## Lecture 28: Random walks

## 1 GI/GI/1 Queueing Model

Consider a GI/GI/1 queue. Customers arrive in accordance with a renewal process having an arbitrary interarrival distribution F, and the service distribution G.

**Proposition 1.1.** Let  $D_n$  be the delay in the queue of the  $n^{th}$  customer in a GI/GI/1 queue with independent inter-arrival times  $X_n$  and service times  $Y_n$ . Let  $S_n$  be a random walk with iid steps  $U_n = Y_n - X_{n+1}$  for all  $n \in \mathbb{N}$ . Then, we can write

$$\Pr\{D_{n+1} \ge c\} = \Pr\{S_j \ge c, \text{ for some } j \in [n]\}.$$
(1)

*Proof.* The following recursion for  $D_n$  is easy to verify

$$D_{n+1} = (D_n + Y_n - X_{n+1}) \mathbf{1}_{\{D_n + Y_n - X_{n+1} \ge 0\}} = \max\{0, D_n + U_n\}.$$

Iterating the above relation with  $D_1 = 0$  yields

$$D_{n+1} = \max\{0, U_n + \max\{0, D_{n-1} + U_{n-1}\}\} = \max\{0, U_n, U_n + U_{n-1} + D_{n-1}\}.$$

For the random walk  $S_n$  with steps  $U_n$ , we can write delay in terms of random walk  $S_n$  as

$$D_{n+1} = \max\{0, S_n - S_{n-1}, S_n - S_{n-2}, \dots, S_n - S_0\}.$$

Using the duality principle, we can rewrite the following equality for delay in distribution

$$D_{n+1} = \max\{0, S_1, S_2, \dots, S_n\}.$$

**Corollary 1.2.** If  $\mathbb{E}U_n \ge 0$ , then for all c, we have  $\Pr\{D_\infty \ge c\} \triangleq \lim_{n \in \mathbb{N}} \Pr\{D_n \ge c\} = 1$ .

*Proof.* It follows from Proposition 1.1 that  $Pr\{D_{n+1} \ge c\}$  is nondecreasing in *n*. Hence, by MCT the limit exists and is denoted by  $Pr\{D_{\infty} \ge c\} = \lim_{n \in \mathbb{N}} Pr\{D_n \ge c\}$ . Therefore, by continuity of probability, we have from (1), that

$$\Pr\{D_{\infty} \ge c\} = \Pr\{S_n \ge c \text{ for some } n\}.$$
(2)

If  $\mathbb{E}U_n = \mathbb{E}Y_n - \mathbb{E}X_{n+1}$  is positive, then by strong law of large numbers the random walk  $S_n$  will converge to positive infinity with probability 1. The above will also be true when  $E[U_n] = 0$ , then the random walk is recurrent.

*Remark* 1.3. Hence, we get that  $\mathbb{E}Y_n < \mathbb{E}X_{n+1}$  implies the existence of a stationary distribution.

**Proposition 1.4 (Spitzer's Identity).** Let  $M_n = \max\{0, S_1, S_2, \dots, S_n\}$  for  $n \in \mathbb{N}$ , then

$$\mathbb{E}M_n = \sum_{k=1}^n \frac{1}{k} \mathbb{E}S_k^+.$$

*Proof.* We can decompose  $M_n$  as

$$M_n = 1_{\{S_n > 0\}} M_n + 1_{\{S_n \le 0\}} M_n.$$

We can rewrite first term in decomposition as,

$$1_{\{S_n>0\}}M_n = 1_{\{S_n>0\}} \max_{i\in[n]} S_i = 1_{\{S_n>0\}} (X_1 + \max\{0, S_2 - S_1, \dots, S_n - S_1\})$$

Hence, taking expectation and using exchangeability, we get

$$\mathbb{E}[M_n \mathbb{1}_{\{S_n > 0\}}] = \mathbb{E}[X_1 \mathbb{1}_{\{S_n > 0\}}] + \mathbb{E}[M_{n-1} \mathbb{1}_{\{S_n > 0\}}].$$

Since  $X_i$ ,  $S_n$  has the same joint distribution for all i,

$$\mathbb{E}S_n^+ = \mathbb{E}[S_n 1_{\{S_n > 0\}}] = \mathbb{E}\sum_{i=1}^n X_i 1_{\{S_n > 0\}}] = n\mathbb{E}[X_1 1_{\{S_n > 0\}}].$$

Therefore, it follows that

$$\mathbb{E}[1_{\{S_n>0\}}M_n] = \mathbb{E}[1_{\{S_n>0\}}M_{n-1}] + \frac{1}{n}\mathbb{E}[S_n^+].$$

Also,  $S_n \leq 0$  implies that  $M_n = M_{n-1}$ , it follows that

$$1_{\{S_n \le 0\}} M_n = 1_{\{S_n \le 0\}} M_{n-1}.$$

Thus, we obtain the following recursion,

$$\mathbb{E}[M_n] = \mathbb{E}[M_{n-1}] + \frac{1}{n}\mathbb{E}[S_n^+].$$

Result follow from the fact that  $M_1 = S_1^+$ .

*Remark* 1.5. Since  $D_{n+1} = M_n$ , we have  $\mathbb{E}[D_{n+1}] = \mathbb{E}[M_n] = \sum_{k=1}^n \frac{1}{k} \mathbb{E}[S_k^+]$ .

## 2 Martingales for Random Walks

**Proposition 2.1.** A random walk  $S_n$  with step size  $X_n \in [-M,M] \cap \mathbb{Z}$  for some finite M is a recurrent DTMC iff  $\mathbb{E}X = 0$ .

*Proof.* If  $\mathbb{E}X \neq 0$ , the random walk is clearly transient since, it will diverge to  $\pm \infty$  depending on the sign of  $\mathbb{E}X$ . Conversely, if  $\mathbb{E}X = 0$ , then  $S_n$  is a martingale. Assume that the process starts in state *i*. We define

$$A = \{-M, -M+1, \cdots, -2, -1\}, \qquad A_j = \{j+1, \dots, j+M\}, \ j > i.$$

Let *N* denote the hitting time to *A* or *A<sub>j</sub>* by random walk *S<sub>n</sub>*. Since *N* is a stopping time and  $S_{N \wedge n} \leq |M| + j$ , by optional stopping theorem, we have

$$\mathbb{E}_i[S_N] = \mathbb{E}_i[S_0] = i.$$

Thus we have

$$i = \mathbb{E}_i[S_N] \ge -M\mathbb{P}_i\{S_N \in A\} + j(1 - \mathbb{P}_i\{S_N \in A\}).$$

Rearranging this, we get a bound on probability of random walk  $S_n$  hitting A over  $A_i$  as

$$\mathbb{P}_i \{ S_n \in A \text{ for some } n \} \ge \mathbb{P}_i \{ S_N \in A \} \ge \frac{j-i}{j+M}$$

Taking limit  $j \to \infty$ , we see that for any  $i \ge 0$ , we have  $\mathbb{P}_i \{S_n \in A \text{ for some } n\} = 1$ . Similarly, taking  $B = \{1, 2, \dots, M\}$ , we can show that for any  $i \ge 0$ ,  $\mathbb{P}_i \{S_n \in B \text{ for some } n\} = 1$ . Result follows from combining the above two arguments to see that for any  $i \ge 0$ ,

$$\mathbb{P}_i \{ S_n \in A \cup B \text{ for some } n \} = 1.$$

**Proposition 2.2.** Consider a random walk  $S_n$  with mean step size  $\mathbb{E}[X] \neq 0$ . For A, B > 0, let  $P_A$  denote the probability that the walk hits a value greater than A before it hits a value less than -B. Then, for  $\theta \neq 0$  such that  $\mathbb{E}e^{\theta X_1} = 1$ , we have

$$P_A \approx rac{1-e^{- heta B}}{e^{ heta A}-e^{- heta B}}$$

Approximation is an equality when step size is unity and A and B are integer valued.

*Proof.* For any A, B > 0, we can define stopping times

$$T_A = \inf\{n \in \mathbb{N} : S_n \ge a\}, \qquad T_{-B} = \inf\{n \in \mathbb{N} : S_n \le -B\}.$$

We are interested in computing the probability

$$P_A = \Pr\{T_A < T_{-B}\}.$$

Now let  $Z_n = e^{\theta S_n}$ . We can see that  $Z_n$  is a martingale with mean 1. Define a stopping time  $N = T_A \wedge T_{-B}$ . From Doob's Theorem,  $\mathbb{E}[e^{S_N}] = 1$ . Thus we get

$$1 = \mathbb{E}[e^{\theta S_N} | S_N \ge A] P_A + \mathbb{E}[e^{\theta S_N} | S_N \le -B](1 - P_A).$$

We can obtain an approximation for  $P_A$  by neglecting the overshoots past A or -B. Thus we get

$$\mathbb{E}[e^{\theta S_N}|S_N \ge A] \approx e^{\theta A}, \qquad \qquad E[e^{\theta S_N}|S_N \le -B] \approx e^{-\theta B}.$$

The result follows.