



# Markov limit of line of decent types in a multitype supercritical branching process

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## ABSTRACT

In a multitype ( $d$  types) supercritical positively regular Galton–Watson branching process, let  $\{X_n, X_{n-1}, \dots, X_0\}$  denote the types of a randomly chosen (i.e., uniform distribution) individual from the  $n$ th generation and this individual's  $n$  ancestors. It is shown here that this sequence converges in distribution to a Markov chain  $\{Y_0, Y_1, \dots\}$  with transition probability matrix  $(p_{ij})_{1 \leq i, j \leq d}$  and having the stationary distribution. We also consider the critical case conditioned on non-extinction.

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## 1. Introduction

Let  $\mathbf{Z}_n = (Z_{n,1}, Z_{n,2}, \dots, Z_{n,d})$  be the population vector in the  $n$ th generation,  $n = 0, 1, 2, \dots$ , where  $Z_{n,i}$  is the number of individuals of type  $i$  in the  $n$ th generation. We assume that each individual of type  $i$ ,  $i = 1, 2, \dots, d$ , lives a unit of time and, upon death, produces children of all types according to the offspring distribution  $\{p^{(i)}(\mathbf{j}) \equiv p^{(i)}(j_1, j_2, \dots, j_d)\}_{\mathbf{j} \in \mathbb{N}^d}$  and independently of other individuals, where  $p^{(i)}(j_1, j_2, \dots, j_d)$  is the probability that a type  $i$  parent produces  $j_1$  children of type 1,  $j_2$  children of type 2,  $\dots$ ,  $j_d$  children of type  $d$ .

Let  $m_{ij} = E(Z_{1,j} | \mathbf{Z}_0 = \mathbf{e}_i)$  be the expected number of type  $j$  offspring of a single type  $i$  individual in one generation for any  $i, j = 1, 2, \dots, d$ . Then,

$$\mathbf{M} \equiv \{m_{ij} : i, j = 1, 2, \dots, d\} \tag{1.1}$$

is called the offspring mean matrix.

In a discrete-time multi-type positively regular Galton–Watson branching process, by the Perron–Frobenius theorem, the matrix  $\mathbf{M}$  has a maximal eigenvalue  $\rho$  and has associated strictly positive normalized right and left eigenvectors  $\mathbf{u} = (u_1, u_2, \dots, u_d)$  and  $\mathbf{v} = (v_1, v_2, \dots, v_d)$  such that

$$\mathbf{u} \cdot \mathbf{v} = 1 \quad \text{and} \quad \mathbf{u} \cdot \mathbf{1} = 1. \tag{1.2}$$

The maximal eigenvalue  $\rho$  of the offspring mean matrix  $\mathbf{M}$  plays a crucial role. The process is called a supercritical, critical or subcritical branching process according as  $1 < \rho < \infty$ ,  $\rho = 1$  or  $\rho < 1$ , respectively (see Athreya and Ney, 2004 for the details).

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Now, we consider Galton–Watson branching process with a finite offspring mean matrix  $\mathbf{M}$  whose maximal eigenvalue  $1 < \rho < \infty$  and with no extinction. Then choose an individual at random, i.e., uniform distribution, from the  $n$ th generation and denote its type  $X_n$ . Let  $X_{n-1}$  be the type of its parent,  $X_{n-2}$  the type of its grandparent, etc., down to  $X_0$  being the type of the ancestor in the generation 0. We show that, for any integer  $k$ ,  $(X_n, X_{n-1}, \dots, X_{n-k})$  converges in distribution to  $(Y_0, Y_1, \dots, Y_k)$  where  $\{Y_n\}_{n \geq 0}$  is a Markov chain with a unique stationary distribution.

The work of Jagers and Nerman (1996) considers a similar process in a more general setting. However, their principal assumption is that the process has been evolving for an infinite amount of time and is already in a stable state. In this paper, we consider the case when the population has been evolving up to  $n$  generations and prove a limit result about the types of the ancestors of a random chosen individual as  $n \rightarrow \infty$ . Thus, the work reported here is related to but different from that in Jagers and Nerman (1996).

**2. Main results**

The first result is for the supercritical case. Without loss of generality, we assume that each individual in this supercritical process produces at least one offspring with probability 1 upon death, that is,  $P(\mathbf{Z}_1 = \mathbf{0} | \mathbf{Z}_0 = \mathbf{e}_i) = 0$  for all  $i = 1, 2, \dots, d$ . Thus, the probability of extinction is 0.

**Theorem 2.1.** *Let  $1 < \rho < \infty$ ,  $|\mathbf{Z}_0| = 1$ ,  $E(\|\mathbf{Z}_1\| \log \|\mathbf{Z}_1\| | \mathbf{Z}_0 = \mathbf{e}_i) < \infty$  for any  $i = 1, 2, \dots, d$ . Then, for any integer  $k \geq 0$ , there exists a random vector  $(Y_0, Y_1, \dots, Y_k)$  such that*

$$(X_n, X_{n-1}, \dots, X_{n-k}) \xrightarrow{d} (Y_0, Y_1, \dots, Y_k) \text{ as } n \rightarrow \infty,$$

and, for any  $i_0, i_1, \dots, i_k \in \{1, 2, \dots, d\}$ ,

$$P(Y_0 = i_0, Y_1 = i_1, \dots, Y_k = i_k) = \frac{v_{i_k} m_{i_k i_{k-1}} m_{i_{k-1} i_{k-2}} \cdots m_{i_1 i_0}}{(\mathbf{1} \cdot \mathbf{v}) \rho^k}.$$

Moreover,  $\{Y_n\}_{n \geq 0}$  is a Markov chain with the state space  $\{1, 2, \dots, d\}$ ,

(a) initial distribution  $\lambda_0 \equiv (\lambda_0(1), \lambda_0(2), \dots, \lambda_0(d))$  where

$$\lambda_0(i) = \frac{v_i}{\mathbf{1} \cdot \mathbf{v}} \text{ for any } i = 1, 2, \dots, d,$$

(b) transition probability  $\mathbf{P} \equiv (p_{ij} : i, j = 1, 2, \dots, d)$ , where

$$p_{ij} = \frac{v_j m_{ji}}{v_i \rho} \text{ for any } n = 0, 1, 2, \dots,$$

(c) and a unique stationary distribution  $\pi \equiv (\pi_1, \pi_2, \dots, \pi_d)$  where

$$\pi_i = \frac{u_i v_i}{\mathbf{u} \cdot \mathbf{v}} \text{ for any } i = 1, 2, \dots, d.$$

A similar result also holds for the critical case conditioned on non-extinction:

**Theorem 2.2.** *Let  $\rho = 1$ ,  $|\mathbf{Z}_0| = 1$  and  $E\|\mathbf{Z}_1\|^2 < \infty$ . Then, for any integer  $k \geq 0$ , there exists a random vector  $(Y_0, Y_1, \dots, Y_k)$  such that*

$$(X_n, X_{n-1}, \dots, X_{n-k}) \Big|_{|\mathbf{Z}_n| > 0} \xrightarrow{d} (Y_0, Y_1, \dots, Y_k) \text{ as } n \rightarrow \infty,$$

and, for any  $i_0, i_1, \dots, i_k \in \{1, 2, \dots, d\}$ ,

$$P(Y_0 = i_0, Y_1 = i_1, \dots, Y_k = i_k) = \frac{v_{i_k} m_{i_k i_{k-1}} m_{i_{k-1} i_{k-2}} \cdots m_{i_1 i_0}}{(\mathbf{1} \cdot \mathbf{v})}.$$

Moreover,  $\{Y_n\}_{n \geq 0}$  is a Markov chain with the state space  $\{1, 2, \dots, d\}$ ,

(a) initial distribution  $\lambda_0 \equiv (\lambda_0(1), \lambda_0(2), \dots, \lambda_0(d))$  where

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(b) transition probability  $\mathbf{P} \equiv (p_{ij} : i, j = 1, 2, \dots, d)$ , where

$$p_{ij} = \frac{v_j m_{ji}}{v_i} \text{ for any } n = 0, 1, 2, \dots,$$

(c) and a unique stationary distribution  $\pi \equiv (\pi_1, \pi_2, \dots, \pi_d)$  where

$$\pi_i = \frac{u_i v_i}{\mathbf{u} \cdot \mathbf{v}} \text{ for any } i = 1, 2, \dots, d.$$

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