

# Lecture 27: Random Walks

## 1 Introduction

Let  $X = \{X_n : n \in \mathbb{N}\}$  be a sequence of *iid* random variables with  $\mathbb{E}|X_n| < \infty$ . Let  $S_0 = 0$  and  $S_n = \sum_{i=1}^n X_i$ . Then the sequence  $S = \{S_n : n \in \mathbb{N}_0\}$  is called a *random walk process*.

**Example 1.1 (Simple random walk).** Let  $X_i \in \{-1, 1\}$  and  $\Pr\{X_i = 1\} = p$ , then random walk  $S$  is called a simple random walk.

**Example 1.2.** Stock prices each day can be modeled by a random walk.

Random walks are generalizations of renewal processes. If  $X$  was a sequence of non-negative random variables indicating inter-renewal times, then  $S_n$  is the instant of the  $n$ th renewal event.

## 2 Duality in random walks

**Lemma 2.1 (Duality principle).** For any finite  $n \in \mathbb{N}$ , the joint distributions of finite sequence  $(X_1, X_2, \dots, X_n)$  and the reversed sequence  $(X_n, X_{n-1}, \dots, X_1)$  are identical.

*Proof.* Since  $X$  is a sequence of *iid* random variables, it is exchangeable. The reversed sequence is  $(X_{\pi(1)}, \dots, X_{\pi(n)})$  where  $\pi : [n] \rightarrow [n]$  is permutation with  $\pi(i) = n - i + 1$ .  $\square$

**Corollary 2.2.** Distribution of  $S_k$  and  $S_n - S_{n-k}$  are identical for any  $k \in [n]$ .

*Proof.* Using duality principle, we can write the following equality for any  $x \in \mathbb{R}, k \in [n]$

$$\Pr\{S_k \leq x\} = \Pr\left\{\sum_{i=1}^k X_i \leq x\right\} = \Pr\left\{\sum_{i=1}^k X_{n-i+1} \leq x\right\} = \Pr\left\{\sum_{i=n-k+1}^n X_i \leq x\right\} = \Pr\{S_n - S_{n-k} \leq x\}.$$

$\square$

**Proposition 2.3.** Suppose  $\{X_n : n \in \mathbb{N}\}$  is a sequence of *iid* random variables with positive mean. Let  $S_n = \sum_{k=1}^n X_k$  be a random walk with step size  $X_n$ . If  $N = \min\{n \in \mathbb{N} : S_n > 0\}$ , then  $\mathbb{E}N < \infty$ .

*Proof.* Consider a discrete process  $\{T_k \in \mathbb{N}_0 : k \in \mathbb{N}_0\}$ , where  $T_0 = 0$  and for each  $k \in \mathbb{N}_0$

$$T_{k+1} = \inf\{n > T_k : S_n \leq S_{N_k}\} = N_k + \inf\{n \in \mathbb{N} : S_{N_k+n} \leq S_{N_k}\}.$$

Since for any finite  $n \in \mathbb{N}$ , the distribution of  $(X_1, \dots, X_n)$  is identical to that of  $(X_{T_k+1}, \dots, X_{T_k+n})$ , and

$$T_{k+1} - T_k = \inf\{n \in \mathbb{N} : \sum_{i=1}^n X_{T_k+i} \leq 0\},$$

we observe that  $T_k - T_{k-1}$  are *iid* for each  $k \in \mathbb{N}$ , with complementary distribution

$$\bar{F}(m) = \Pr\{T_{k+1} - T_k > m\} = \Pr\{S_1 > 0, S_2 > 0, \dots, S_m > 0\}.$$

Therefore,  $\{T_k : k \in \mathbb{N}_0\}$  is a renewal process and at each renewal instant  $\{T_k = n\}$ ,

$$S_n \leq S_{n-1}, S_n \leq S_{n-2}, \dots, S_n \leq 0.$$

Hence,  $T_k$  denotes the  $k$ th renewal instant corresponding to the random walk  $S_n$  hitting  $k$ th low. We can define the inverse counting process  $\{N_n \in \mathbb{N}_0 : n \in \mathbb{N}_0\}$  for this renewal process as

$$N_n = \sum_{j=1}^n 1_{\{T_j \leq n\}}, \quad \text{or } \{N_n \geq k\} = \{T_k \leq n\}.$$

From definition of stopping time  $N$  and duality principle, we can write

$$\Pr\{N > n\} = \Pr\{S_1 \leq 0, \dots, S_n \leq 0\} = \Pr\{S_n \leq S_{n-1}, \dots, S_n \leq 0\}.$$

The event of renewal process hitting a new low at  $n$  is same as some renewal occurring at time  $n$ . That is,

$$\sum_{n \in \mathbb{N}} 1_{\{S_n \leq S_{n-1}, S_n \leq S_{n-2}, \dots, S_n \leq S_0\}} = \sum_{n \in \mathbb{N}} \sum_{k=1}^n 1_{\{T_k = n\}} = \sum_{k \in \mathbb{N}} \sum_{n \geq k} 1_{\{T_k = n\}} = \sum_{k \in \mathbb{N}} 1_{\{T_k \leq \infty\}} = N_\infty.$$

Therefore, we can write the mean of stopping time  $N$  as

$$\mathbb{E}N = 1 + \sum_{n \in \mathbb{N}} \Pr\{N > n\} = 1 + \sum_{n \in \mathbb{N}} \sum_{k=1}^n \Pr\{T_k = n\} = 1 + \sum_{k \in \mathbb{N}} \sum_{n \geq k} \Pr\{T_k = n\} = 1 + \mathbb{E}N_\infty.$$

Since  $\mathbb{E}X_1 > 0$ , it follows from strong law of large numbers that  $S_n \rightarrow \infty$ . Hence, the expected number of renewals that occur is finite. Thus  $\mathbb{E}N < \infty$ .  $\square$

The number of distinct values of  $(S_0, \dots, S_n)$  is called **range**, denoted by  $R_n$ . We define by stopping time  $T_k$ , first return of random walk to  $k$

$$T_k = \inf\{n \in \mathbb{N} : S_n = k\}.$$

**Proposition 2.4.** For a simple random walk,

$$\lim_{n \in \mathbb{N}} \frac{\mathbb{E}R_n}{n} = \Pr\{T_0 > \infty\}.$$

*Proof.* We can define indicator function for  $S_k$  being a distinct number from  $S_1, \dots, S_{k-1}$ , as

$$I_k = 1_{\{S_k \neq S_{k-1}, \dots, S_k \neq S_1\}}.$$

Then, we can write range  $R_n$  in terms of indicator  $I_k$  as

$$R_n = 1 + \sum_{k=1}^n I_k$$

Further, using the duality principle, we can write

$$\mathbb{E}R_n = 1 + \sum_{k=1}^n \Pr\{S_1 \neq 0, \dots, S_k \neq 0\} = \sum_{k=0}^n \Pr\{T_0 > k\}$$

Result follows by dividing both sides by  $n$  and taking limits.  $\square$

## 2.1 Simple random walk

**Theorem 2.5 (range).** *For a simple random walk with  $\Pr\{X_1 = 1\} = p$ , the following holds*

$$\lim_{n \in \mathbb{N}} \frac{\mathbb{E}R_n}{n} = \begin{cases} 2p - 1, & p > \frac{1}{2} \\ 2(1 - p) - 1, & p \leq \frac{1}{2}. \end{cases}$$

*Proof.* When  $p = \frac{1}{2}$ , this random walk is recurrent and thus

$$\Pr\{T_0 > \infty\} = 0 = \lim_{n \in \mathbb{N}} \frac{\mathbb{E}R_n}{n}.$$

For  $p > \frac{1}{2}$ , let  $\alpha = \Pr\{T_0 < \infty | X_1 = 1\}$ . Since  $\mathbb{E}X > 0$ , we know that  $S_n \rightarrow \infty$  and hence  $\Pr\{T_0 < \infty | X_1 = -1\} = 1$ . We can write unconditioned probability of return of random walk to 0 as

$$\Pr\{T_0 < \infty\} = \alpha p + 1 - p.$$

Conditioning on  $X_2$  and from strong law of large numbers, we yield

$$\alpha = \Pr\{T_0 < \infty | X_1 = 1\} = p \Pr\{T_0 < \infty | S_2 = 2\} + (1 - p).$$

From Markov property and homogeneity of random walk process, it follows that

$$\Pr\{T_0 < \infty | S_2 = 2\} = \Pr\{T_0 < \infty | S_{T_1} = 1, T_1 < \infty\} \Pr\{T_1 < \infty | S_2 = 2\} = \alpha^2.$$

We conclude  $\alpha = \alpha^2 p + 1 - p$ , and since  $\alpha < 1$  due to transience, we get  $\alpha = \frac{1-p}{p}$ , and hence the result follows. We can show similarly for the case when  $p < 1/2$ .  $\square$