

Worked with Sydney Wong and Taylor Hines

**Q 1.** Let  $Z_n$  be a branching process with offspring distribution  $X$ . Suppose that  $X$  has mean  $\mu < 1$ . Calculate the total expected number of offspring in all the generations.

*Proof.*  $\mathbb{E}[Z_n] = \mu^n$ , so

$$\mathbb{E} \left[ \sum_{n=0}^{\infty} Z_n \right] = \sum_{n=0}^{\infty} \mathbb{E}[Z_n] = \sum_{n=0}^{\infty} \mu^n = \frac{1}{1-\mu}.$$

We know that this sum converges to the given value because  $\mu < 1$ .  $\square$

**Q 2.** Let  $Z_n$  be a branching process with offspring distribution  $X \sim \text{Geom}(\frac{1}{2})$ . We showed in class that the branching process becomes extinct with probability 1. Calculate the expected number of steps to extinction.

*Proof.* Let  $T$  be the time of extinction. We know that  $\mathbb{P}(T < \infty) = 1$ , and we can express  $T$  as  $T = \sum_{n=0}^{\infty} I_n$ , where  $I_n$  indicates the event  $\{Z_n > 0\}$ . Note that  $\mathbb{E}[I_n] = \mathbb{P}(Z_n > 0) = \frac{1}{n+1}$ , since we have shown in lecture that this is true when  $X \sim \text{Geom}(\frac{1}{2})$ . Therefore,

$$\mathbb{E}[T] = \mathbb{E} \left[ \sum_{n=0}^{\infty} I_n \right] = \sum_{n=0}^{\infty} \mathbb{E}[I_n] = \sum_{n=0}^{\infty} \frac{1}{n+1} = -1 + \sum_{n=0}^{\infty} \frac{1}{n} = \infty$$

by the divergence of the harmonic series.  $\square$

**Q 3.** Let  $Z_n$  be a branching process with offspring distribution  $X$ . Suppose that  $X$  has mean  $\mu$  and variance  $\sigma^2$ . Find

$$\text{Cov}(Z_t, Z_s).$$

*Proof.* We have shown in class that

$$\text{Var}(Z_n) = \begin{cases} n\sigma^2 & \mu = 1 \\ \sigma^2 \mu^{n-1} \frac{\mu^n - 1}{\mu - 1} & \mu \neq 1 \end{cases}.$$

We may assume WLOG that  $s \leq t$ . Note that

$$\begin{aligned} \mathbb{E}[Z_t Z_s] &= \mathbb{E}[\mathbb{E}[Z_t Z_s \mid Z_s]] = \mathbb{E}[Z_s \mathbb{E}[Z_t \mid Z_s]] = \mathbb{E}[Z_s \mathbb{E}[Z_{t-s} \mid Z_0 = Z_s]] \\ &= \mathbb{E}[Z_s^2 \mu^{t-s}] = \mu^{t-s} (\text{Var}(Z_s) + \mathbb{E}[Z_s]^2) \\ &= \begin{cases} \mu^{t-s} (s\sigma^2 + \mu^{2s}) & \mu = 1 \\ \mu^{t-s} (\sigma^2 \mu^{s-1} \frac{\mu^s - 1}{\mu - 1} + \mu^{2s}) & \mu \neq 1 \end{cases} \\ &= \begin{cases} s\sigma^2 + 1 & \mu = 1 \\ \sigma^2 \mu^{t-1} \frac{\mu^s - 1}{\mu - 1} + \mu^{t+s} & \mu \neq 1. \end{cases} \end{aligned}$$

So

$$\begin{aligned} \text{Cov}(Z_s, Z_t) &= \mathbb{E}[Z_s Z_t] - \mathbb{E}[Z_s] \mathbb{E}[Z_t] \\ &= \begin{cases} s\sigma^2 + 1 - \mu^{t+s} & \mu = 1 \\ \sigma^2 \mu^{t-1} \frac{\mu^s - 1}{\mu - 1} + \mu^{t+s} - \mu^{t+s} & \mu \neq 1 \end{cases} \\ &= \begin{cases} s\sigma^2 & \mu = 1 \\ \sigma^2 \mu^{t-1} \frac{\mu^s - 1}{\mu - 1} & \mu \neq 1. \end{cases} \end{aligned}$$

Obviously, if  $t < s$  then all we need to do is to switch  $t$  and  $s$  in this expression.  $\square$

**Q 4.** Let  $Z_n$  be a branching process with offspring distribution  $X \sim \text{Geom}(p)$ . Find the extinction probability as a function of  $p$ .

*Proof.* We have shown in lecture that the extinction probability, in this case, is the smaller fixed point of the generating function  $G_X(s)$ . Thus  $s^*$  solves

$$\begin{aligned} s &= G_X(s) = \frac{1}{1 - (1-p)s} \\ \implies 0 &= (1-p)s^2 - s + p \\ &= \frac{1}{1-p}(s-1)\left(s - \frac{p}{1-p}\right). \end{aligned}$$

If  $p \leq \frac{1}{2}$ , then  $\frac{p}{1-p}$  is the smaller root, and thus the extinction probability. Otherwise, it is 1.  $\square$

**Q 5.** Students come to office hours according to a rate 5 per hour Poisson process. They stay for 10 minutes and then leave. Conditional that 8 came during the hour, what is the distribution of the number still there at the end?

*Proof.* A person is still in office hours at the end of the hour if and only if they arrived during the last 10 minutes. The arrivals in a Poisson process fall uniformly throughout the interval, so it is clear that the distribution is  $\text{Bin}(8, \frac{1}{6})$ , since this 10 minute interval is  $\frac{1}{6}$  of the entire interval. To see this proven, suppose that  $N$  is the number of arrivals in some length 1 time interval of a rate  $\lambda$  Poisson process, and that  $M$  is the number of arrivals in some length  $p$  subinterval. Then  $M$  is independent from  $N - M$ , since  $N - M$  is the number of arrivals in the complement of the length  $p$  interval, and these two intervals are disjoint. So

$$\begin{aligned} \mathbb{P}(M = k \mid N = n) &= \frac{\mathbb{P}(M = k, N - M = n - k)}{\mathbb{P}(N = n)} \\ &= \frac{\mathbb{P}(M = k)\mathbb{P}(N - M = n - k)}{\mathbb{P}(N = n)} \\ &= \frac{e^{-p\lambda} \frac{(p\lambda)^k}{k!} e^{-(1-p)\lambda} \frac{((1-p)\lambda)^{n-k}}{(n-k)!}}{e^{-\lambda} \frac{(\lambda)^n}{n!}} \\ &= \frac{n!}{k!(n-k)!} p^k (1-p)^{n-k} \\ &= \binom{n}{k} p^k (1-p)^{n-k}. \end{aligned}$$

Thus the number of arrivals in the length  $p$  subinterval, conditional on the number of arrivals  $n$  in the entire interval, is distributed  $\text{Binomial}(n, p)$ .  $\square$