IV** Assume  $(S_n : n \geq 0)$  is a symmetry simple random walk with  $S_0 = i \in \mathbb{Z}$ . Prove that  $\forall a \in \mathbb{Z}$ , let  $\tau_a := \min\{n \in \mathbb{N} : S_n = a\}$ , then  $\mathbb{P}(\tau_a < \infty) = 1$ .

**SOLUTION.** Without loss of generality assume  $a < 0, i = 0$ . Take  $N \in \mathbb{N}^+$ . Consider  $\tau := \min\{n \in \mathbb{N} : S_n = a \vee S_n = N\}$ . From Problem ?? we can easily know  $\mathbb{P}(\tau < \infty) = 1$  because  $\{\sup_n |S_n| = \infty\} \subset \{\tau < \infty\}$ , a.s. So we get  $\{\tau_a = \tau\} \subset \{\tau_a < \infty\}$ , a.s. Let  $Y_n := S_{n \wedge \tau} := S_{\min\{n, \tau\}}$ . Easily

$(S_n : n \in \mathbb{N})$  is a martingale, and  $\tau$  is a stopping time, so we get  $(Y_n : n \in \mathbb{N})$  is a martingale, too. And easily  $Y_n \in [a, N]$ , so  $Y_n$  is bounded. So we get  $\mathbb{E}(S_\tau) = \lim_{n \rightarrow \infty} \mathbb{E}(Y_n) = \mathbb{E}(Y_0) = 0$ . Easily to know  $\mathbb{E}(S_\tau) = \mathbb{P}(\tau = \tau_a)a + \mathbb{P}(\tau \neq \tau_a)N = 0$ . And  $\mathbb{P}(\tau = \tau_a) + \mathbb{P}(\tau \neq \tau_a) = 1$ , so easily  $\mathbb{P}(\tau = \tau_a) = \frac{N}{N-a}$ . So  $\mathbb{P}(\tau_a < \infty) \geq \frac{N}{N-a}$ . Let  $N \rightarrow \infty$ , we get  $\mathbb{P}(\tau_a < \infty) = 1$ .  $\square$