

### Definition of stopping time

Let a random process,  $\{X_t, t \in T\}$  defined on some probability space and taking values in the set of integers,  $D$ . Random variable  $\tau$  is said to be a stopping time w.r.t.  $\{X_t, t \in T\}$  if the event  $\{\tau = m\}, \forall m \in D$  can be determined by  $X_0, X_1, \dots, X_m$ .

### Definition of strong Markov property

Let  $\{X_t, t \in T\}$  be a Markov process and let  $\tau$  be a stopping time w.r.t.  $\{X_t, t \in T\}$ . The process  $\{X_t, t \in T\}$  satisfies the strong Markov property if  $\forall k = 1, 2, \dots$ ,

$$P(X_{\tau+k} \in A | X_\tau = x, X_{\tau-1} = x, \dots, X_0 = x) = P(X_{\tau+k} \in A | X_\tau = x)$$

(note that when  $\tau$  is a constant, it goes back to Markov property)

### Definition of Markov Chain

A Markov chain is a Markov process  $\{X_t, t \in T\}$  with finite or countably infinite state space. To characterize a Markov chain statistically, we need the following

$$\begin{aligned} \{\pi_s(x) : \forall x \in S, \pi_0(x) = p(X_0 = 0)\} \\ \{P_t(x, y) := p(X_{t+1} = y | X_t = x), x, y \in S, t \in T\} \end{aligned}$$

where  $S$  is the state of space of the Markov chain.

### Definition of Random Walk

Consider  $\{\xi_1, \dots, \xi_k, \dots\}$ , a collection of i.i.d. random variables taking values in a set of integers. Let  $X_t$  be a random variable that takes values in the set of integers and is independent of  $\{\xi_1, \dots, \xi_k, \dots\}$ ,  $\{X_t := X_0 + \xi_1 + \dots + \xi_t\}$  is called a random walk.

### Definition of hitting times

Considering a Markov Chain,  $\{X_t, t \geq 0\}$  with state space  $S$  and transition probability,  $P(x, y) = P(X_{t+1} = y | X_t = x), \forall x, y \in S, \forall t$ . Let  $A \subset S$ , a stopping time  $T_A$  w.r.t.  $\{X_t, t = 0, 1, 2, \dots\}$  is defined as

$$T_A := \min\{t > 0 : X_t \in A\}.$$

### Properties of Markov Chain

1.  $P^n(x, y) = \sum_{m=1}^n P(T_y = m | X_0 = x) P^{n-m}(y, y);$

2. If  $a$  is an absorbing state, then

$$P(X_n = a | X_0 = x) = P(T_a \leq n | X_0 = x)$$

### Definition of transient and recurrent states

Let  $\rho_{xy} := P(T_y < +\infty | X_0 = x)$ ,  $\rho_{yy} = P(T_y < +\infty | X_0 = y)$ ,  $x, y \in S$ . A state  $y \in S$  is said to be recurrent if  $\rho_{yy} = 1$ . A state  $y \in S$  is transient if  $\rho_{yy} < 1$ .

### State and prove the properties of a transient state

Let  $y \in S$  be a transient state, then

$$P(N(y) < +\infty | X_0 = x) = 1$$

$$G(x, y) = \mathbf{E}(N(y) | X_0 = x) = \frac{\rho_{xy}}{1 - \rho_{yy}}, \forall x \in S$$

where  $N(y) := \sum_{n=1}^{+\infty} \mathbf{1}_{\{y\}}(X_n)$

### State and prove properties of a recurrent state

Let  $y \in S$  be a recurrent state, then

$$P(N(y) = +\infty | X_0 = y) = 1$$

$$G(y, y) = \mathbf{E}(N(y) | X_0 = y) = +\infty$$

$$P(N(y) < +\infty | X_0 = x) = \rho_{xy}, \forall x \in S$$

where  $N(y) := \sum_{n=1}^{+\infty} \mathbf{1}_{\{y\}}(X_n)$ .

If  $\rho_{xy} = 0$ , then  $G(x, y) = 0$ ; if  $\rho_{xy} > 0$ , then  $G(x, y) = +\infty$ .