

# Math 589 – Winter 2020 – Assignment 2

Assigned on January 31, 2020.

Due on February 9, 2020 at 2 PM (by email or in my office)

- Exercise 13.2 from the course notes.
- [How many leaves does a tree have?]** For a node  $v$  of a tree  $t$ , say that  $v$  is a *leaf* of  $t$  if  $c(v; t) = 0$ . Fix an offspring distribution  $\mu$  and write  $\alpha = \sum_{i \geq 0} i\mu(i)$ . Then let  $T$  be a  $\text{GW}_\mu$ -distributed random tree, and for  $n \geq 0$  let  $L_n = |\{v \in T_n : v \text{ is a leaf of } T\}|$ . Suppose that  $\mu(0) > 0$ .

- (a) Prove that as  $n \rightarrow \infty$ ,

$$\frac{L_n}{|T_n|} \xrightarrow{\text{a.s.}} \mu(0)\mathbf{1}_{[M \neq 0]},$$

where  $M := \limsup M_n$  is the almost sure limit of the martingale  $M_n = |T_n|/\alpha^n$ . Interpret the ratio as zero if  $|T_n| = 0$ .

**Hint.** First bound

$$\mathbf{P} \left\{ \left| \frac{L_n}{|T_n|} - \mu(0) \right| > \epsilon \mid |T_n| = k \right\},$$

then use information about the growth of  $|T_n|$  to conclude.

- (b) Prove that as  $n \rightarrow \infty$ ,

$$\frac{L_n}{\alpha^n} \xrightarrow{\text{a.s.}} M\mu(0).$$

where  $M := \limsup M_n$  is the almost sure limit of the martingale  $M_n = |T_n|/\alpha^n$ .

- [Conditioning a Galton-Watson tree to be finite.]** Let  $T$  be  $\text{GW}_\mu$ -distributed as in the previous question, and write  $q = \mathbf{P}\{|T| < \infty\}$ . Assume  $\mu$  is such that  $q > 0$ , and define a new measure  $\mu_{\text{fin}}$  on  $\mathbb{N}_{\geq 0}$  by  $\mu_{\text{fin}}(i) = \mu(i) \cdot q^{i-1}$ .
  - Prove that  $\mu_{\text{fin}}$  is a probability measure.
  - Prove that if  $\sum_{i \geq 0} i\mu(i) > 1$  then  $\sum_{i \geq 0} i\mu_{\text{fin}}(i) < 1$ .
  - Let  $T'$  be  $\text{GW}_{\mu_{\text{fin}}}$ -distributed. Prove that the conditional distribution of  $T$ , given that extinction occurs, is that of  $T'$ . In other words, prove that for any tree  $t$  and any  $n \geq 1$ ,

$$\mathbf{P}\{T'_{\leq n} = t_{\leq n}\} = \mathbf{P}\{T_{\leq n} = t_{\leq n} \mid |T| < \infty\}.$$

- [The effect of size-biased sampling.]** A size-biased sample from  $x_1, \dots, x_n$  is a random variable  $I$  with  $\mathbf{P}\{I = i\} \propto x_i$ . This question is about the distributional properties of a size-biased sample from a collection of iid random variables.

Let  $X_1, \dots, X_n$  be iid positive random variables (where  $n \geq 3$ ), and let  $S_n$  be their sum. Let  $J$  be a size-biased sample from  $X_1, \dots, X_n$ , so

$$\mathbf{P}\{J = j \mid X_1, \dots, X_n\} = X_j/S_n.$$

This determines the joint distribution of  $J$  and  $X_1, \dots, X_n$ .

Next, for  $1 \leq i \leq n-1$ , let

$$X_i^* = \begin{cases} X_i & \text{if } i < J \\ X_{i+1} & \text{if } i \geq J, \end{cases}$$

so that  $(X_1^*, \dots, X_{n-1}^*) = (X_1, \dots, X_{J-1}, X_{J+1}, \dots, X_n)$  is “ $(X_1, \dots, X_n)$  with a size-biased sample removed.” Let  $S_{n-1}^* = X_1^* + \dots + X_{n-1}^* = S_n - X_J$ .

- (a) Prove that  $X_J$  is conditionally independent of  $(X_1^*, \dots, X_{n-1}^*)$  given  $S_{n-1}^*$ , in that for any bounded Borel functions  $f : \mathbb{R}^{n-1} \rightarrow \mathbb{R}$  and  $g : \mathbb{R} \rightarrow \mathbb{R}$ ,

$$\mathbf{E}[f(X_1^*, \dots, X_{n-1}^*)g(X_J) \mid S_{n-1}^*] = \mathbf{E}[f(X_1^*, \dots, X_{n-1}^*) \mid S_{n-1}^*] \mathbf{E}[g(X_J) \mid S_{n-1}^*].$$

- (b) Prove that the conditional distribution of  $(X_1^*, \dots, X_{n-1}^*)$  given  $S_{n-1}^*$  is equivalent to the conditional distribution of  $(X_1, \dots, X_{n-1})$  given  $X_1 + \dots + X_{n-1}$ , in that for any bounded Borel function  $f : \mathbb{R}^{n-1} \rightarrow \mathbb{R}$ ,

$$\mathbf{E}[f(X_1^*, \dots, X_{n-1}^*) \mid S_{n-1}^*] \stackrel{d}{=} \mathbf{E}[f(X_1, \dots, X_{n-1}) \mid S_{n-1}].$$

(Hint: Use the definition of conditional expectation.)