

**Solution to Problem Set 7**

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**Exercise 1.**

(a) For every  $s \in [-1, 1]$ , we have

$$\left| \sum_{j=0}^{+\infty} p_j s^j \right| \leq \sum_{j=0}^{+\infty} |p_j s^j| = \sum_{j=0}^{+\infty} p_j |s^j| \leq \sum_{j=0}^{+\infty} p_j = 1,$$

and thus the series in the statement is dominated by a convergent series that does not depend on  $s$ . Therefore, the series converges uniformly.

(b) Since  $Z$  is a discrete random variable, for every  $s \in [-1, 1]$  we have

$$\mathbb{E}[s^Z] = \sum_{j=0}^{+\infty} s^j \mathbb{P}\{Z = j\} = \sum_{j=0}^{+\infty} p_j s^j = g_Z(s).$$

(c) Using (b) above and the independence of  $Z$  and  $Z'$ , it follows that

$$g_{Z+Z'}(s) = \mathbb{E}[s^{Z+Z'}] = \mathbb{E}[s^Z s^{Z'}] = \mathbb{E}[s^Z] \mathbb{E}[s^{Z'}] = g_Z(s) g_{Z'}(s).$$

(d) Since  $\sum_{j=0}^{+\infty} j p_j < +\infty$ , by dominated convergence theorem we can differentiate term by term, which gives

$$g'_Z(s) = \sum_{j=1}^{+\infty} j p_j s^{j-1}$$

and

$$g'_Z(1) = \sum_{j=1}^{+\infty} j p_j = \sum_{j=1}^{+\infty} j \mathbb{P}(Z = j) = \mathbb{E}[Z].$$

Similarly, note that if  $\sum_{j=0}^{+\infty} j^2 p_j < +\infty$ , then  $\sum_{j=0}^{+\infty} j p_j$  converges, as does  $\sum_{j=0}^{+\infty} j(j-1)p_j$ . Consequently,

$$g''_Z(s) = \sum_{j=2}^{+\infty} j(j-1) p_j s^{j-2}$$

and

$$g''_Z(1) = \sum_{j=2}^{+\infty} j(j-1) p_j = \mathbb{E}[Z(Z-1)].$$

Since  $\text{Var}(Z) = \mathbb{E}[Z^2] - \mathbb{E}[Z]^2$ , we then obtain

$$\begin{aligned} \text{Var}(Z) &= \mathbb{E}[Z(Z-1)] + \mathbb{E}[Z] - \mathbb{E}[Z]^2 \\ &= g''_Z(1) + \mathbb{E}[Z] - \mathbb{E}[Z]^2. \end{aligned}$$

### Exercise 2.

Given that  $\alpha \leq 1$ , we have

$$\mathbb{E}[|S_n|] = \mathbb{E}[\alpha^{X_n}] = \sum_{j \geq 0} \mathbb{P}(X_n = j) \alpha^j \leq 1.$$

Recall that  $\alpha$  is the solution of the equation  $g_Z(s) = s$ . Suppose  $X_0 = i_0, \dots, X_n = i_n$  and denote by  $Z_1, \dots, Z_{i_n}$  the  $i_n$  random variables whose sum defines  $X_{n+1}$ . Then, using the fact that  $(X_n)$  is a branching process, that the random variables  $Z_1, \dots, Z_{i_n}$  are mutually independent and also independent of  $X_n$ , and that they share the same distribution, we obtain

$$\begin{aligned} \mathbb{E}[S_{n+1} \mid X_0 = i_0, \dots, X_n = i_n] &= \mathbb{E}[\alpha^{X_{n+1}} \mid X_0 = i_0, \dots, X_n = i_n] \\ &= \mathbb{E}[\alpha^{Z_1 + \dots + Z_{i_n}} \mid X_0 = i_0, \dots, X_n = i_n] \\ &= \mathbb{E}[\alpha^{Z_1 + \dots + Z_{i_n}}] \\ &= \prod_{i=1}^{i_n} \mathbb{E}[\alpha^{Z_i}] \\ &= (g_Z(\alpha))^{i_n} = \alpha^{i_n}, \end{aligned}$$

which means that  $\mathbb{E}[S_{n+1} \mid X_0, \dots, X_n] = \alpha^{X_n} = S_n$ . This proves the result.

### Exercise 3.

This problem can be modeled using a branching process. Suppose that at time 1, there are  $Z_1$  customers in line. They form generation 1. The direct descendants of a customer are those who arrive while that customer is being served. Generation  $n+1$  is formed by the direct descendants of the customers in generation  $n$ . Thus, each of the  $X_n$  customers in generation  $n$  is served during one minute, during which a certain number  $Z_i$  (for  $i = 1, \dots, X_n$ ) of customers arrive. These are the direct descendants. Once the  $X_n$  customers of generation  $n$  have been served, we are left with  $X_{n+1} = Z_{t_n+1} + \dots + Z_{t_n+X_n}$  new customers in line. Here,  $t_n$  is the time spent in the generations before. Hence,  $X_n$  is a branching process.

We thus see that the probability that the server can take a break is equal to the extinction probability  $\alpha$  of the process  $(X_n)$ . Indeed, if  $X_n = 0$ , then there are no more customers in line.

There are three cases to consider.

- If there exists some  $j \geq 1$  such that  $p_j = 1$ , then the extinction probability is zero.
- If  $p_1 < 1$  and  $p_0 + p_1 = 1$ , then, by the continuity of probabilities,

$$\mathbb{P}\{X_n \geq 1 \ \forall n \geq 1\} = \lim_{m \rightarrow \infty} \mathbb{P}\{X_n \geq 1 \ \forall 1 \leq n \leq m\} \leq \limsup_{m \rightarrow \infty} (1 - p_0^{X_0})^m = 0,$$

and therefore the extinction probability is 1.

- Finally, if  $p_j < 1$  for all  $j \geq 0$  and  $p_0 + p_1 < 1$ , then the assumptions  $A$  and  $B$  of the course are satisfied, and hence the extinction probability is the smallest solution of the equation  $g_Z(s) = s$  (one can verify that this holds even in the first two cases).

*Numerical application:* We seek the solutions of  $\alpha = \frac{1}{5} + \frac{1}{5}\alpha + \frac{3}{5}\alpha^2$ . We find  $\alpha = 1$  or  $\alpha = \frac{1}{3}$ , with the latter being the smallest solution. Hence, the probability that the server can take a break is  $\frac{1}{3}$ .

**Exercise 4.**

- Since  $-X_n$  is a submartingale, the a.s.-submartingale convergence theorem tells us that  $-X_n$  converges, which means that  $X_n$  converges a.s. as well.
- 2 is just the a.s.-submartingale convergence theorem. For 1, non-negative means that  $\mathbb{E}[X_n^-] = 0 < \infty$ , thus we have  $X_n$  converges a.s. by part (a). 3 is similar to 1.

**Exercise 5.**

- Let  $(S_n)_n$  be a SRW on  $\mathbb{Z}$  and  $\tau_1$  its first hitting time of 1. We saw that  $(S_{n \wedge \tau_1})_n$  is a martingale, which almost surely converges to  $S_{\tau_1} = 1$ . Latter follows from the fact that  $\tau_1$  is almost surely finite, which we have seen in class. However, for all  $n$ ,

$$0 = \mathbb{E}[S_0] = \lim_n \mathbb{E}[S_{n \wedge \tau_1}] \neq \mathbb{E}[S_{\tau_1}] = 1.$$

Therefore,  $(S_{n \wedge \tau_1})_n$  can not converge in  $L^1$ .

- Let  $(X_n)_n$  be iid standard normal random variables. Clearly,  $\sup_n \mathbb{E}[|X_n|] = \mathbb{E}[|X_n|] \leq 1$ . And by Borel-Cantelli II applied to  $(E_n(C) = \{X_n > C\})_n$ ,

$$\mathbb{P}[\limsup_n X_n \geq C] \geq \mathbb{P}[\limsup_n E_n] = 1.$$

This holds for any  $C > 0$ . Therefore, when taking  $C \uparrow \infty$  the monotonicity of the probability measure implies that  $\limsup_n X_n = \infty$  almost surely.<sup>1</sup>

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<sup>1</sup>A more general statement is the following: let  $(X_n)_n$  be iid random variables with c.d.f.  $F$ , then  $\limsup_n X_n = M$  almost surely, where  $M = \inf\{t \in \mathbb{R} : F(t) = 1\}$ .