

# Lecture-18: Stopping Times

## 1 Stopping times

Let  $(\Omega, \mathcal{F}, P)$  be a probability space. Consider a random process  $X : \Omega \rightarrow \mathcal{X}^T$  defined on this probability space with state space  $\mathcal{X} \subseteq \mathbb{R}$  and ordered index set  $T \subseteq \mathbb{R}$  considered as time.

**Definition 1.1.** A collection of event spaces denoted  $\mathcal{F}_\bullet = (\mathcal{F}_t \subseteq \mathcal{F} : t \in T)$  is called a filtration if  $\mathcal{F}_s \subseteq \mathcal{F}_t$  for all  $s \leq t$ .

*Remark 1.* For the random process  $X : \Omega \rightarrow \mathcal{X}^T$ , we can find the event space generated by all random variables until time  $t$  as  $\mathcal{G}_t \triangleq \sigma(X_s, s \leq t)$ . The collection of event spaces  $\mathcal{G}_\bullet = (\mathcal{G}_t : t \in T)$  is a filtration.

**Definition 1.2.** The natural filtration associated with a random process  $X : \Omega \rightarrow \mathcal{X}^T$  is given by  $\mathcal{F}_\bullet = (\mathcal{F}_t : t \in T)$  where  $\mathcal{F}_t \triangleq \sigma(X_s, s \leq t)$ .

*Remark 2.* For a random sequence  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ , the natural filtration is a sequence  $\mathcal{F}_\bullet = (\mathcal{F}_n : n \in \mathbb{N})$  of event spaces  $\mathcal{F}_n \triangleq \sigma(X_1, \dots, X_n)$  for all  $n \in \mathbb{N}$ .

*Remark 3.* If the random sequence  $X$  is independent, then the random sequence  $(X_{n+j} : j \in \mathbb{N})$  is independent of the event space  $\sigma(X_1, \dots, X_n)$ .

**Example 1.3.** For a random walk  $S$  with step size sequence  $X$ , the natural filtration of the random walk is identical to that of the step size sequence. That is,  $\sigma(S_1, \dots, S_n) = \sigma(X_1, \dots, X_n)$  for all  $n \in \mathbb{N}$ . This follows from the fact that for all  $n \in \mathbb{N}$ , we can write  $S_j = \sum_{i=1}^j X_i$  and  $X_j = S_j - S_{j-1}$  for all  $j \in [n]$ . That is, there is a bijection between  $(X_1, \dots, X_n)$  and  $(S_1, \dots, S_n)$ .

**Definition 1.4.** A random variable  $\tau : \Omega \rightarrow T$  is called a **stopping time** with respect to a filtration  $\mathcal{F}_\bullet$  if

- (a) the event  $\tau^{-1}(-\infty, t] \in \mathcal{F}_t$  for all  $t \in T$ , and
- (b) the random variable  $\tau$  is finite almost surely, i.e.  $P\{\tau < \infty\} = 1$ .

*Remark 4.* Intuitively, if we observe the process  $X$  sequentially, then the event  $\{\tau \leq t\}$  can be completely determined by the observation  $(X_s, s \leq t)$  until time  $t$ . The intuition behind a stopping time is that its realization is determined by the past and present events but not by future events. That is, given the history of the process until time  $t$ , we can tell whether the stopping time is less than or equal to  $t$  or not. In particular,  $\mathbb{E}[\mathbb{1}_{\{\tau \leq t\}} \mid \mathcal{F}_t] = \mathbb{1}_{\{\tau \leq t\}}$  is either one or zero.

**Definition 1.5 (First hitting time).** For a process  $X : \Omega \rightarrow \mathcal{X}^T$  and any Borel measurable set  $A \in \mathcal{B}(\mathcal{X})$ , we define the **first hitting time**  $\tau_X^A : \Omega \rightarrow T \cup \{\infty\}$  for the process  $X$  to hit states in  $A$ , as  $\tau_X^A \triangleq \inf\{t \in T : X_t \in A\}$ .

**Example 1.6.** We observe that the event  $\{\tau_X^A \leq t\} = \{X_s \in A \text{ for some } s \leq t\} \in \mathcal{F}_t$  for all  $t \in T$ . It follows that, if  $\tau_A$  is finite almost surely, then  $\tau_A$  is a stopping time with respect to filtration  $\mathcal{F}_\bullet$ .

**Theorem 1.7.** For a random sequence  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ , an almost sure finite discrete random variable  $\tau : \Omega \rightarrow \mathbb{N} \cup \{\infty\}$  is a **stopping time** with respect to this random sequence  $X$  iff the event  $\{\tau = n\} \in \sigma(X_1, \dots, X_n)$  for all  $n \in \mathbb{N}$ .

*Proof.* From Definition 1.4, we have  $\{\tau = n\} = \{\tau \leq n\} \setminus \{\tau \leq n-1\} \in \sigma(X_1, \dots, X_n)$ . Conversely, from the theorem hypothesis, it follows that  $\{\tau \leq n\} = \cup_{m=1}^n \{\tau = m\} \in \sigma(X_1, \dots, X_n)$ .  $\square$

**Example 1.8.** Consider a random sequence  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ , with the natural filtration  $\mathcal{F}_\bullet$ , and a measurable set  $A \in \mathcal{B}(\mathcal{X})$ . If the first hitting time  $\tau_X^A : \Omega \rightarrow \mathbb{N} \cup \{\infty\}$  for the sequence  $X$  to hit set  $A$  is almost surely finite, then  $\tau_X^A$  is a hitting time. This follows from the fact that  $\{\tau_X^A = n\} = \bigcap_{k=1}^{n-1} \{X_k \notin A\} \cap \{X_n \in A\} \in \mathcal{F}_n$  for each  $n \in \mathbb{N}$ .

**Definition 1.9.** Consider a random process  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{R}^+}$  with discrete state space  $\mathcal{X} \subseteq \mathbb{R}$ . For each state  $y \in \mathcal{X}$ , we define  $\tau_X^{\{y\},0} \triangleq 0$  and inductively define the  $k$ th hitting time to a state  $y$  after time  $t = 0$ , as

$$\tau_X^{\{y\},k} \triangleq \inf \left\{ t > \tau_X^{\{y\},k-1} : X_t = y \right\}, \quad k \in \mathbb{N}.$$

*Remark 5.* We observe that  $\{\tau_X^{\{y\},k} \leq t\} \in \mathcal{F}_t$  for all times  $t \in \mathbb{R}_+$ . Hence if  $\tau_X^{\{y\},k}$  is almost surely finite, then it is a stopping time for the process  $X$ .

**Definition 1.10.** For a discrete valued random sequence  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{N}}$ , the number of visits to a state  $y \in \mathcal{X}$  in first  $n$  time steps is defined as  $N_y(n) \triangleq \sum_{k=1}^n \mathbb{1}_{\{X_k=y\}}$  for all  $n \in \mathbb{N}$ .

*Remark 6.* We observe that  $N_y : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{Z}_+}$  is a random walk with the Bernoulli step size sequence  $(\mathbb{1}_{\{X_k=y\}} : k \in \mathbb{N})$ . Further,  $\tau_X^{\{y\},k} = \tau_{N_y}^{\{k\}} = \inf \{n \in \mathbb{N} : N_y(n) = k\}$ . We also observe that  $\{N_y(n) \leq k\} = \{\tau_y^{(k+1)} > n\}$  and  $\{N_y(n) = k\} = \{\tau_y^k \leq n < \tau_y^{(k+1)}\}$ .

*Remark 7.* We observe that number of visits to state  $y$  in first  $n$  steps of  $X$  is also given by

$$N_y(n) = \sup \left\{ k \in \mathbb{Z}_+ : \tau_X^{\{y\},k} \leq n \right\} = \inf \left\{ k \in \mathbb{N} : \tau_X^{\{y\},k} > n \right\} - 1 = \sum_{k \in \mathbb{N}} \mathbb{1}_{\{\tau_X^{\{y\},k} \leq n\}}.$$

This implies that  $N_y(n) + 1$  is the first hitting time to state  $(n + 1)$  for the increasing random sequence  $(\tau_X^{\{y\},k} : k \in \mathbb{N})$ .

**Lemma 1.11 (Wald's Lemma).** Consider a random walk  $S : \Omega \rightarrow \mathbb{R}^{\mathbb{Z}_+}$  with i.i.d. step-sizes  $X : \Omega \rightarrow \mathbb{R}^{\mathbb{N}}$  having finite  $\mathbb{E}|X_1|$ . Let  $\tau$  be a finite mean stopping time with respect to this random walk. Then,

$$\mathbb{E}[S_\tau] = \mathbb{E}[X_1] \mathbb{E}[\tau].$$

*Proof.* Recall that the event space generate by the random walk and the step-sizes are identical. From the independence of step sizes, it follows that  $X_n$  is independent of  $\sigma(X_0, X_1, \dots, X_{n-1})$ . Since  $\tau$  is a stopping time with respect to random walk  $S$ , we observe that  $\{\tau \geq n\} = \{\tau > n - 1\} \in \sigma(X_0, X_1, \dots, X_{n-1})$ , and hence it follows that random variable  $X_n$  and indicator  $\mathbb{1}_{\{\tau \geq n\}}$  are independent and  $\mathbb{E}[X_n \mathbb{1}_{\{\tau \geq n\}}] = \mathbb{E}X_1 \mathbb{E} \mathbb{1}_{\{\tau \geq n\}}$ . Therefore,

$$\mathbb{E}[S_\tau] = \mathbb{E}\left[\sum_{n=1}^{\tau} X_n\right] = \mathbb{E}\left[\sum_{n \in \mathbb{N}} X_n \mathbb{1}_{\{\tau \geq n\}}\right] = \sum_{n \in \mathbb{N}} \mathbb{E}X_1 \mathbb{E}\left[\mathbb{1}_{\{\tau \geq n\}}\right] = \mathbb{E}X_1 \mathbb{E}\left[\sum_{n \in \mathbb{N}} \mathbb{1}_{\{\tau \geq n\}}\right] = \mathbb{E}[X_1] \mathbb{E}[\tau].$$

We exchanged limit and expectation in the above step, which is not always allowed. We were able to do it by the application of dominated convergence theorem.  $\square$

**Corollary 1.12.** Consider the stopping time  $\tau_S^{\{i\}} \triangleq \min \{n \in \mathbb{N} : S_n = i\}$  for an integer random walk  $S : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{Z}_+}$  with i.i.d. step size sequence  $X : \Omega \rightarrow \mathbb{Z}^{\mathbb{N}}$ . Then, the mean of stopping time  $\mathbb{E}\tau_S^{\{i\}} = i/\mathbb{E}X_1$ .

*Proof.* This follows from the Wald's Lemma and the fact that  $S_{\tau_i} = i$ .  $\square$

## 1.1 Properties of stopping time

**Lemma 1.13.** Let  $\tau_1, \tau_2$  be two stopping times with respect to filtration  $\mathcal{F}_\bullet$ . Then the following hold true.

- i.  $\min \{\tau_1, \tau_2\}$  is a stopping time.
- ii. If  $T$  is separable, then  $\tau_1 + \tau_2$  is a stopping time.

*Proof.* Let  $\tau_1, \tau_2$  be stopping times with respect to a filtration  $\mathcal{F}_\bullet = (\mathcal{F}_t : t \in T)$ .

- i. Result follows since the event  $\{\min\{\tau_1, \tau_2\} > t\} = \{\tau_1 > t\} \cap \{\tau_2 > t\} \in \mathcal{F}_t$ .
- ii. A topological space is called if it contains a countable dense set. Since  $\mathbb{R}_+$  is separable and ordered, we assume  $T = \mathbb{R}_+$  without any loss of generality. It suffices to show that the event  $\{\tau_1 + \tau_2 \leq t\} \in \mathcal{F}_t$  for  $T = \mathbb{R}_+$ . To this end, we observe that  $\{\tau_1 + \tau_2 \leq t\} = \bigcup_{s \in \mathbb{Q}_+ : s \leq t} \{\tau_1 \leq t - s, \tau_2 \leq s\} \in \mathcal{F}_t$ .

□

**Theorem 1.14 (Strong independence property).** Let  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{R}_+}$  be an independent random process with natural filtration  $\mathcal{F}_\bullet$ , and  $\tau : \Omega \rightarrow \mathbb{R}_+$  a stopping time with respect to  $\mathcal{F}_\bullet$ , then  $(X_{\tau+s} : s \in \mathbb{R}_+)$  is independent of history  $(X_s : s \leq \tau)$ .

**Definition 1.15.** We can define the  $k$ th return time to state  $y$  for the random process  $X : \Omega \rightarrow \mathcal{X}^{\mathbb{R}_+}$  as the interval between two successive visits to state  $y$ , that is for all  $k \in \mathbb{N}$

$$H_X^{\{y\},k} \triangleq \tau_X^{\{y\},k} - \tau_X^{\{y\},k-1} = \inf \left\{ s \in \mathbb{R}_+ : X_{\tau_X^{\{y\},k-1} + s} = y \right\}.$$

*Remark 8.* We observe that  $\tau_X^{\{y\},k} = \sum_{j=1}^k H_X^{\{y\},j}$ . Therefore, if  $H_X^{\{y\},j}$  is almost sure finite for all  $j \in [k]$ , then the finite sum  $\tau_X^{\{y\},k}$  is almost sure finite.

*Remark 9.* From the bijection between hitting and return times, we observe that  $\sigma(\tau_X^{\{y\},k} : k \in [n]) = \sigma(H_X^{\{y\},j} : j \in [n])$ . If passage times are independent and *i.i.d.* from second passage time, then it follows from the Wald's Lemma that

$$\mathbb{E}H_X^{\{y\},1} + \mathbb{E}[N_y(n)]\mathbb{E}H_X^{\{y\},2} = (n+1)$$

**Example 1.16.** We also observe that the  $k$ th hitting time to  $\{1\}$  by a Bernoulli step size sequence  $X : \Omega \rightarrow \{0,1\}^{\mathbb{N}}$  is the first hitting time to  $\{k\}$  by random walk  $S : \Omega \rightarrow \mathbb{Z}_+^{\mathbb{N}}$ . That is,

$$\tau_X^{\{1\},k} = \tau_S^{\{k\}} = \inf \{n \in \mathbb{N} : S_n = k\} = \tau_S^{\{k-1\}} + \inf \left\{ n \in \mathbb{N} : S_{\tau_S^{\{k-1\}} + n} - S_{\tau_S^{\{k-1\}}} = 1 \right\}.$$

**Lemma 1.17.** For an *i.i.d.* Bernoulli random sequence  $X : \Omega \rightarrow \{0,1\}^{\mathbb{N}}$  where  $\mathbb{E}X_1 \in (0,1)$ , then the  $k$ th hitting time to state 1 is a stopping time, and  $\tau_X^{\{1\},k} = \sum_{i=1}^k Y_i$  where  $Y : \Omega \rightarrow \mathbb{N}^{\mathbb{N}}$  is an *i.i.d.* random sequence distributed identically to  $\tau_X^{\{1\}}$ .

*Proof.* When the Bernoulli step size sequence  $X$  is *i.i.d.* with  $\mathbb{E}X_1 = p \in (0,1)$ , we get that  $P\{\tau_S^{\{1\}} = n\} = (1-p)^{n-1}p$  for all  $n \in \mathbb{N}$ . It follows that

$$P\{\tau_S^{\{1\}} < \infty\} = P\left(\bigcup_{n \in \mathbb{N}} \{\tau_S^{\{1\}} = n\}\right) = \sum_{n \in \mathbb{N}} P\{\tau_S^{\{1\}} = n\} = 1.$$

Hence, the random time  $\tau_S^{\{1\}}$  is finite almost surely. We will show that  $\tau_S^{\{k\}}$  is finite almost surely for all  $k \in \mathbb{N}$  by induction. By induction hypothesis,  $\tau_S^{\{k-1\}}$  is finite almost surely. Then  $S_{\tau_S^{\{k-1\}} + n} - S_{\tau_S^{\{k-1\}}} = \sum_{j=1}^n X_{\tau_S^{\{k-1\}} + j}$  is the sum of  $n$  *i.i.d.* Bernoulli random variables, and hence has distribution identical to  $S_n$ . Further, This implies that  $\tau_S^{\{k\}} = \tau_S^{\{k-1\}} + \tau_S^{\{1\}}$ , where  $\tau_S^{\{1\}}$  has the identical distribution to  $\tau_X^{\{1\}}$  and is independent of  $\tau_S^{\{k-1\}}$ . □