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CRITICAL CONTROLLED BRANCHING PROCESSES AND THEIR RELATIVES*

George P. Yanev

This survey aims at collecting and presenting results for one-type, discrete time branching processes with random control functions. In particular, the subclass of critical migration processes with different regimes of immigration and emigration is reviewed in detail. Critical controlled branching processes with continuous state space are also discussed.

1. Introduction

The independence of individuals' reproduction is a fundamental assumption in branching processes. Since the 1960s, a number of authors have been studying models allowing different forms of population size dependence. Sevastyanov and Zubkov (1974) proposed a class of branching processes in which the number of reproductive individuals in one generation decreases or increases depending on the size of the previous generation through a set of control functions. The individual reproduction law (offspring distribution) is not affected by the control and remains independent of the population size. These processes are known as controlled or φ -branching processes (CBP). N. Yanev (1975) (no relation to the author) extended the class of CBP by introducing random control functions. The so called φ -processes with random φ can be defined as follows.

Key words: critical branching processes, immigration, emigration, control, extinction

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Definition. The process $\{Z_n, n = 0, 1, ...\}$ is called controlled branching process (CBP) if

(1)
$$Z_{n+1} = \sum_{i \in I} \sum_{j=1}^{\varphi_{i,n}(Z_n)} \xi_{j,n}(i), \quad n \ge 0; \quad Z_0 = z_0 > 0,$$

where I is an (finite or infinite) index set and for $i \in I$

- (i) $\xi_i = \{\xi_{j,n}(i), j = 1, 2, ...; n = 0, 1, ...\}$ are i.i.d., non-negative, integer-valued r.v.'s, (independent for different i's). Denote $\xi := \xi_1$.
- (ii) $\varphi_i = \{\varphi_{i,n}(k), k = 0, 1, ...; n = 0, 1, ...\}$ are non-negative, integer-valued r.v.'s, independent from ξ_i , (independent for different i's), and such that $P(\varphi_{i,n}(k) = j) = p_k(j)$, for j = 0, 1, ... Denote $\varphi(k) := \varphi_1(k)$.

The recurrence (1) describes a very large class of stochastic processes including, for instance, all Markov chains with discrete time. Among the particular cases of CBPs is the classical Galton-Watson process (GWP) as well as popular discrete time branching processes such as: (i) processes with immigration: $I = \{1, 2\}, \ \varphi_{1,n}(k) = k, \ \text{and} \ \varphi_{2,n}(k) \equiv 1; \ \text{(iii)} \ \text{processes} \ \text{with state-dependent immigration:} \ I = \{1, 2\}, \ \varphi_{1,n}(k) = k, \ \text{and} \ \varphi_{2,n}(k) = \max\{1 - k, 0\}; \ \text{and} \ \text{(iii)} \ \text{processes} \ \text{with random migration} \ \text{(to be discussed in Sections 3–5):} \ I = \{1, 2\}, \ \varphi_{1,n}(k) = \max\{\min\{k, k + \beta_n\}, 0\}, \ \text{and} \ \varphi_{2,n}(k) = \max\{\beta_n, 0\}, \ \text{where for} \ p + q + r = 1 \ \text{we have} \ P(\beta_n = -1) = p, \ P(\beta_n = 0) = q, \ \text{and} \ P(\beta_n = 1) = r. \ \text{In all these subclasses of CBPs, the controlled functions satisfy the condition}$

$$\lim_{n \to \infty} \sum_{i \in I} \varphi_{i,n}(k) = \infty \quad \text{a.s.}, \qquad k \ge 1,$$

which can be identified as a general property of CBPs.

In Section 2 we present a classification of CBPs into subcritical, critical, and supercritical based on their mean growth rate. Two sets of conditions for extinction and non-extinction are presented in relation to this classification. Finally, limit theorems for the critical CBPs are given. The next three sections, are devoted to critical processes with different regimes of migration. In Section 3, processes with migration, stopped and non-stopped at zero, are defined and limit theorems in the critical case are discussed. Processes with time non-homogeneous migration are treated in Section 4. In Section 5, a more general type of migration is considered, utilizing a regenerative construction. CBPs with continuous state

space are discussed in Section 6. Finally, the paper ends with some concluding remarks and a list of references.

2. General Class of Controlled Branching Processes

In this section we shall discuss a classification of CBPs, which is similar to that of the classical Galton-Warson processes. Then we will focus our attention on the critical case.

As it will become clear below, the asymptotic behavior of the CBPs depends crucially on the so-called mean growth rate. Following Bruss (1984), we define the mean growth rate per individual in a population with k mothers by

$$\tau_k := k^{-1} E[Z_{n+1} \mid Z_n = k] = k^{-1} E[\varphi(k)] E[\xi].$$

In the particular case of GWP, we have $\tau_k = E[\xi]$, i.e., the mean growth rate equals the offspring mean and remains constant for any k.

2.1. Extinction and Classification of CBPs

Extinction, along with growth and composition of the population, is a principal subject of interest in the theory of branching processes. If the control functions satisfy $\varphi_n(0) \equiv 0$ a.s., then $\{Z_n\}$ is a Markov chain with absorption state 0. Furthermore, it can be proven (see [33]) that if $P(\xi = 0) > 0$ or $P(\varphi(k) = 0) > 0$ for k = 1, 2, ..., then the classical extinction-explosion duality

$$P(Z_n \to 0) + P(Z_n \to \infty) = 1$$

holds. The following key theorem for the extinction probability of CBPs is proven in Gonzales et al. (2002).

Theorem 2.1 ([9]).

- (i) If $\limsup_{k\to\infty} \tau_k < 1$, then $P(Z_n \to 0 \mid Z_0 = N) = 1$ for $N \ge 1$.
- (ii) If $\liminf_{k\to\infty} \tau_k > 1$, then there exists N_0 such that for $N \geq N_0$ we have $P(Z_n \to 0 \mid Z_0 = N) < 1$.

Referring to Theorem 2.1, Gonzalez et al. (2005) classify the CBPs as follows.

Definition. The class of CBPs can be partitioned into three subclasses:

- (i) subcritical if $\limsup_{k\to\infty} \tau_k < 1$;
- (ii) critical if $\liminf_{k \to \infty} \tau_k \le 1 \le \limsup_{k \to \infty} \tau_k$;
- (iii) supercritical if $\liminf_{k\to\infty} \tau_k > 1$.

Unlike the supercritical GWP, if the number of ancestors in the supercritical CBP is not sufficiently large, then the extinction probability might be one. This resembles the situation with the two–sex branching processes (e.g., [11]). On the other hand, the critical CBP does not always have extinction probability one, as it is seen in (4) below.

Gonzalez at al. (2005) study in detail the extinction probability of the critical CBP considering different rates of convergence of τ_k to one. Their findings rely on analysis of the stochastic difference equation

(2)
$$Z_{n+1} = Z_n + h(Z_n) + \delta_{n+1}$$
 a.s., $n = 0, 1, ...,$

where $h(k) = E[\varphi(k)]E[\xi] - k$ and $\delta_{n+1} = Z_{n+1} - E[Z_{n+1}|Z_n]$ (martingale difference). Denote

$$2\sigma_s(k) := E[|\delta_{n+1}|^s | Z_n = k]$$
 $s > 0.$

It is proven in [10] that if

(3)
$$\lim_{k \to \infty} \tau_k = 1 \quad \text{and} \quad \tau(k) \ge 1,$$

then for $N \geq 1$

(4)
$$P(Z_n \to 0 \mid Z_0 = N) \begin{cases} = 1 & \text{if } \lim \sup_{k \to \infty} (\tau_k - 1) k^2 (\sigma_2(k))^{-1} < 1; \\ < 1 & \text{if } \lim \inf_{k \to \infty} (\tau_k - 1) k^2 (\sigma_2(k))^{-1} > 1. \end{cases}$$

A different set of conditions for extinction and non-extinction of $\{Z_n\}$ is obtained by N. Yanev (1975) using a random walk construction. It is proven in [33] that if the control functions have a linear growth a.s., that is

$$\varphi_n(k) = \alpha_n k \ (1 + o(1))$$
 a.s. $k \to \infty$,

where $\{\alpha_n\}$ are i.i.d. and independent of the reproduction, then for $N \geq 1$

$$P(Z_n \to 0 \mid Z_0 = N) \begin{cases} = 1 & \text{if } E[\log(\alpha_1 E \xi)] < 0; \\ < 1 & \text{if } E[\log(\alpha_1 E \xi)] > 0. \end{cases}$$

Bruss (1980) shows that the independence of reproduction assumption for $\{\alpha_n\}$ can be removed.

2.2. Limit Theorems for Critical CBPs

Assuming (3), let us turn to the critical CBPs. It follows from (4) that, depending on the rate of convergence of τ_k to one, the extinction probability is either one or less than one. We will consider these two cases separately. Utilizing a gammalimit theorem (see [14]) for the stochastic difference equation (2), Gonzalez et al. (2005) prove the following two theorems.

Case A. The extinction is almost sure, i.e., $P(Z_n \to 0 \mid Z_0 = N) = 1$.

Theorem 2.2 ([10]). Assume

(i)
$$\tau_k = 1 + ck^{-1}$$
, $c > 0$, $k = 1, 2, ...$;

(ii)
$$\sigma_2(k) = 2ak + O(1)$$
, $a > 0$, as $k \to \infty$;

(iii)
$$\sup_{k\geq 1} \left(g_k^{1/k}\right)^{\prime\prime\prime}(1) < \infty$$
, where $g_k(s) := E\left[s^{\varphi(k)}\right]$, $0 \leq s \leq 1$.

If c < a, then

(5)
$$\lim_{n \to \infty} P\left(\frac{Z_n}{an} \le x | Z_n > 0\right) = 1 - e^{-x}.$$

Note that the limiting distribution in (5) is exponential as in the critical GWP. However, one difference is that $P(Z_n > 0) \sim c_1 n^{-(1-c/a)}$, $c_1 > 0$ whereas the survival probability in the GWP has a decay rate $(bn)^{-1}$.

Case B. Positive non-extinction probability, i.e., $P(Z_n \to 0 \mid Z_0 = N) < 1$.

Theorem 2.3 ([10]). Assume as $k \to \infty$

(i)
$$\tau_k = 1 + ck^{-(1-\alpha)} + o\left(k^{-(1-\alpha)}\right), \quad c > 0, \ 0 < \alpha < 1;$$

(ii)
$$\sigma_2(k) = 2ak^{1+\alpha} + o(k^{1+\alpha}), \quad a > 0;$$

(iii)
$$\sigma_{2+s}(k) = O\left((\sigma_2(k))^{1+s/2}\right)$$
 for some $s > 0$.

If c > a, then

(6)
$$\lim_{n \to \infty} P\left(\frac{Z_n^{1-\alpha}}{(1-\alpha)^2 a n} \le x \mid Z_n > 0\right) = \frac{1}{\Gamma(\gamma)} \int_0^x t^{\gamma - 1} e^{-t} dt,$$

where $\gamma = (c - a\alpha)/(a(1 - \alpha))$ and $\Gamma(x)$ is the Gamma function.

Obviously, if $\alpha = 0$ and c = a, then (6) coincides with (5).

3. Branching Processes with Migration

In the context of queueing theory, stochastic models with migration were discussed in [12]. A model of branching process with emigration-immigration (migration) was introduced in Nagaev and Han (1980). The readers are referred to the survey [23] and the monograph [20] for a detailed account of results for processes with a variety of immigration and emigration regimes.

In Sections 3–5 we shall review results for a class of CBPs, called branching processes with migration, introduced by N. Yanev and Mitov in 1980 when a detailed study of these branching models began. These processes were already mentioned in Section 1 as a special class of CBPs. The particular choice of control functions $\varphi(k)$ allows for a detailed analysis which in turn leads to interesting new findings. On the other hand, branching processes with migration are sufficiently general to include as subclasses previously studied models with different regimes of immigration and emigration.

Definition. The process $\{Y_n, n = 0, 1, ...\}$ is called a branching process with migration if $Y_0 > 0$ and for n = 0, 1, ...

(7)
$$Y_{n+1} = \begin{cases} \sum_{k=1}^{Y_n} \xi_{k,n} + M_n^+ & \text{if } Y_n > 0; \\ M_n^0 & \text{if } Y_n = 0, \end{cases}$$

where for p + q + r = 1

(8)
$$M_n^+ = \begin{cases} -\xi_{1,n} & probab. & p, & (emigration) \\ 0 & probab. & q, & (no migration) \\ \eta_n & probab. & r, & (immigration) \end{cases}$$

is the migration outside zero and

$$M_n^0 = \begin{cases} 0 & probab. & 1-r, & (no migration) \\ \eta_n & probab. & r, & (immigration at 0), \end{cases}$$

is the migration at zero. The number of immigrants $\{\eta_n, n = 1, 2, ...\}$ are i.i.d. non-negative, integer-valued, and independent from the offspring variables.

The process $\{Y_n\}$ can be interpreted as follows. Three scenarios are possible: (i) the offspring of one individual is removed (emigration) with probability p; (ii) there is no migration with probability q; or (iii) η_n individuals join the population (immigration) with probability r. The state zero is a reflecting barrier for $\{Y_n\}$. The emigration here can be regarded as "reversed" (negative) immigration since the branching process is modified to allow both positive and negative increments. See [19] and the references within for other approaches to emigration.

3.1. Branching Migration Processes with Reflection Barrier at Zero In Sections 3–5 we assume (unless stated otherwise) that $\{Y_n\}$ is critical with finite offspring variance and finite immigration mean, i.e.,

(9)
$$E[\xi] = 1$$
, $2b := Var[\xi] < \infty$, and $d := E[\eta_n] < \infty$.

The long-term behavior of the critical $\{Y_n\}$ depends crucially on the parameter

(10)
$$\theta := \frac{EM_n^+}{(Var[\xi])/2} = \frac{rE[\eta_n] - pE[\xi]}{(Var[\xi])/2} = \frac{rd - p}{b},$$

i.e., the ratio of the mean migration outside zero over half of the offspring variance. Depending on the values of θ , the aperiodic and irreducible Markov chain $\{Y_n\}$ can be classified as

$$\{Y_n\} = \begin{cases} \text{non-recurrent} & \theta > 1\\ \text{null-recurrent} & 0 \le \theta \le 1\\ \text{positive-recurrent} & \theta < 0. \end{cases}$$

The following limiting results are obtained in [30].

Theorem 3.1 ([30]). Assume (9).

(A) If $\theta > 0$ (dominating immigration) and $Var[\eta_n] < \infty$, then

$$\lim_{n \to \infty} P\left(\frac{Y_n}{bn} \le x\right) = \frac{1}{\Gamma(\theta)} \int_0^x t^{\theta - 1} e^{-t} dt.$$

(B) If $\theta = 0$ (zero average migration), then

$$\lim_{n \to \infty} P\left(\frac{\log Y_n}{\log n} \le x\right) = x, \qquad 0 < x < 1.$$

(C) If $\theta < 0$ (dominating emigration), then there is a limiting-stationary distribution, i.e.,

$$\lim_{n \to \infty} P(Y_n = k) = v_k, \qquad \sum_{k=0}^{\infty} v_k = 1$$

and $V(s) = \sum_{k=0}^{\infty} v_k s^k$ is the unique p.g.f. solution of a functional equation.

Remark. It is worth pointing out here a limit theorem due to Dyakonova (1997) for the "close to critical" process $\{Y_n\}$, i.e., assuming that the offspring mean $m := E[\xi] \uparrow 1$. It is known that if m < 1, then $\{Y_n\}$ has a limiting-stationary distribution. Let V be the limiting random variable with this distribution. Then it is proven in [8] that

$$\lim_{m \uparrow 1} P\left(\frac{\log V}{\log \frac{1}{1-m}} \le x\right) = x, \qquad x \in (0,1).$$

3.2. Branching Migration Processes with Absorbing State Zero

In this subsection we shall consider the branching process with no migration when it hits zero, i.e., $M_n^0 = 0$ a.s. in (7) and hence zero is an absorbing state.

Definition. Let $Y_0^0 > 0$ and for n = 1, 2, ...

$$Y_n^0 = Y_n I_{\{Y_n > 0\}}$$
 a.s.,

where I_A denotes the indicator of the event A. Then $\{Y_n^0\}$ is called a migration process with absorption at zero.

It is shown in [29] and [32], under some additional finite moment conditions, that the probability of the process surviving to time n satisfies as $n \to \infty$

$$P(Y_n^0 > 0) \sim \begin{cases} c_{\theta} > 0 & \theta > 1 \\ c_{\theta} (\log n)^{-1} & \theta = 1 \\ c_{\theta} n^{-(1-|\theta|)} & 0 \le \theta < 1, \\ c_{\theta} n^{-(1+|\theta|)} & \theta < 0. \end{cases}$$

Referring to these results, one can adopt the following classification for the critical process $\{Y_n^0\}$: (i) critical-supercritical for $\theta > 1$; critical-critical for $0 \le \theta \le 1$; and (iii) critical-subcritical for $\theta < 0$.

The next two limit results were proven in [38] (for $\theta > 0$) and [29] (for $\theta < 0$).

Theorem 3.2 ([38], [29]). Assume (9).

(A) If $\theta > 1$ (strongly dominating immigration), then

$$\lim_{n\to\infty}P\left(\frac{Y_n^0}{bn}\leq x|Y_n^0>0\right)=\frac{1}{\Gamma(\theta)}\int_0^xt^{\theta-1}e^{-t}\,dt.$$

(B) Assume $\theta \leq 1$ and some additional moment conditions when $\theta < 0$ (see [29]). Then

$$\lim_{n \to \infty} P\left(\frac{Y_n^0}{bn} \le x | Y_n^0 > 0\right) = 1 - e^{-x}.$$

Remarks. (i) If the rate of migration is not too high, i.e., $\theta \leq 1$, then the long-term behavior of $\{Y_n^0\}$ over the non-extinction trajectories is the same as in the critical GWP. The observation made after (5) applies here too.

(ii) One extension of Theorem 3.2(A), when the distribution of the initial number of ancestors Y_0^0 belongs to the domain of attraction of a stable law with parameter in (0,1], is given in [31].

4. Time Non-Homogeneous Migration

N. Yanev and Mitov (1985) study branching processes with time non-homogeneous migration defined as follows.

Definition. The process $\{\tilde{Y}_n : n = 0, 1, ...\}$ is called a branching process with non-homogeneous migration if $\tilde{Y}_0 > 0$ and for n = 1, 2, ...

(11)
$$\tilde{Y}_{n+1} = \begin{cases} \sum_{k=1}^{\tilde{Y}_n} \xi_{k,n} + \tilde{M}_n^+ & if \quad \tilde{Y}_n > 0; \\ \tilde{M}_n^0 & if \quad \tilde{Y}_n = 0, \end{cases}$$

where the migration is given for $p_n + q_n + r_n = 1$ by

(12)
$$d\tilde{M}_{n}^{+} = \begin{cases} -\xi_{1,n} & probab. & p_{n}, & (emigration) \\ 0 & probab. & q_{n}, & (no \ migration) \\ \eta_{n} & probab. & r_{n}, & (immigration) \end{cases}$$

and

$$\tilde{M}_{n}^{0} = \begin{cases} 0 & probab. & 1 - r_{n}, & (no \ migration) \\ \eta_{n} & probab. & r_{n}, & (immigration \ at \ \theta). \end{cases}$$

Unlike (8), here the probabilities p_n, q_n and r_n controlling the migration are time-dependent. Thus, $\{\tilde{Y}_n\}$ is a non-homogeneous Markov chain. In addition to (9), suppose that the immigration variance is finite, i.e.,

$$(13) Var[\eta_n] < \infty.$$

In the rest of this section we also assume that the migration decreases to 0, i.e., $\lim_{n\to\infty}q_n=1$.

Case A. Decreasing to Zero Migration and $p_n = o(r_n)$.

Theorem 4.1 ([35], see also [36]). Suppose (9) and (13). If as $n \to \infty$

$$r_n \sim \frac{r}{\log n}, \qquad p_n = o(r_n),$$

then

$$\lim_{n \to \infty} P\left(\frac{\log Z_n}{\log n} \le x\right) = e^{-rd(1-x)/b} \quad 0 \le x \le 1.$$

Theorem 4.2 ([37], see also [36]). Suppose (9) and (13). If as $n \to \infty$

$$r_n \sim \frac{l_n}{\log n}$$
 and $p_n = o(r_n),$

where $l_n \sim o(\log n) \to \infty$, then for $x \ge 0$

$$\lim_{n \to \infty} P\left(l_n \left(1 - \frac{\log Z_n}{\log n}\right) \le x\right) = 1 - e^{-dx/b}.$$

Case B. Decreasing to Zero Migration and $p_n = dr_n$.

If both immigration and emigration decrease to zero at the same rate, then a key role for the limiting behavior of the process is played by the series $\sum_{n=0}^{\infty} p_n$

and $\sum_{n=0}^{\infty} r_n$. This observation is made precise in the next three theorems.

Theorem 4.3 ([36], [7]). Suppose (9), (13), and $p_n = dr_n$. If one of the following two conditions holds as $n \to \infty$

(i) $p_n \sim l_n n^{-v}$ for 0 < v < 1 where l_n is a s.v.f. at ∞ .

(ii) $p_n = O\left((\log n)^{-1}\right)$, then

$$\lim_{n \to \infty} P\left(\frac{\log \tilde{Y}_n}{\log n} \le x | \tilde{Y}_n > 0\right) = x, \quad x \in (0, 1).$$

Theorem 4.3 is an analog of Foster's result for processes with immigration at zero only. Unlike Foster's model, $\{\tilde{Y}_n\}$ is a non-homogeneous Markov chain.

Theorem 4.4 ([36]). Suppose (9), (13), and $p_n = dr_n$. Assume as $n \to \infty$

(i) $p_n \sim l_n n^{-1}$, where l_n is a s.v.f. at ∞ ;

(ii)
$$\lim_{n \to \infty} \frac{p_n n \log n}{\sum_{k=1}^n p_k} = C \text{ for } 0 \le C \le \infty.$$

Then,

(14)
$$\lim_{n \to \infty} P\left(\frac{\log \tilde{Y}_n}{\log n} \le x | \tilde{Y}_n > 0\right) = \frac{C}{1+C}x =: G_1(x) \quad 0 < x < 1$$

and

(15)
$$\lim_{n \to \infty} P\left(\frac{\tilde{Y}_n}{bn} \le x | \tilde{Y}_n > 0\right) = \frac{C}{1+C} + \frac{1}{1+C} \left(1 - e^{-x}\right) =: G_2(x) \quad x > 0.$$

It is worth pointing out that, since $\lim_{x\to 1} G_1(x) = \lim_{x\to 0} G_2(x)$, the limiting distributions in (14) and (15) represent the two different types of non-degenerate trajectories of $\{\tilde{Y}_n\}$:

(A)
$$\tilde{Y}_n \sim n^{\eta_1}$$
, where $\eta_1 \in U(0,1)$ with probab. $\frac{C}{1+C}$;

(B)
$$\tilde{Y}_n \sim \eta_2 n$$
, where $\eta_2 \in Exp(b)$ with probab. $\frac{1}{1+C}$.

We will have a similar situation with the processes considered in Section 6.

Theorem 4.5 ([36]). Suppose (9), (13), and
$$p_n = dr_n$$
. If $\sum_{k=1}^{\infty} p_k < \infty$, then

$$\lim_{n \to \infty} P\left(\frac{\tilde{Y}_n}{bn} \le x | \tilde{Y}_n > 0\right) = 1 - e^{-x}, \quad x \ge 0.$$

Theorem 4.5 is an analog of the classical Kolmogorov–Yaglom result for GWPs. It turns out that the convergence of $\sum_{n=0}^{\infty} p_n$ and $\sum_{n=0}^{\infty} r_n$ ensure that the migration disappears without a trace so fast that the process with non-homogeneous migration has the same asymptotic behavior as the standard GWP.

5. Regenerative Branching Processes with Migration

Quoting [27], "A regenerative process is a stochastic process with the property that after some (usually) random time, it starts over in the sense that, from the random time on, the process is stochastically equivalent to what it was at the beginning". Regenerative processes can be intuitively seen as comprising of i.i.d. cycles. For classical regenerative processes, cycles and cycle lengths are i.i.d.

Consider a random vector (W,R) with non-negative and independent coordinates. The sequence of its i.i.d. copies (W_j,R_j) for $j=1,2,\ldots$ defines an alternating renewal process (e.g., [16]). The random variables W and R can be interpreted as the "working" and "repairing" time periods, respectively, of an operating system. Denote $S_0=0$ and for $n=1,2,\ldots$

$$S_n := \sum_{j=1}^n (R_j + W_j)$$
 and $N(t) := \max\{n \ge 0 : S_n \le t\}.$

Define

$$\sigma(t) := t - S_{N(t)} - R_{N(t)+1}, \qquad t \ge 0.$$

The random variable $\sigma(t)$ takes on positive or negative values depending on whether at t the system is working or repairing, respectively. Let associate with each W_i , $j \geq 1$ a cycle given by the process $\{Z_i(t), 0 \leq t \leq W_i\}$ such that

$$Z_j(0) = 0$$
, $Z_j(t) > 0$ for $0 < t < W_j$, $Z_j(W_j) = 0$.

Definition. An alternating regenerative process (ARP) is defined by

$$Z(t) := \left\{ \begin{array}{ll} Z_{N(t)+1}(\sigma(t)) & \textit{when } \sigma(t) \geq 0 & \textit{(the system is working)} \\ 0 & \textit{when } \sigma(t) < 0 & \textit{(the system is repairing)}. \end{array} \right.$$

Example. Recall the process with migration $\{Y_t, t=0,1,\ldots\}$ defined by (7). It is an ARP with $P(R=k) = \left[P(M_t^0=0)\right]^{k-1} \left[1-P(M_t^0=0)\right]$, a geometrically distributed repairing time. Consider the sequence $\{Y_{t,j}^0, j=1,2,\ldots\}$ of corresponding migration processes with absorption at 0 and let $W_j = tI_{\{Y_{t,j}^0>0\}}$ for $j \geq 0$. Thus, $\{Y_t\}$ is an ARP with cycle process $\{Y_{t,j}^0\}$. It regenerates whenever it visits state zero.

The migration process in the example above can be generalized as follows.

Definition. Define a regenerative branching process with migration by $X_0 = 0$ and for t = 1, 2, ...

$$X_t = \begin{cases} Y_{N(t)+1,\sigma(t)}^0 & when \ \sigma(t) \ge 0\\ 0 & when \ \sigma(t) < 0, \end{cases}$$

where $\{Y_{j,t}^0, j=1,2,\ldots\}$ are migration processes with absorbing state zero.

Note that, unlike the process with migration $\{Y_t\}$, in the generalized regenerative branching process with migration $\{X_t\}$ the repairing time periods R_j are not necessary geometrically distributed.

Possible Scenario. The queueing systems are good examples for discrete time regenerative processes. Consider a single-server queue with Poisson arrivals. The service periods are composed of a busy part (not-empty queue) W_i and

an idle part (empty queue) R_j . The customers arriving during the service time of a customer are her "offspring". The "immigrants" (probably from another customer pool) will be served in the end of the entire "generation". Alternatively, some "emigrants" may give up and leave the queue.

Let $\{X_t\}$ be critical and $0 < \theta < 1/2$, where θ is from (10). Assume that either E[R] is finite or $P(R > t) \sim L(t)t^{-\alpha}$ for $\alpha \in (1/2, 1]$, where L(t) is a s.v.f. Under some additional moment assumptions for the reproduction and migration, the following limiting results are obtained in G. Yanev, Mitov, and N. Yanev (2006). The proofs make use of theorems from Mitov and N. Yanev (2001) for regenerative processes.

(i) If "the working time dominates over the repairing time", i.e.,

$$0 \le c := \lim_{t \to \infty} \frac{P(R > t)}{P(W > t)} < \infty,$$

then for $x \geq 0$

$$\lim_{t \to \infty} P\left(\frac{X_t}{bt} \le x\right) = \frac{c}{c+1} + \frac{1}{c+1} \frac{1}{B(\theta, 1-\theta)} \int_0^1 y^{\theta-1} (1-y)^{-\theta} \left(1 - e^{-x/y}\right) dy,$$

where B(x, y) is the Beta function. The expected value of the limiting random variable is $\theta/(c+1)$.

(ii) If "the repairing time dominates over the working time", i.e.,

$$\lim_{t \to \infty} \frac{P(R > t)}{P(W > t)} = \infty,$$

then for $x \geq 0$

$$\lim_{t \to \infty} P\left(\frac{X_t}{bt} \le x \mid X_t > 0\right) = \frac{1}{B(\theta, \alpha)} \int_0^1 y^{\theta - 1} (1 - y)^{\alpha - 1} \left(1 - e^{-x/y}\right) dy.$$

Note that the distribution of the limiting random variable has a mixture of beta and exponential distributions and a mean of $\theta/(\theta + \alpha)$.

6. Controlled Branching Processes with Continuous State Space

A branching process with continuous state space models situations when it is difficult to count the number of individuals in the population, but a related non-negative variable (e.g., volume or weight) associated with the "individuals" is measured instead.

Let us make the following assumptions.

- (i) For fixed n, let $U_n := \{U_{i,n}, i \geq 1\}$ be a sequence of i.i.d., non-negative random variables and the double array $U := \{U_n, n \geq 1\}$ consists of independent sequences U_n , n=1,2,...
- (ii) Each of the stochastic processes $N_n := \{N_n(t), t \in T\}$, n=1,2,... has state space Z^+ , the set of non-negative integers. They are independent processes with stationary and independent increments (s.i.i.) and $N_n(0) = 0$ a.s. Here T is either $[0, \infty)$ or Z^+ .
- (iii) The sequence $V := \{V_n, n \ge 1\}$ consists of independent and non-negative random variables.
- (iv) The processes $N := \{N_n, n \ge 1\}$, U, and V are independent.
- (v) The random variable X_0 is non-negative and independent from all processes introduced in (i)-(iv).
- (vi) The components N_n and U_n for $n \geq 1$ of the processes N and U, respectively, are identically distributed.

The following class of branching processes is introduced by Adke and Gadag (1995) and studied by Rahimov (2007) and Rahimov and Al–Sabah (2008).

Definition. A controlled branching process with continuous state space is defined by the recursive relation

(16)
$$X_{n+1} = \sum_{i=1}^{N_{n+1}(X_n)} U_{i,n+1} + V_{n+1}, \quad n = 0, 1, \dots, \quad X_0 = 0.$$

Notice that if X_0 , U, and V are integer-valued, then $\{X_n\}$ is a CBP. If, in addition, we choose in (1) the index set to be $I = \{1, 2\}$ and the control functions to be $\varphi_{1,n}(k) = N_n(k)$, and $\varphi_{2,n}(k) \equiv 1$, we obtain (16).

It is proven in [1] that $\tilde{Z}_n := N_n(X_{n-1})$ for n = 1, 2, ... is a GWP with time-depended immigration given by

$$\tilde{Z}_{n+1} = \sum_{i=1}^{\tilde{Z}_n} \xi_{i,n+1} + \eta_{n+1}, \quad n = 0, 1, \dots, \quad \tilde{Z}_0 = 0,$$

where $\xi_{i,n+1} \stackrel{d}{=} N_{n+1}(U_n)$ and $\eta_{n+1} \stackrel{d}{=} N_{n+1}(V_n)$. Exploring this duality, Rahimov (2007) transferred results from GWPs with immigration to $\{X_n\}$. Below we present one theorem from [21] for the critical $\{X_n\}$. Denote $2\tilde{b} := Var[\xi_{i,n}]$, $\beta_n := E[\eta_n(\eta_n - 1)]$, and $\gamma_n := E[V_n]$. For simplicity, some of the assumptions of the next theorem are given in terms of moments of $\xi_{i,n}$ and η_n , which can be expressed as functions of the moments of N, U, and V (see [21]).

Theorem 6.1 ([21]). Suppose $E[N_1(1)]E[U_1] = 1$ and $\tilde{b} < \infty$. Assume

(i)
$$\beta_n = o(\gamma_n \log n) \to 0 \text{ as } n \to \infty$$
;

(ii)
$$\lim_{n \to \infty} \gamma_n \log n = 0$$
 and $\lim_{n \to \infty} \frac{\gamma_n n \log n}{\sum_{k=1}^n \gamma_k} = C$ for $0 \le C \le \infty$.

Then,

$$\lim_{n \to \infty} P\left(\frac{\log X_n}{\log n} \le x | X_n > 0\right) = \frac{C}{1 + C}x, \quad 0 < x < 1$$

and

$$\lim_{n \to \infty} P\left(\frac{X_n}{\tilde{h}_m} \le x | X_n > 0\right) = \frac{C}{1+C} + \frac{1}{1+C} \left(1 - e^{-x}\right) \quad x > 0.$$

The similarities between Theorem 4.4 and Theorem 6.1 are striking. The phenomenon of having different limiting distributions under different normalizations in GWPs with decreasing time-dependent immigration was first observed by Badalbaev and Rahimov (1978) (see also [20], p.109 and p.122).

7. Concluding Remarks

This survey is by no means exhaustive. Not included here are some classes CBPs such as: branching processes with barriers (see Zubkov (1972), Bruss (1978), Schuh (1976), Sevastyanov (1995)), CBPs with random environments, and the more recently introduced alternating branching processes (see Mayster (2005)).

Controlled branching processes are part of Sevastyanov's legacy. Over time, particular subclasses were introduced and studied in details. We paid special attention to the processes with migration, which have been a subject of systematical research investigations by the Bulgarian school in branching processes under the direction of its founder Professor Nikolay Yanev a.k.a. the Captain. Closed relations were established between CBP and other classes, e.g., two-sex processes

and population-size-dependent processes. There is no doubt, that CBPs have great potential as modeling tools. In my opinion, they deserve more attention from the branching processes' community.

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