

# Topics in Applied Stochastic Processes

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# Lecture 1

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## 1.1 Definitions

Random walks serve as very useful models in many applications. They are simple to state and understand, yet they lead to lots of intractable questions.

**Notation.**  $\mathbb{N} = \{k \in \mathbb{Z} : k \geq 1\}$  and  $\mathbb{N}_0 = \mathbb{N} \cup \{0\}$

We now proceed to construct what is called a “simple random walk” on  $\mathbb{Z}$  of finite length  $N \in \mathbb{N}$ . The sample space  $\Omega_N$  and the event space  $\mathcal{F}_N$  are described below.

$$\Omega_N := \{(\omega_1, \dots, \omega_N) : \omega_i \in \{-1, 1\} \forall 1 \leq i \leq N\}$$

$$\mathcal{F}_N := \{A : A \subseteq \Omega_N\}$$

The probability function  $\mathbb{P}_N : \Omega_N \rightarrow [0, 1]$  is defined as

$$\mathbb{P}_N(A) := |A| 2^{-N}$$

We also define random variables  $X_k$  and  $S_k$  on  $\Omega_N$  for  $1 \leq k \leq N$  as

$$X_k : \Omega_N \rightarrow \{-1, 1\} ; X_k(\omega) := \omega_k$$

$$S_k : \Omega_N \rightarrow \mathbb{Z} ; S_k(\omega) := \sum_{i=1}^k X_i(\omega) ; S_0(\omega) := 0 \text{ for all } \omega \in \Omega_N$$

**Definition 1.1.1.** Fix  $N \in \mathbb{N}$ . The sequence of random variables  $\{S_k\}_{k=1}^N$  on  $(\Omega_N, \mathcal{F}_N, \mathbb{P}_N)$  is called a (symmetric) simple random walk on  $\mathbb{Z}$ , of finite length  $N$ , starting at 0.

In what follows, we suppress the subscript  $N$  while referring to the probability space  $(\Omega_N, \mathcal{F}_N, \mathbb{P}_N)$ , and we assume that  $N \in \mathbb{N}$  is fixed.

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<sup>†</sup> added illustrations

**Observations.**

- (a)  $\{X_k\}_{k=1}^N$  are iid, i.e. independent and identically distributed.

*Proof.*

$$\begin{aligned}\mathbb{P}(X_k = 1) &= \mathbb{P}(\{\omega \in \Omega : \omega_k = 1\}) = 2^{-N} |\{\omega \in \Omega : \omega_k = 1\}| \\ &= 2^{-N} 2^{N-1} \\ &= \frac{1}{2} \\ &= \mathbb{P}(X_k = -1)\end{aligned}$$

So  $\{X_k\}_{k=1}^N$  are identically distributed. Independence is left as an exercise.  $\square$

- (b) (Independent increments) For  $1 \leq k_1 \leq k_2 \leq \dots \leq N$ ,  $\{S_{k_i} - S_{k_{i-1}} : 1 \leq i \leq N\}$  are independent random variables.

*Proof.* Observe that, for  $1 \leq k < l \leq N$ , we have  $S_l - S_k = \sum_{i=k+1}^l X_i$ . Therefore, if  $1 \leq a < b \leq c < d \leq N$ , we see that  $S_b - S_a$  and  $S_d - S_c$  are functions of disjoint sets of independent random variables, and hence the claim is true.  $\square$

- (c) (Stationary in increments) For  $1 \leq k < m \leq N$ ,  $\mathbb{P}(S_m - S_k = \alpha) = \mathbb{P}(S_{m-k} = \alpha)$  for every  $\alpha \in \mathbb{Z}$ .

*Proof.* We use the fact that  $\{X_i\}_{i=1}^N$  are identically distributed in the following argument.

$$\mathbb{P}(S_m - S_k = \alpha) = \mathbb{P}\left(\sum_{i=k+1}^m X_i = \alpha\right) = \mathbb{P}\left(\sum_{i=1}^{m-k} X_i = \alpha\right) = \mathbb{P}(S_{m-k} = \alpha)$$

$\square$

- (d) (Markov Property) For  $\alpha_i \in \mathbb{Z}$ ,  $1 \leq i \leq N$  and  $0 \leq n \leq N$ ,

$$\mathbb{P}(S_n = \alpha_n \mid S_{n-1} = \alpha_{n-1}, \dots, S_1 = \alpha_1) = \mathbb{P}(S_n = \alpha_n \mid S_{n-1} = \alpha_{n-1}),$$

assuming (of course) that the conditional probabilities are well defined.

*Proof.* Left as an exercise.  $\square$

- (e) (Conditional Law) For  $1 \leq k < m \leq N$ ,  $\mathbb{P}(S_m = b \mid S_k = a) = \mathbb{P}(S_{m-k} = b - a)$ .

*Proof.* Left as an exercise.  $\square$

- (f) (Moments) For  $1 \leq k \leq N$ , we have  $\mathbb{E}[X_k] = \mathbb{E}[S_k] = 0$  and  $\text{Var}[S_k] = k$ .

*Proof.* By definition of expected value,  $\mathbb{E}[X_k] = 1(1/2) - 1(1/2) = 0$ . By linearity of expected values,  $\mathbb{E}[S_k] = \sum_{i=1}^k \mathbb{E}[X_i] = 0$ .

Since  $\mathbb{E}[S_k] = 0$ ,  $\text{Var}[S_k] = \mathbb{E}[(\sum_{i=1}^k X_i)^2] = \sum_{i=1}^k \mathbb{E}[X_i^2] = k$ . As an exercise, show that  $\mathbb{E}[(\sum_{i=1}^k X_i)^2] = \sum_{i=1}^k \mathbb{E}[X_i^2]$ .  $\square$

(g) (Distribution of  $S_n$ ) For  $x \in \{-n, -n+2, \dots, n-2, n\}$ , we have

$$\mathbb{P}(S_n = x) = \mathbb{P}(S_n = -x) = \binom{n}{\frac{n+x}{2}} 2^{-n}$$

*Proof.* We only provide a sketch of the proof, which is left as an exercise. For  $0 \leq j \leq N$ ,  $\{S_n = 2j - n\} = \{S_n = j - (n - j)\}$ . So there must be a total of  $j$  steps to the right and  $n - j$  steps to the left. Therefore

$$\mathbb{P}(S_n = 2j - n) = 2^{-N} |\{\omega \in \Omega : \dots\}| = 2^{-n} \binom{n}{j}$$

$\square$

(h) (Mode) The mode of the above distribution is achieved in the middle, i.e. at  $x = 0$  and at  $x = 1, -1$  for  $S_{2n}$  and  $S_{2n-1}$  respectively.

*Proof.*

$$\mathbb{P}(S_{2n} = 0) = \mathbb{P}(S_{2n-1} = 1) = \binom{2n}{n} 2^{-2n}$$

$\square$

(i) (Stirling's formula) Using Stirling's approximation, for large  $n$ , we have

$$\binom{2n}{n} = \frac{2n!}{n!n!} \sim \frac{(2n)^{2n} e^{-2n} \sqrt{4\pi n}}{n^{2n} e^{-2n} \sqrt{2\pi n} \sqrt{2\pi n}} \sim \frac{2^{2n}}{\sqrt{\pi n}} \quad (*)$$

Therefore,

$$\mathbb{P}(S_{2n} = 0) = \binom{2n}{n} \frac{1}{2^{2n}} \sim \frac{1}{\sqrt{\pi n}} \quad \text{as } n \rightarrow \infty$$

This approximation, although correct, has a caveat - we chose to keep  $N$  fixed, but as  $n \rightarrow \infty$ , we must also let  $N \rightarrow \infty$ , and this requires subtler arguments. A few consequences of this approximation are mentioned in the exercises.

## 1.2 Stopping times

Motivation for this section comes from the classic Gambler's ruin problem. We can interpret a simple random walk as a fair game between two players, where in round  $k$ , a player wins the amount  $X_k$ . Then  $S_n$  denotes the capital of one player over the other after  $n$  rounds.

We would like to answer the following question - "Is it possible to stop the game in a favorite moment, i.e., can clever stopping lead to a positive expected gain?". In other words, can we design a  $T(\omega)$  for every  $\omega \in \Omega$  such that  $\mathbb{E}[S_T] > 0$ ? Of course, the decision to stop may only depend on the trajectory until that time: no "insider knowledge" about the future of the trajectory is permitted.

To formalize this setting, we make the following definition.

**Definition 1.2.1.** An event  $A \subseteq \Omega$  is said to be observable by time  $n$  if it is a (possibly empty) union of basic / elementary events of the form

$$\{\omega \in \Omega : \omega_1 = o_1, \dots, \omega_n = o_n\}$$

where  $o_i \in \{-1, 1\}$  for  $1 \leq i \leq n$ .

We also define  $\mathcal{A}_0 = \{\phi, \Omega\}$  and set

$$\mathcal{A}_n := \{A \in \mathcal{F} : A \text{ is observable by time } n\}.$$

Immediately, we observe that

$$\{\phi, \Omega\} = \mathcal{A}_0 \subseteq \mathcal{A}_1 \subseteq \dots \subseteq \mathcal{A}_{N-1} \subseteq \mathcal{A}_N = \mathcal{F}$$

As an easy exercise, verify that each  $\mathcal{A}_n$  is closed with respect to taking complement, union and intersection. Such a sequence  $\{\mathcal{A}_i\}_{i=0}^N$  is called a *filtration*.

**Definition 1.2.2.** A function  $T : \Omega \rightarrow \{0, 1, \dots, N\} \cup \{\infty\}$  is called a *stopping time* if for each  $0 \leq n \leq N$ ,

$$\{T = n\} = \{\omega \in \Omega : T(\omega) = n\} \in \mathcal{A}_n$$

**Example.** For  $a \in \mathbb{Z}$ , let  $\sigma_a = \inf\{n : S_n = a, 0 \leq n \leq N\}$  denote the *first* hitting time of  $a$ . As an exercise, show that  $\sigma_a$  is a stopping time.

**Example.** For  $a \in \mathbb{Z}$ , let  $L_a = \max\{n : S_n = a, 0 \leq n \leq N\}$  denote the *last* hitting time of  $a$ . As an exercise, show that  $L_a$  is NOT a stopping time.

**Theorem 1.** Let  $T : \Omega \rightarrow \{0, 1, \dots, N\}$  be a stopping time. Then

$$\mathbb{E}[S_T] = 0$$

where  $S_T : \Omega \rightarrow \mathbb{Z}$  maps  $\omega \mapsto S_{T(\omega)}(\omega)$ .

*Proof.*

$$\begin{aligned}
S_T &= \sum_{k=1}^N S_k \mathbb{1}\{T = k\} = \sum_{k=1}^N S_k (\mathbb{1}\{T \geq k\} - \mathbb{1}\{T \geq k+1\}) \\
&= \sum_{k=1}^N (S_k - S_{k-1}) \mathbb{1}\{T \geq k\} \\
&= \sum_{k=1}^N X_k \mathbb{1}\{T \geq k\}
\end{aligned}$$

where we take  $\mathbb{1}\{T \geq N+1\} = 0$ . Now, we can write  $\mathbb{E}[S_T]$  as

$$\mathbb{E}[S_T] = \sum_{k=1}^N \mathbb{E}[X_k \mathbb{1}\{T \geq k\}] \quad (\dagger)$$

Observe that for  $1 \leq k \leq N$ , we have

$$X_k \mathbb{1}\{T \geq k\} = \begin{cases} 1, & \text{for } X_k = 1, T \geq k \\ -1, & \text{for } X_k = -1, T \geq k \\ 0, & \text{otherwise.} \end{cases}$$

$$\mathbb{E}[X_k \mathbb{1}\{T \geq k\}] = \mathbb{P}(X_k = 1, T \geq k) - \mathbb{P}(X_k = -1, T \geq k) \quad (\dagger\dagger)$$

Now,

$$\{T \geq k\} = \{T < k\}^c = \left( \bigcup_{l=0}^{k-1} \{T = l\} \right)^c \in \mathcal{A}_{k-1}$$

Using the fact that  $\{T \geq k\} \in \mathcal{A}_{k-1}$ , one can show that (details left as an exercise)

$$\mathbb{P}(X_k = 1, T \geq k) = \mathbb{P}(X_k = -1, T \geq k) = \frac{1}{2} \mathbb{P}(T \geq k)$$

Substituting the above values in  $(\dagger)$  and  $(\dagger\dagger)$ , we finally have

$$\mathbb{E}[S_T] = 0$$

□

As an exercise, compute  $\text{Var}[S_T]$ .

**Definition 1.2.3.** A bet sequence / game system is a sequence of random variables  $V_k : \Omega \rightarrow \mathbb{R}$  such that

$$\{V_k = c\} \in \mathcal{A}_{k-1} \text{ for every } c \in \mathbb{R} \text{ and } 1 \leq k \leq N$$

**Theorem 2.** Let  $\{V_k\}_{k=1}^N$  be a bet sequence. Then

$$\mathbb{E}[S_N^V] = 0 \quad \text{where} \quad S_N^V = \sum_{k=1}^N V_k X_k$$

In this setting,  $S_N^V$  is interpreted as the “total gain”.

*Proof.* Since  $\Omega$  is finite, we may write

$$\text{Range}(V_k) = \{c_i^k : 1 \leq i \leq m_k\} \text{ where } c_i^k \in \mathbb{R}$$

$$V_k = \sum_{i=1}^{m_k} c_i^k \mathbb{1}\{V_k = c_i^k\}$$

Now, since  $\mathbb{E}[X_k] = 0$ , and since  $X_k \perp \mathbb{1}\{V_k = c_i^k\}$ , we get

$$\begin{aligned} \mathbb{E}[S_N^V] &= \sum_{k=1}^N \mathbb{E}[V_k X_k] = \sum_{k=1}^N \mathbb{E}\left[X_k \sum_{i=1}^{m_k} c_i^k \mathbb{1}\{V_k = c_i^k\}\right] \\ &= \sum_{k=1}^N \sum_{i=1}^{m_k} c_i^k \mathbb{E}[X_k \mathbb{1}\{V_k = c_i^k\}] \\ &= \sum_{k=1}^N \sum_{i=1}^{m_k} c_i^k \mathbb{E}[X_k] \mathbb{P}(V_k = c_i^k) \\ &= 0 \end{aligned}$$

□

### 1.3 Exercises

1. Show that  $\{X_k\}_{k=1}^N$  are independent.
2. Show that  $\{S_n\}_{n=0}^N$  satisfies the Markov property.
3. For  $1 \leq k < m \leq N$ ,  $\mathbb{P}(S_m = b \mid S_k = a) = \mathbb{P}(S_{m-k} = b - a)$ .
4. Show that  $\mathbb{E}[S_n^2] = \sum_{i=1}^n \mathbb{E}[X_i^2]$ .
5. (a) Show that for any  $a, b \in \mathbb{R}$ ,

$$\mathbb{P}(a \leq S_n < b) \leq (b - a) \mathbb{P}(S_n \in \{-1, 0, 1\}).$$

- (b) Using (a), conclude that

$$\mathbb{P}(a \leq S_n < b) \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$

Thus, we observe that the walk exits any finite interval as  $n \rightarrow \infty$ .

6. Verify that each  $\mathcal{A}_n$ ,  $0 \leq n \leq N$ , is closed with respect to taking complement, union and intersection.
7. For  $a \in \mathbb{Z}$ , let  $\sigma_a = \inf\{n : S_n = a, 0 \leq n \leq N\}$ . Show that  $\sigma_a$  is a stopping time.
8. For  $a \in \mathbb{Z}$ , let  $L_a = \max\{n : S_n = a, 0 \leq n \leq N\}$ . Show that  $L_a$  is not a stopping time.
9. Let  $T : \Omega \rightarrow \{0, 1, \dots, N\}$  be a stopping time. Compute  $\text{Var}[S_T]$ .
10. Show that  $X_k$  and  $\mathbb{1}\{T \geq k\}$  are independent.