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# Size-biased branching processes with overlapping generations

(based on a paper in progress together with P.Jagers)

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# 1 Size-biased Galton-Watson process

The Galton-Watson process  $\{Z(n)\}_{n \geq 0}$  is a Markov chain describing

- consecutive generation sizes of a population of (mortal) particles
- where particles reproduce independently with a common reproduction law  $p_k = P(\nu = k)$ ,  $k \geq 0$  with mean  $m = E(\nu)$ .

Size-biased GWP:  $\widehat{\text{GWP}} = \text{an immortal particle} + \text{GW}\widehat{\text{I}}$

GW $\widehat{\text{I}}$  = GWP with immigration

- an iid inflow of particles
- it is easier to study the limit behavior of GW $\widehat{\text{I}}$  than that of GWP.

Size-biased reproduction law  $\hat{p}_k = \frac{kp_k}{m}$ ,  $k \geq 1$ .

- independent offspring numbers  $\nu_1, \dots, \nu_N$ , where  $N \gg 1$
- pick a particle from  $S_N = \nu_1 + \dots + \nu_N$  particles, what is the probability that it belongs to a family of size  $k$ ?

- the number of offspring particles from families of size  $k$

$$\nu_1 1_{\{\nu_1=k\}} + \dots + \nu_N 1_{\{\nu_N=k\}} = k \cdot \text{Bin}(N, p_k)$$

- the proportion of offspring particles from families of size  $k$  is

$$\frac{k \cdot \text{Bin}(N, p_k)}{S_N} \approx \frac{kNp_k}{Nm} = \frac{kp_k}{m}$$

The immortal particle has the size-biased offspring distribution

- one of the offspring is the immortal particle itself, other offspring are mortal particles
- the joint distribution of the offspring number  $k$  and the label  $j$  assigned to the immortal daughter

$$\hat{p}_{k,j} = 1_{\{1 \leq j \leq k\}} m^{-1} p_k$$

Lyons-Pemantle-Peres [5] approach

- two measures on a tree space:  $P$  for GWP and  $\hat{P}$  for  $\widehat{\text{GWP}}$
- the Radon-Nikodym derivative  $d\hat{P}/dP = m^{-n} Z(n)$  on the sigma algebra generated by the trees stopped at height  $n$
- translate the easier results on GWI to the GWP

## 2 CMJ process in discrete time

Age-dependent reproduction law

- an individual can give birth several times during its life
- time unit is a year rather than a generation
- the joint distribution of the offspring number and the ages at birth

$$p_k(n_1, \dots, n_k), \quad k \geq 0, \quad 1 \leq n_1 \leq \dots \leq n_k < \infty$$

- the marginal offspring distribution of the offspring number

$$p_k = \sum_{1 \leq n_1 \leq \dots \leq n_k < \infty} p_k(n_1, \dots, n_k)$$

Individual life probability space  $(\Omega_0, \mathcal{A}_0, P_0)$

- an element  $\omega_0 \in \Omega_0$  is a vector

$$\omega_0 = (n_1, \dots, n_k), \quad 1 \leq n_1 \leq \dots \leq n_k < \infty, \quad k \geq 0$$

- $P_0(\omega_0) = p_k(n_1, \dots, n_k)$
- the offspring number  $\nu = \nu(\omega_0) = k$
- consecutive ages at birth  $\tau_i = \tau_i(\omega_0)$

$$\tau_i = n_i, \quad i = 1, \dots, k; \quad \tau_i = \infty, \quad i > k$$

- the joint distribution of the offspring number and the ages at birth

$$P_0(\nu = k, \tau_1 = t_1, \dots, \tau_k = t_k) = p_k(t_1, \dots, t_k)$$

$\mathbb{I} = \cup_{k=0}^{\infty} \mathbb{N}^k$  is the set of all conceivable individuals

- (0) stands for the progenitor
- (1), (2), (3), ... progenitor's daughters in the order of appearance
- (1, 1), (1, 2), ..., (2, 1), ... the granddaughters, and so on

The CMJ process is defined on the product probability space

$$(\Omega, \mathcal{A}, P) = \prod_{x \in \mathbb{I}} (\Omega_x, \mathcal{A}_x, P_0)$$

- $(\Omega_x, \mathcal{A}_x, P_0)$  is a copy of  $(\Omega_0, \mathcal{A}_0, P_0)$
- an outcome  $\omega_x \in \Omega_x$  describes a potential life  $(n_{x1}, \dots, n_{xk_x})$
- an element  $\omega = \{\omega_x\}_{x \in \mathbb{I}} \in \Omega$  contains information about all conceivable individuals, but only some of them will be recruited

### 3 The size-biased CMJ process

The size-biased life distribution found by Jagers and Nerman [4]

$$\hat{p}_{k,j}(n_1, \dots, n_k) = 1_{\{1 \leq j \leq k\}} e^{-\alpha n_j} p_k(n_1, \dots, n_k)$$

$\alpha$  is the *Malthusian parameter*:  $\sum_{n=1}^{\infty} e^{-\alpha n} m_n = 1$

- $m = \sum_{n=1}^{\infty} m_n$  where  $m_n =$  mean offspring number at age  $n$
- $\alpha = 0$  if  $m = 1$ ,  $\alpha > 0$  if  $m > 1$ , and  $\alpha < 0$  if  $m < 1$
- in the GW case  $\alpha = \ln m$ , all  $n_i = 1$ , and

$$\hat{p}_{k,j} = 1_{\{1 \leq j \leq k\}} m^{-1} p_k$$



Enhanced individual life space  $(\hat{\Omega}_0, \hat{\mathcal{A}}_0)$

- an element  $\hat{\omega}_0 \in \hat{\Omega}_0$  is a vector

$$\hat{\omega}_0 = (j, n_1, \dots, n_k), \quad 0 \leq j \leq k, \quad 1 \leq n_1 \leq \dots \leq n_k < \infty, \quad k \geq 0$$

- if  $j = 0$ , then the individual is mortal, otherwise the individual is immortal and  $j$  specifies its immortal daughter
- put  $\gamma = \gamma(\hat{\omega}_0) = j$

Two probability measures  $P_0$  and  $\hat{P}_0$  on  $(\hat{\Omega}_0, \hat{\mathcal{A}}_0)$

$$P_0(\gamma = j, \nu = k, \tau_1 = n_1, \dots, \tau_k = n_k) = p_k(n_1, \dots, n_k) 1_{\{j=0\}}$$

$$\hat{P}_0(\gamma = j, \nu = k, \tau_1 = n_1, \dots, \tau_k = n_k) = \hat{p}_{k,j}(n_1, \dots, n_k)$$

The regeneration age  $\tau_\gamma$  of the immortal individual

- $\hat{P}_0(\tau_\gamma = n) = e^{-\alpha n} m_n$
- pick an individual in an old supercritical population and follow backwards in time its lineage, then  $\tau_\gamma \approx$  the age at childbearing
- average age at birth for the mortal individuals

$$\beta = \hat{E}_0(\tau_\gamma) = E_0(\tau_1 e^{-\alpha\tau_1} + \dots + \tau_\nu e^{-\alpha\tau_\nu})$$

The size-biased distribution of the offspring number

$$\hat{P}_0(\nu = k) = E_0(\xi \cdot 1_{\{\nu=k\}})$$

- where  $\xi = e^{-\alpha\tau_1} + \dots + e^{-\alpha\tau_\nu}$  [=  $\frac{\nu}{m}$  in the GW case]
- $E_0(\xi) = \sum_{n=1}^{\infty} e^{-\alpha n} m_n = 1$

Jagers [3] approach for constructing multitype branching processes based on Tulcea's theorem.

A measure  $\hat{P}$  for the size-biased CMJ process is defined on

$$(\hat{\Omega}, \hat{\mathcal{A}}) = \prod_{x \in \mathbb{I}} (\hat{\Omega}_x, \hat{\mathcal{A}}_x)$$

- individual life spaces  $(\hat{\Omega}_x, \hat{\mathcal{A}}_x)$  are copies of  $(\hat{\Omega}_0, \hat{\mathcal{A}}_0)$
- assign the immortal measure  $\hat{P}_0$  to the progenitor's life  $(\hat{\Omega}_0, \hat{\mathcal{A}}_0)$
- given  $\hat{\omega}_0 = (j, n_1, \dots, n_k)$  assign  $\hat{P}_0$  to  $(\hat{\Omega}_j, \hat{\mathcal{A}}_j)$
- given  $\hat{\omega}_j = (j', n'_1, \dots, n'_{k'})$  assign  $\hat{P}_0$  to  $(\hat{\Omega}_{jj'}, \hat{\mathcal{A}}_{jj'})$  and so on
- assign the mortal measure  $P_0$  to  $(\hat{\Omega}_x, \hat{\mathcal{A}}_x)$  off the immortal lineage

## 4 CMJ trees

A CMJ tree defined on  $(\Omega, \mathcal{A}, P) = \prod_{x \in \mathbb{I}} (\Omega_x, \mathcal{A}_x, P_0)$

- $\omega_x \in \Omega_x$  describes a potential branch at vertex  $x$
- let  $\sigma_x = \sigma_x(\omega)$  stand for the height of the vertex  $x$ :  $\sigma_0 = 0$  and

$$\sigma_{xi} = \sigma_x + n_{xi}, \text{ given } \omega_x = (n_{x1}, \dots, n_{xk_x})$$

- the set of branches  $T(\omega) = \{\omega_x\}_{x:\sigma_x < \infty}$  builds a CMJ tree
- the tree  $T(\omega)$  is independent of  $\{\omega_x\}_{x:\sigma_x = \infty}$
- the CMJ tree  $[T]_n = \{\omega_x\}_{x:\sigma_x \leq n}$  stopped at level  $n$

$$P([T]_n = t) = p_k(n_1, \dots, n_k) \prod_{i: n_i \leq n} P([T]_{n-n_i} = t^{(i)})$$

A measure on the tree space induced by  $(\hat{\Omega}, \hat{\mathcal{A}}, \hat{P})$

- $\hat{\omega}_x = (j, \omega_x)$  describes a potential branch of a tree with an immortal lineage
- $\hat{T}(\hat{\omega}) = \{\hat{\omega}_x\}_{x:\sigma_x < \infty}$  defines a CMJ tree  $T(\hat{\omega}) = \{\omega_x\}_{x:\sigma_x < \infty}$  with an immortal lineage  $\{j, j', \dots\}$
- the stopped tree  $[\hat{T}]_n = \{\hat{\omega}_x\}_{x:\sigma_x \leq n}$  has a discrete distribution satisfying the recursion

$$\hat{P}([\hat{T}]_n = \hat{t}) = \hat{p}_{k,j}(n_1, \dots, n_k) \hat{P}([\hat{T}]_{n-n_j} = \hat{t}^{(j)}) \prod_{i \neq j: n_i \leq n} P([T]_{n-n_i} = t^{(i)})$$

## Radon-Nikodym derivative

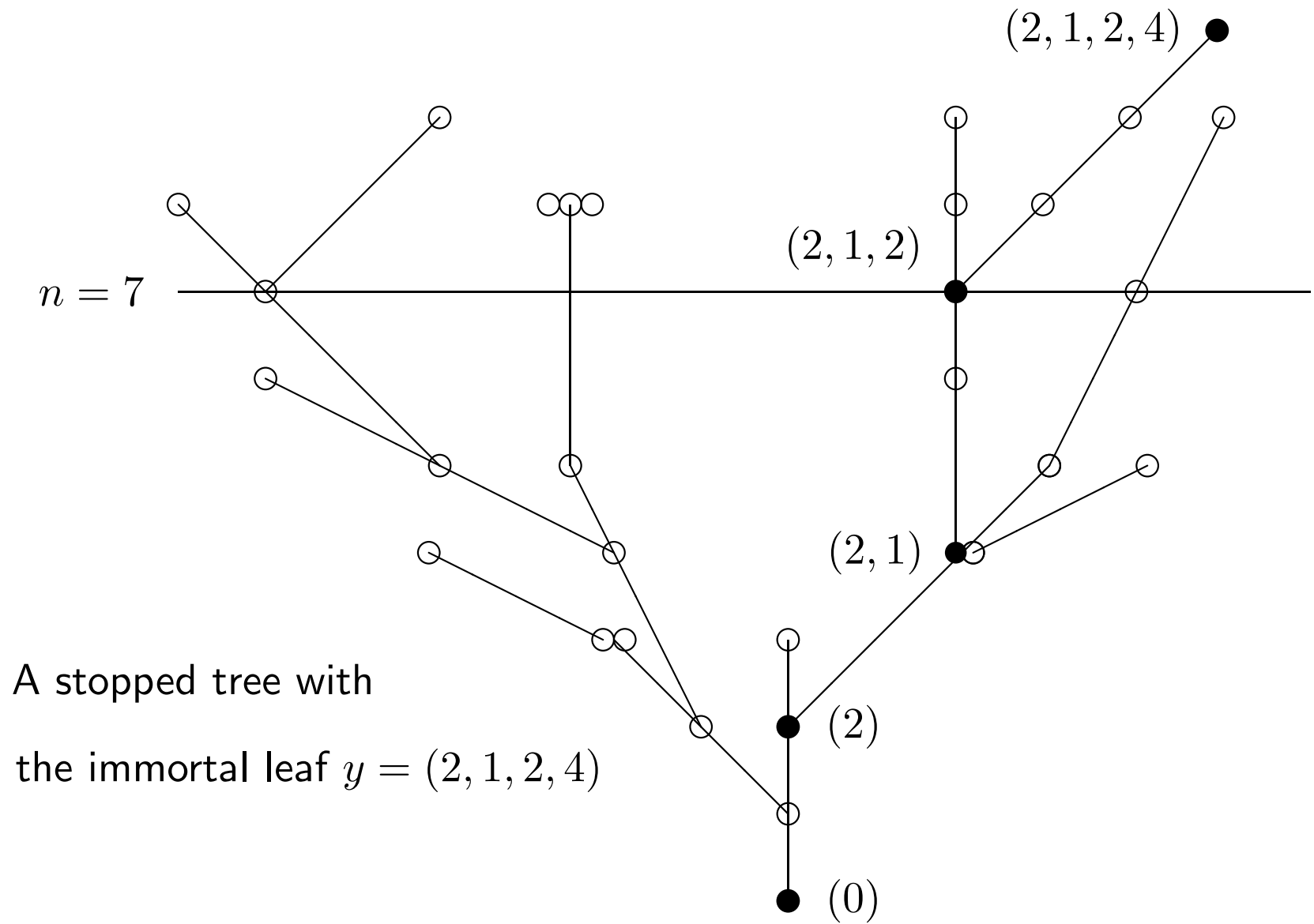
- Nerman's martingale  $Y_n = \sum_{x \in I_n} e^{-\alpha \sigma_x}$   
the set of leaves  $I_n \subset \mathbb{I}$  of the stopped tree  $[T]_n$
- the immortal leaf  $y \in I_n$  defines the stopped immortal lineage
- the two recursions for the stopped trees imply

$$\hat{P}([\hat{T}]_n = (t, y)) = e^{-\alpha \sigma_y} \cdot P([T]_n = t)$$

- summing over possible immortal lineages

$$\hat{P}([T]_n = t) = Y_n \cdot P([T]_n = t)$$

- $\left. \frac{d\hat{P}}{dP} \right|_{\mathcal{B}_n} = Y_n$  on the  $\sigma$ -algebra  $\mathcal{B}_n \subset \mathcal{A}$  generated by the complete lives of all individuals born up to time  $n$ : even if  $x$  is born at  $n$  we still include her future in  $\mathcal{B}_n$



## 5 $x \log x$ condition

Individual contributions  $\chi_x(n) = \chi(n, w_x)$  by  $x \in \mathbb{I}$ .

Jagers' population sum

$$X_n = \sum_{x: \sigma_x \leq n} \chi_x(n - \sigma_x)$$

has different distributions under the probability measures  $P$  and  $\hat{P}$

- $X_n$  is adapted to the filtration  $\mathcal{B}_n$  and  $\left. \frac{d\hat{P}}{dP} \right|_{\mathcal{B}_n} = Y_n$
- thus for any positive Borel function  $g$

$$\hat{E}[g(X_n)] = E[Y_n g(X_n)]$$



In the supercritical case

$$(Y_n, e^{-\alpha n} X_n) \rightarrow (W, cW), \quad n \rightarrow \infty$$

and therefore

$$\hat{E}(e^{-\lambda e^{-\alpha n} X_n}) \rightarrow E(W e^{-\lambda cW})$$

The limit  $W$  is not degenerate provided the  $\xi \log \xi$  condition:

$$E(\xi \log \xi) < \infty,$$

where  $\xi = e^{-\alpha\tau_1} + \dots + e^{-\alpha\tau_\nu}$  and  $E(\xi \log \xi) = \hat{E}(\log \xi)$

Under  $\hat{P}$  the weak convergence  $e^{-\alpha n} X_n \rightarrow c\hat{W}$  holds with  $\hat{E}(e^{-\lambda \hat{W}}) = E(W e^{-\lambda W})$

Since  $\nu \geq \xi \geq \nu e^{-\alpha\tau_\nu}$ , condition  $(\xi \log \xi)$  follows from

$$E(\xi \log \nu) < \infty$$

(a condition arising in the subcritical case) which in turn is weaker than

$$E(\nu \log \nu) < \infty$$

which is equivalent to  $(\xi \log \xi)$  given  $\tau_\nu \leq \text{const.}$

In general  $(\xi \log \xi)$  is not equivalent to  $(\nu \log \nu)$

Counterexample: the Sevastyanov model with  $\tau_1 = \dots = \tau_\nu = \tau$

- $\xi = e^{-\alpha\tau} \nu$ ,  $\beta = E(\tau\xi)$ , and  $E(\xi \log \xi) = E(\xi \log \nu) - \alpha\beta$
- if  $\beta < \infty$  condition  $(\xi \log \xi)$  becomes equivalent to  $(\xi \log \nu)$
- $\tau = \nu$ ,  $E(\nu \log \nu) = \infty$ ,  $E(e^{-\alpha\nu} \nu^2) < \infty \Rightarrow \beta < \infty$  and  $(\xi \log \nu)$

In general if  $\alpha > 0$  and  $\beta < \infty$  (supercritical Malthusian case), then  $(\xi \log \xi)$  is equivalent to  $(\xi \log \nu)$

A sketch of the proof

- we verify that if  $\beta < \infty$  and  $\hat{E}(\log \nu) = \infty$ , then  $E(\xi \log \xi) = \infty$
- turn to the total number of individuals  $X_n$  born up to time  $n$
- since  $\beta < \infty$ , the size-biased process is bounded from below by a GWI process with iid immigration distributed as  $(\nu - 1)$  under  $\hat{P}$
- given  $\hat{E}(\log \nu) = \infty$  we modify the argument in Section 3 of [5] to show  $e^{-cn} X_n \rightarrow 0$  for the ordinary CMJ process whatever is  $c > 0$
- according to [4] it follows that condition  $(\xi \log \xi)$  does not hold

## 6 Further questions

1. In the supercritical case: if  $\beta < \infty$  is necessary for the equivalence  $(\xi \log \xi) \Leftrightarrow (\xi \log \nu)$ ?
2. In the critical case obtain alternative proofs for the asymptotic results: classical and Vatutin's discrete theorems
3. In the subcritical case obtain an alternative proof for the limit theorem under condition  $(\xi \log \nu)$
4. Alternative proofs for the reduced CMJ processes

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