

# PROBLEM I

1. Assume  $\{Y_1(n) : n \geq 0\}, \{Y_2(n) : n \geq 0\}$  are two independent migrating branch process with offspring distribution  $(p(i) : i \in \mathbb{N})$  and the migrating probability respectively are  $(\gamma_1(i) : i \in \mathbb{Z}_+), (\gamma_2(i) : i \in \mathbb{N})$ . Prove:  $\{Y_1(n) + Y_2(n) : n \geq 0\}$  is migrating branching process with offspring distribution  $p(i) : i \in \mathbb{N}$  and migrating probability  $\gamma_1 * \gamma_2$ .
2. Let  $\{Y(n) : n \in \mathbb{N}\}$  be migrating branch process with offspring distribution  $p(j) : j \in \mathbb{N}$  and the migrating distribution  $\gamma(i) : i \in \mathbb{N}$ .  $P_n^\gamma = (p_n^\gamma(i, j); i, j \in \mathbb{N})$  is the  $n$ -th transition matrix. Prove:  $\forall i, n \geq 1$

$$\sum_{j=0}^{\infty} p_n^\gamma(i, j) z^j = g_n(z)^i \prod_{k=1}^n h(g_{k-1}(z)), |z| \leq 1$$

where  $h$  is the generating function of  $(\gamma(j) : j \in \mathbb{N})$ .  $g$  is the generating function of  $(p(j) : j \in \mathbb{N})$ .

3.  $h, g$  are defined as above. Assume  $m := g'(1-) < \infty, \mu := h'(1-) < \infty$ . Prove:  $\forall i, n \geq 1$ ,

$$\mathbb{P}(Y_n | Y_0 = i) = im^n + \mu \sum_{k=1}^n m^{k-1}$$

**SOLUTION.** 1. Since  $Y_1, Y_2$  are independent Markov chain, we easily get  $\mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | \sigma(Y_1(j), Y_2(j) : 0 \leq j \leq n)) = \mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | Y_1(n), Y_2(n))$ . So to prove  $Y_1 + Y_2$  is Markov chain, we only need to prove  $\mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | Y_1(n), Y_2(n)) = \mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | Y_1(n) + Y_2(n))$ .

$$\begin{aligned} & \mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | Y_1(n) = j, Y_2(n) = k) \\ &= \sum_{x+y=i} \mathbb{P}(Y_1(n+1) = x | Y_1(n) = j) \mathbb{P}(Y_2(n+1) = y | Y_2(n) = k) \\ &= \sum_{x+y=i} p^{*j} * \gamma_1(x) p^{*k} * \gamma_2(y) \\ &= p^{*j} * \gamma_1 * p^{*k} * \gamma_2(i) \\ &= p^{*(j+k)} * \gamma_1 * \gamma_2(i) \end{aligned}$$

So  $\mathbb{P}(Y_1(n+1) + Y_2(n+1) = i | Y_1(n), Y_2(n)) = p^{*(Y_1(n) + Y_2(n))} * (\gamma_1 * \gamma_2)(i) \in \sigma(Y_1(n) + Y_2(n)) \subset \sigma(Y_1(n), Y_2(n))$ . So  $Y_1 + Y_2$  is Markov chain. More over, we have obtained  $\mathbb{P}(Y_1(n+1) + Y_2(n+1) = j | Y_1(n) + Y_2(n) = i) = p^{*i} * (\gamma_1 * \gamma_2)(j)$ . So  $\{Y_1(n) + Y_2(n) : n \geq 0\}$  is migrating branching process with offspring distribution  $p(i) : i \in \mathbb{N}$  and migrating probability  $\gamma_1 * \gamma_2$ .

2. Use MI to prove it. Write  $G_n(i, z) := \sum_{j=0}^{\infty} p_n^\gamma(i, j) z^j$ . When  $n = 0$ , we have  $p_0^\gamma(i, j) = \delta_{ij}$ , so  $G_0(i, z) = z^i = g_0(z)^i$ . When  $n = 1$ , we have  $p_1^\gamma(i, j) = p^{*i} * \gamma(j)$ . So  $G_1(i, z) = g(z)^i h(z)$ . Assume for certain  $n$  we have proved that  $G_n(i, z) = g_n(z)^i \prod_{k=1}^n h(g_{k-1}(z))$ , Consider  $n + 1$ .

Easily  $p_{n+1}^\gamma(i, j) = \sum_{k \in \mathbb{N}} p_n^\gamma(k, j) p(i, \cdot) * \gamma(k)$ . So

$$\begin{aligned}
 G_{n+1}(i, z) &= \sum_{j \in \mathbb{N}} \sum_{k \in \mathbb{N}} p_1^\gamma(i, k) p_n^\gamma(k, j) z^j \\
 &= \sum_{k \in \mathbb{N}} p_1^\gamma(i, k) G_n(k, z) \\
 &= \prod_{k=1}^n h(g_{k-1}(z)) \sum_{k \in \mathbb{N}} p_1^\gamma(i, k) g_n(z)^k \\
 &= \prod_{k=1}^n h(g_{k-1}(z)) G_1(i, g_n(z)) \\
 &= g_{n+1}(z) \prod_{k=1}^{n+1} h(g_{k-1}(z))
 \end{aligned}$$

3. Easily  $\mathbb{P}(Y_n \mid Y_0 = i) = D_z G_n(i, z) \mid_{z \rightarrow 1-}$ . Noting  $g(1) = h(1) = 1$ , easy to get that  $\mathbb{P}(Y_n \mid Y_0 = i) = im^n + \mu \sum_{k=1}^n m^{k-1}$ .

□

**PROBLEM II** Assume  $b \in (0, 1), p \in (0, 1)$ . Let  $\mu(0) = \frac{1-b-p}{1-p}, \mu(j) = bp^{j-1}, j \geq 1$ . Prove:

1.  $(\mu(j) : j \in \mathbb{N})$  is probability distribution and

$$g(z) := \sum_{j=0}^{\infty} \mu(j) z^j = \frac{1-b-p}{1-p} + \frac{bz}{1-pz}.$$

2. Let  $b = (1-p)^2$ . Then  $g'(1) = 1$  and

$$g(z) = p + \frac{(1-p)^2 z}{1-pz} = \frac{p - (2p-1)z}{1-pz}.$$

Prove:  $\forall n \geq 1$ ,

$$g_n(z) = \frac{np - ((n+1)p-1)z}{1 + (n-1)p - npz}.$$

**SOLUTION.** 1. Easily  $\sum_{j=1}^{\infty} \mu(j) = \frac{b}{1-p}$ . So  $\sum_{j=0}^{\infty} \mu(j) = 1$ . Easily  $\sum_{j=1}^{\infty} \mu(j) z^j = \frac{bz}{1-pz}$ . So  $g(z) = \mu(0) + \frac{bz}{1-pz} = \frac{1-b-p}{1-p} + \frac{bz}{1-pz}$ .

2.  $g_{n+1}(z) = g(g_n(z)) = \frac{p - (2p-1)g_n(z)}{1 - pg_n(z)}$ . So  $g_{n+1}(z) - 1 = \frac{(g_n(z)-1)(1-p)}{1 - pg_n(z)}$ . Thus, we obtain  $\frac{1}{g_{n+1}(z)-1} = \frac{1}{g_n(z)-1} - \frac{p}{1-p}$ . So  $\frac{1}{g_n(z)-1} = \frac{1}{z-1} - \frac{np}{1-p}$ , and finally we get  $g_n(z) = \frac{np - ((n+1)p-1)z}{1 + (n-1)p - npz}$ .

□

**PROBLEM III** Let  $\{X(n) : n \in \mathbb{N}\}$  be branch process with offspring distribution  $p(j) : j \in \mathbb{N}$ . And  $g$  is the generating function. Let  $m_2 := g'(1) + g''(1) < \infty$ . Let  $m = g'(1) < \infty$ .  $\forall k \geq 1$ ,  $X_n^{(k)} = k^{-1} X_n$ . Prove:  $\forall \varepsilon > 0, i, n \geq 1, \mathbb{P}(|X_n^{(k)} - im^n| \geq \varepsilon \mid X_0^{(k)} = i) \rightarrow 0, k \rightarrow \infty$ .

**SOLUTION.** In fact, we don't need  $m_2 < \infty$ . We let  $(Y(k, n) : n \in \mathbb{N}), k \in \mathbb{N}$  are independent branch process with offspring distribution  $p(j) : j \in \mathbb{N}$  and  $Y(k, 0) = i$ . Then  $\sum_{j=1}^k Y(j, n)$  is branch process with offspring distribution  $p(j) : j \in \mathbb{N}$  and initial value  $ki$ . So  $\sum_{j=1}^k Y(j, n) \stackrel{d}{=} X_n^{(k)} | X_0^{(k)} = i$ . So  $\mathbb{P}(|X_n^{(k)} - im^n| \geq \varepsilon | X_0^{(k)} = i) = \mathbb{P}(|\frac{\sum_{j=1}^k Y(j, n)}{k} - im^n| \geq \varepsilon)$ . By LLN we obtain  $\frac{1}{k} \sum_{j=1}^k Y(j, n) \xrightarrow{\text{a.s.}} im^n$ . So finally we get  $\lim_{k \rightarrow \infty} \mathbb{P}(|X_n^{(k)} - im^n| \geq \varepsilon | X_0^{(k)} = i) = \lim_{k \rightarrow \infty} \mathbb{P}(|\frac{\sum_{j=1}^k Y(j, n)}{k} - im^n| \geq \varepsilon) = 0$ .  $\square$

**PROBLEM IV** Let  $\{X(n) : n \in \mathbb{N}\}$  be branch process with offspring distribution  $p(j) : j \in \mathbb{N}$ . And  $g$  is the generating function, where  $m := g'(1) \in (1, \infty), m_2 := g'(1) + g''(1) < \infty$ . Let  $\sigma^2 := m_2 - m^2 = \mathbb{D}(X(1))$ . It is well known that  $\exists W, \lim_{n \rightarrow \infty} \frac{X_n}{m^n} = W$ . Prove:

$$\lim_{n \rightarrow \infty} \mathbb{E}_1[(m^{-n} X_n - W)^2] = 0, \mathbb{D}_1(W) = \sigma^2 m^{-1} (m - 1)^{-1}$$

**SOLUTION.** For convinence we write  $\mathbb{E}, \mathbb{D}$  instead of  $\mathbb{E}_1, \mathbb{D}_1$ . Easy to get that  $\mathbb{E}(m^{-2n} X_n^2) = \frac{\sigma^2(1-m^{-n})}{m^2-m} + 1$ . So by Fatou theorem we get that  $\mathbb{E}(W^2) \leq \lim_{n \rightarrow \infty} \mathbb{E}(m^{-2n} X_n^2) = \frac{\sigma^2}{m^2-m} + 1 < \infty$ . And by Doob Stochastic Processes p317 theorem 3.4 we get that  $\mathbb{E}(\max_{n \in \mathbb{N}} m^{-2n} X_n^2) < \infty$ . Thus,  $m^{-2n} X_n^2$  are integrable uniformly, and so do  $(m^{-n} X_n - W)^2$ . So by LCDT we can get  $\lim_{n \rightarrow \infty} \mathbb{E}((m^{-n} X_n - W)^2) = 0$ . Noting that

$$\mathbb{E}(m^{-2n} X_n^2 - W^2) = \mathbb{E}((m^{-n} X_n + W)(m^{-n} X_n - W)) \leq \sqrt{\mathbb{E}((m^{-n} X_n + W)^2) \mathbb{E}((m^{-n} X_n - W)^2)} \rightarrow 0$$

, we get  $\mathbb{E}(W^2) = \lim_{n \rightarrow \infty} \mathbb{E}(m^{-2n} X_n^2) = \frac{\sigma^2}{m^2-m} + 1$ . Also,  $\mathbb{E}(|m^{-n} X_n - W|)^2 \leq \mathbb{E}((m^{-n} X_n - W)^2)$ , so  $\mathbb{E}(W) = \lim_{n \rightarrow \infty} \mathbb{E}(m^{-n} X_n) = 1$ . So  $\mathbb{D}(W) = \mathbb{E}(W^2) - \mathbb{E}(W)^2 = \frac{\sigma^2}{m(m-1)}$ .  $\square$

**PROBLEM V** Let  $\{Y(n) : n \in \mathbb{N}\}$  be branch process with offspring distribution  $p(j) : j \in \mathbb{N}$ . And  $g$  is the generating function, where  $m := g'(1) \leq 1$ . Prove  $(p^\gamma(j) : j \in \mathbb{N})$  is the steady-state vector of transition matrix  $P_n^\gamma$ , that is  $\sum_{i=0}^\infty p^\gamma(i) p_n^\gamma(i, j) = p^\gamma(j), i \geq 0$ .

**SOLUTION.** Since  $\lim_{m \rightarrow \infty} p_m^\gamma(i, j) = p^\gamma(j)$ , and fix  $k \in \mathbb{N}$ , we have  $\sum_{j=0}^\infty p_m^\gamma(k, i) p_n^\gamma(i, j) = p_{n+m}^\gamma(k, j)$ , we only need to prove that  $\lim_{m \rightarrow \infty} \sum_{i=0}^\infty (p_m^\gamma(k, i) - p^\gamma(i)) p_n^\gamma(i, j) = 0$ . Since  $\lim_{m \rightarrow \infty} p_m^\gamma(k, i) = p^\gamma(i)$  and  $\sum_{i \in \mathbb{N}} p_m^\gamma(k, i) = 1$ , we can easily get that  $\sum_{i \in \mathbb{N}} p^\gamma(i) = 1$ . For  $\varepsilon > 0$ , we let  $N$  large enough such that  $\sum_{k=N}^\infty p^\gamma(k) < \varepsilon$ . Then we let  $M$  large enough such that  $\forall i : 0 \leq i < N, \forall m \geq$

$M, |p_m^\gamma(k, i) - p^\gamma(k)| < \frac{\varepsilon}{N}$ . Then

$$\begin{aligned}
& \left| \sum_{i=0}^{\infty} (p_m^\gamma(k, i) - p^\gamma(i)) p_n^\gamma(i, j) \right| \\
& \leq \sum_{i=0}^{\infty} |p_m^\gamma(k, i) - p^\gamma(i)| p_n^\gamma(i, j) \\
& \leq \sum_{i=0}^{N-1} |p_m^\gamma(k, i) - p^\gamma(i)| p_n^\gamma(i, j) + \sum_{i=N}^{\infty} (p_m^\gamma(k, i) + p^\gamma(i)) p_n^\gamma(i, j) \\
& \leq \sum_{i=0}^{N-1} \frac{\varepsilon}{N} + \sum_{i=N}^{\infty} p_m^\gamma(k, i) + p^\gamma(i) \\
& \leq \varepsilon + \sum_{i=N}^{\infty} p^\gamma(i) + 1 - \sum_{i=1}^{N-1} p_m^\gamma(k, i) \\
& \leq \varepsilon + \varepsilon + 1 - \sum_{i=1}^{N-1} p^\gamma(i) + \sum_{i=1}^{N-1} |p_m^\gamma(k, i) - p^\gamma(i)| \\
& \leq 4\varepsilon
\end{aligned}$$

So finally we get  $\lim_{m \rightarrow \infty} \sum_{i=0}^{\infty} (p_m^\gamma(k, i) - p^\gamma(i)) p_n^\gamma(i, j) = 0$ . Thus,  $\sum_{i=0}^{\infty} p^\gamma(i) p_n^\gamma(i, j) = p^\gamma(j), i \geq 0$ .  $\square$

**PROBLEM VI** Let  $\{Y(n) : n \in \mathbb{N}\}$  be branch process with offspring distribution  $p(j) : j \in \mathbb{N}$ . And  $g$  is the generating function, where  $m := g'(1) \leq 1$ . Discuss  $\lim_{n \rightarrow \infty} \mathbb{E}(Y_n | Y_0 = i)$ .

**SOLUTION.** Easy to get that  $\mathbb{E}(Y_n | Y_0 = i) = im^n + \mu \sum_{k=1}^n m^{k-1}$ . When  $m = 1$ , we know  $\mathbb{E}(Y_n | Y_0 = i) \rightarrow \infty$ . When  $m < 1$ , we know  $\mathbb{E}(Y_n | Y_0 = i) \rightarrow \frac{\mu}{1-m}$ .  $\square$