# 2 Branching processes

Notation:  $\mathbb{N} := \{1, 2, \ldots\}, \mathbb{N}_0 := \{0, 1, 2, \ldots\}$ 

## 2.1 Definition and branching property

### 2.1.1 Def. (GWP)

Let  $Y_{nj}$ ,  $n \in \mathbb{N}_0$ ,  $j \in \mathbb{N}$ , be independent and identically distributed (iid) r.v. taking values in  $\mathbb{N}_0$ . The process  $(Z_n)_{n \in \mathbb{N}_0}$ , defined via  $Z_0 := 1$  and

$$Z_{n+1} := \sum_{j=1}^{Z_n} Y_{nj}, \qquad n \in \mathbb{N}_0,$$

is called <u>Bienaymé–Galton–Watson branching process</u> (GWP).  $Z_n$  is interpreted as the size of a population at generation n.

Define  $p_k := P(Z_1 = k) = P(Y_{01} = k), \ k \in \mathbb{N}_0.$ 

We call  $(p_k)_{k \in \mathbb{N}_0}$  the <u>reproduction distribution</u> or <u>offspring distribution</u>.

For  $i, j, i_{n-1}, \ldots, i_0 \in \mathbb{N}_0$  we have (as long as the conditional probability is defined)

$$\pi_{ij} := P(Z_{n+1} = j | Z_n = i, Z_{n-1} = i_{n-1}, \dots, Z_0 = i_0)$$
  
=  $P(Y_{n1} + \dots + Y_{ni} = j) = \sum_{\substack{j_1, \dots, j_i \in \mathbb{N}_0 \\ j_1 + \dots + j_i = j}} p_{j_1} \cdots p_{j_i} =: p_j^{*i},$ 

where  $p_j^{*i} = P(Y_{01} + \cdots + Y_{0i} = j)$  denotes the *i*-fold convolution of the offspring distribution at *j*. Thus,  $(Z_n)_{n \in \mathbb{N}_0}$  is a HMC with transition probabilities

$$\pi_{ij} := P(Z_{n+1} = j | Z_n = i) = p_j^{*i}, \quad i, j \in \mathbb{N}_0.$$

#### 2.1.2 Example (Poisson GWP)

Let  $\alpha > 0$ ,  $p_k := e^{-\alpha} \alpha^k / k!$ ,  $k \in \mathbb{N}_0$ . In this case,  $\pi_{ij} = e^{-\alpha i} (\alpha i)^j / j!$ ,  $i, j \in \mathbb{N}_0$ .

### 2.1.3 Example (Binary GWP)

Let  $p \in [0, 1]$ ,  $p_0 := 1 - p$  and  $p_2 := p$ . In this case,

$$\pi_{ij} = \begin{cases} 0 & \text{if } j \text{ is odd,} \\ {i \choose j/2} p^{j/2} (1-p)^{i-j/2} & \text{if } j \text{ is even.} \end{cases}$$

Thus,

$$P(Z_1 = 2i_1, \dots, Z_n = 2i_n) = \prod_{k=1}^n \left( \binom{2i_{k-1}}{i_k} p^{i_k} (1-p)^{2i_{k-1}-i_k} \right)$$

with the convention  $2i_0 = 1$ . For p = 1 we have  $Z_n = 2^n$  a.s. for all  $n \in \mathbb{N}_0$ .

#### 2.1.4 Lemma

Let  $(\Omega, \mathcal{F}, P)$  be a probability space,  $A, B_i \in \mathcal{F}$   $(i \in I, I \text{ at most countable})$ with  $B_i \cap B_j \forall i, j \in I$  with  $i \neq j$  and assume that  $B := \bigcup_{i \in I} B_i$  satisfies P(B) > 0. If for every  $i \in I$  with  $P(B_i) > 0$  the probability  $P(A|B_i)$  does not depend on i, i.e.  $P(A|B_i) =: c$  for all  $i \in I$  with  $P(B_i) > 0$ , then P(A|B) = c.

### Proof.

We have

$$P(A \cap B) = P(A \cap \bigcup_{i \in I} B_i) = \sum_{i \in I} P(A \cap B_i)$$
  
=  $\sum_{\substack{i \in I \\ P(B_i) > 0}} \underbrace{P(A|B_i)}_{=c} P(B_i) = c \sum_{\substack{i \in I \\ P(B_i) > 0}} P(B_i) = c P(B).$ 

Now divide by P(B) > 0.

#### 2.1.5 Lemma

Let  $(Z_n^{(l)})_{n \in \mathbb{N}_0}$ ,  $l \in \mathbb{N}$ , be independent GWPs, each distributed as  $Z := (Z_n)_{n \in \mathbb{N}_0}$ . Then, for any  $k \in \mathbb{N}$ , the process  $\widetilde{Z} := (\widetilde{Z}_n)_{n \in \mathbb{N}_0}$ , defined via  $\widetilde{Z}_n := \sum_{l=1}^k Z_n^{(l)}$  for all  $n \in \mathbb{N}_0$ , is a HMC with the same transition probabilities as Z.

### Proof.

Let  $j, i_1, \ldots, i_k \in \mathbb{N}_0$  with  $P(Z_n^{(1)} = i_1, \ldots, Z_n^{(k)} = i_k) > 0$ . Define  $i := i_1 + \cdots + i_k$ . The conditional probability

$$P(\tilde{Z}_{n+1} = j \mid Z_n^{(1)} = i_1, \dots, Z_n^{(k)} = i_k)$$

$$= \sum_{\substack{j_1, \dots, j_k \in \mathbb{N}_0 \\ j_1 + \dots + j_k = j}} P(Z_{n+1}^{(1)} = j_1, \dots, Z_{n+1}^{(k)} = j_k \mid Z_n^{(1)} = i_1, \dots, Z_n^{(k)} = i_k)$$

$$= \sum_{\substack{j_1, \dots, j_k \in \mathbb{N}_0 \\ j_1 + \dots + j_k = j}} \prod_{l=1}^k P(Z_{n+1}^{(l)} = j_l \mid Z_n^{(l)} = i_l) = \sum_{\substack{j_1, \dots, j_k \in \mathbb{N}_0 \\ j_1 + \dots + j_k = j}} \prod_{l=1}^k p_{j_l}^{*i_l} = p_j^{*i}$$

only depends via  $i = i_1 + \cdots + i_k$  on  $i_1, \ldots, i_k$ . Lemma 2.1.4.  $\Rightarrow$ 

$$P(\widetilde{Z}_{n+1} = j \mid \widetilde{Z}_n = i) = P(\widetilde{Z}_{n+1} = j \mid \bigcup_{\substack{i_1, \dots, i_k \in \mathbb{N}_0 \\ i_1 + \dots + i_k = i}} \{Z_n^{(1)} = i_1, \dots, Z_n^{(k)} = i_k\})$$
  
=  $p_j^{*i}$ ,  $i, j \in \mathbb{N}_0$ .

The calculation does not change, if the condition  $\widetilde{Z}_n = i$  is replaced by  $\widetilde{Z}_n = i, \widetilde{Z}_{n-1} = i_{n-1}, \ldots, \widetilde{Z}_0 = i_0. \Rightarrow \widetilde{Z}$  is a HMC with the same transition probabilities as Z.

In the following let  $\mathbb{N}_0^{\infty} := \times_{i \in \mathbb{N}} \mathbb{N}_0$ . For  $i \in \mathbb{N}$  let  $\pi_i : \mathbb{N}_0^{\infty} \to \mathbb{N}_0$  be the projection to the *i*-th component, i.e.  $\pi_i(k) = k_i$  for all  $k = (k_i)_{i \in \mathbb{N}} \in \mathbb{N}_0^{\infty}$ . Furthermore, let  $\mathcal{G}$  denote the smallest  $\sigma$ -algebra in  $\mathbb{N}_0^{\infty}$  such that all projections  $\pi_i, i \in \mathbb{N}$ , are measurable.  $\mathcal{G}$  is called the product- $\sigma$ -algebra of  $\mathbb{N}_0^{\infty}$ . It is easily seen that  $\mathcal{G} = \mathcal{F}(\pi_i, i \in \mathbb{N}) = \mathcal{F}(\{\pi_i^{-1}(A_i) : i \in \mathbb{N}, A_i \subseteq \mathbb{N}_0\}).$ 

### 2.1.6 Theorem (Branching Property)

For  $r \in \mathbb{N}_0$ ,  $k \in \mathbb{N}$  and  $A \in \mathcal{G}$ ,

$$P((Z_n)_{n>r} \in A \mid Z_r = k) = P((\widetilde{Z}_n)_{n \in \mathbb{N}} \in A), \qquad (*)$$

where  $\widetilde{Z}_n := \sum_{j=1}^k Z_n^{(j)}$  and  $(Z_n^{(j)})_{n \in \mathbb{N}_0}$ ,  $j \in \{1, \ldots, k\}$ , are independent *GWPs*, all distributed as  $Z := (Z_n)_{n \in \mathbb{N}_0}$ .

#### Proof.

For fixed  $r \in \mathbb{N}_0$  and  $k \in \mathbb{N}$  let  $\mathcal{D}$  be the set of all  $A \in \mathcal{G}$  satisfying (\*). It is easily seen that  $\mathcal{D}$  is a Dynkin system in  $\mathbb{N}_0^{\infty}$ . Consider the system  $\mathcal{E}$  of all A of the form  $A = A_1 \times \cdots \times A_m \times \mathbb{N}_0 \times \mathbb{N}_0 \times \cdots$  with  $m \in \mathbb{N}$  and  $A_1, \ldots, A_m \subseteq \mathbb{N}_0$ . Obviously,  $\mathcal{E}$  is a  $\cap$ -stable generator of  $\mathcal{G}$ , i.e.  $\mathcal{F}(\mathcal{E}) = \mathcal{G}$ . If we can verify that  $\mathcal{E} \subseteq \mathcal{D}$ , then the statement follows, since then  $\mathcal{G} = \mathcal{F}(\mathcal{E}) = \mathcal{D}(\mathcal{E}) \subseteq \mathcal{D}(\mathcal{D}) = \mathcal{D}$ . It remains to verify that (\*) holds for  $A \in \mathcal{E}$ . Each such A is a at most countable union of sets of the form

$$\{k_1\} \times \cdots \times \{k_m\} \times \mathbb{N}_0 \times \mathbb{N}_0 \times \cdots$$
 (\*\*)

with  $m \in \mathbb{N}$  and  $k_1, \ldots, k_m \in \mathbb{N}_0$ . Because of the  $\sigma$ -additivity of the two probability measures on the left-hand and right-hand side in (\*) it suffices to verify (\*) for sets of the form (\*\*). In this case the left-hand side in (\*) is equal to

$$P(Z_{r+1} = k_1, \dots, Z_{r+m} = k_m | Z_r = k)$$
  
=  $\prod_{n=1}^m P(Z_{r+n} = k_n | Z_{r+n-1} = k_{n-1}) = \prod_{n=1}^m \pi_{k_{n-1}, k_n},$ 

where  $k_0 := k$ . The right-hand side in (\*) is as well equal to

$$P(\widetilde{Z}_1 = k_1, \dots, \widetilde{Z}_m = k_m) = \prod_{n=1}^m P(\widetilde{Z}_n = k_n \,|\, \widetilde{Z}_{n-1} = k_{n-1}) = \prod_{n=1}^m \pi_{k_{n-1}, k_n},$$

since, by Lemma 2.1.5,  $\widetilde{Z}$  has the same transition probabilities as Z. **2.1.7 Corollary** 

For  $k, n \in \mathbb{N}_0$  and any function  $h : \mathbb{N}_0 \to [0, \infty)$ ,

$$E(h(Z_{n+1}) | Z_1 = k) = E\left(h\left(\sum_{j=1}^k Z_n^{(j)}\right)\right),$$

where  $(Z_n^{(j)})_{n \in \mathbb{N}_0}, j \in \mathbb{N}$ , are independent copies of Z.

For k = 0 both sides are equal to h(0). Assume now that  $k \in \mathbb{N}$ . For  $h = 1_B$  with  $B \subseteq \mathbb{N}_0$  the statement follows from Theorem 2.1.6 (branching property) with the choice  $A := \pi_n^{-1}(B) \in \mathcal{G}$ . Thus, the statement holds for elementary functions. If  $h : \mathbb{N}_0 \to [0, \infty)$  is arbitrary, then there exist elementary functions  $0 \le h_1 \le h_2 \le \cdots$  with  $\lim_{m\to\infty} h_m = h$ . The statement then follows by two-times applying the theorem of monotone convergence.

In order to compute the mean and the variance of  $Z_n$ , the following lemma will be useful.

### 2.1.8 Lemma

Let  $X_1, X_2, \ldots$  be iid  $\mathbb{N}_0$ -valued r.v. and let Y be a further  $\mathbb{N}_0$ -valued r.v. being independent of  $(X_n)_{n \in \mathbb{N}}$ . If g denotes the probability generation function (pgf) of  $X_1$  and h the pgf of Y, then  $S := \sum_{j=1}^{Y} X_j$  has the pgf  $h \circ g$  and  $E(S) = E(Y)E(X_1) \in [0,\infty]$ . If  $E(S) < \infty$  then Var(S) = $Var(Y)(E(X_1))^2 + E(Y)Var(X_1)$ .

### Proof.

Let  $s \in [0,1]$ . For  $k \in \mathbb{N}_0$ ,  $\mathbb{E}(s^S | Y = k) = \mathbb{E}(s^{X_1 + \dots + X_k} | Y = k) = \mathbb{E}(s^{X_1} \cdots s^{X_k}) = (\mathbb{E}(s^{X_1}))^k = (g(s))^k$ . Multiplication with P(Y = k) and summation over all  $k \in \mathbb{N}_0$  yields

$$E(s^{S}) = \sum_{k=0}^{\infty} E(s^{S} | Y = k) P(Y = k) = \sum_{k=0}^{\infty} (g(s))^{k} P(Y = k) = h(g(s)).$$

Thus, S has the pgf  $h \circ g$ . It follows that  $E(S) = (h \circ g)'(1) = h'(g(1))g'(1) = h'(1)g'(1) = E(Y)E(X_1)$  and

$$E(S(S-1)) = (h \circ g)''(1) = h''(g(1))(g'(1))^2 + h'(g(1))g''(1)$$
  
=  $h''(1)(g'(1))^2 + h'(1)g''(1)$   
=  $E(Y(Y-1))(E(X_1))^2 + E(Y)E(X_1(X_1-1)).$ 

Assume now that  $E(S) < \infty$ . Summation of  $E(S) - (E(S))^2 = E(Y)E(X_1) - (E(Y))^2(E(X_1))^2$  yields

$$Var(S) = E(Y(Y-1))(E(X_1))^2 + E(Y)E(X_1^2) - (E(Y))^2(E(X_1))^2$$
  
=  $E(Y^2)(E(X_1))^2 + E(Y)(E(X_1^2) - (E(X_1))^2) - (E(Y))^2(E(X_1))^2$   
=  $Var(Y)(E(X_1))^2 + E(Y)Var(X_1).$ 

Now let  $f_n$  denote the pgf of  $Z_n$ , i.e.

$$f_n(s) := \mathcal{E}(s^{Z_n}) = \sum_{k=0}^{\infty} P(Z_n = k) s^k, \quad s \in [0, 1].$$

Define  $f := f_1$ , i.e.  $f(s) = \sum_{k=0}^{\infty} p_k s^k$ ,  $s \in [0, 1]$ . Lemma 2.1.8 (applied with  $Y := Z_{n-1}$  and  $X_j := Y_{n-1,j}$ ) yields

$$f_n = f_{n-1} \circ f, \qquad n \in \mathbb{N},$$

and

$$\mathcal{E}(Z_n) = m \mathcal{E}(Z_{n-1}), \qquad n \in \mathbb{N},$$

where  $m := f'(1) = \sum_{k=1}^{\infty} k p_k = E(Z_1)$  is the expected number of offspring of any individual. Moreover, for  $m < \infty$ , Lemma 2.1.8 yields

$$\operatorname{Var}(Z_n) = \sigma^2 \operatorname{E}(Z_{n-1}) + m^2 \operatorname{Var}(Z_{n-1}), \qquad n \in \mathbb{N},$$

where  $\sigma^2 := \text{Var}(Z_1) = \sum_{k=1}^{\infty} k^2 p_k - m^2 = f''(1) + f'(1) - (f'(1))^2$  is the reproductive variance. In particular,

$$f_n = \underbrace{f \circ \cdots \circ f}_{n-\text{times}}$$

is the *n*-fold convolution of f and the mean of  $Z_n$  is  $E(Z_n) = m^n$ ,  $n \in \mathbb{N}_0$ . Moreover, if  $m < \infty$ , an induction on n shows that the variance of  $Z_n$  is

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

### 2.2 Extinction probability

Given: GWP  $Z = (Z_n)_{n \in \mathbb{N}_0}$  with  $Z_0 = 1$  and offspring distribution  $(p_k)_{k \in \mathbb{N}_0}$ . Notation:  $f_n := \text{pgf of } Z_n$ .

Known: 
$$f_n = \underbrace{f \circ \cdots \circ f}_{n-\text{times}}$$
, where  $f(s) := \sum_{k=0}^{\infty} p_k s^k$ ,  $s \in [0, 1]$   
$$m := \operatorname{E}(Z_1) = f'(1-) = \sum_{k=1}^{\infty} k p_k \in [0, \infty]$$

### **2.2.1 Def. (Extinction Probability)** The event

$$Q := \{Z_n = 0 \text{ eventually}\} := \liminf_{n \to \infty} \{Z_n = 0\} := \bigcup_{n \in \mathbb{N}} \bigcap_{m=n}^{\infty} \{Z_m = 0\}$$

is called the <u>extinction event</u> and

$$q := P(Q) = P(Z_n = 0 \text{ eventually}) = \lim_{n \to \infty} P(Z_n = 0) = \lim_{n \to \infty} f_n(0)$$
  
the extinction probability of Z.

#### 2.2.2 Theorem (Fixed Point Theorem)

The fixed point equation f(s) = s has exactly one solution in [0, 1) if m > 1and no solution in [0, 1) if  $m \leq 1$  and  $p_1 < 1$ . The extinction probability q is the smallest fixed point of f in the interval [0, 1].

#### Proof.

We exclude the trivial case  $p_1 = 1$ . From  $f_n(0) \to q$  and the continuity of f it follows that

$$f(q) \leftarrow f(f_n(0)) = f_{n+1}(0) \rightarrow q.$$

Thus, f(q) = q. Now let  $a \in [0, 1]$  be arbitrary with f(a) = a. By induction on  $n \in \mathbb{N}$  it follows that  $f_n(0) \leq a$ : For n = 1 this is clear, since f is nondecreasing and hence  $f_1(0) = f(0) \leq f(a)$ . The induction step from n to n+1 reads  $f_{n+1}(0) = f(f_n(0)) \leq f(a) = a$ . Letting  $n \to \infty$  yields  $q \leq a$ , i.e. q is the smallest solution of the equation f(s) = s in [0, 1].

Define  $\varphi(s) := f(s) - s, s \in [0, 1].$ 

Assume first that  $m \leq 1$  and  $p_1 < 1$ . Then, for all  $s \in [0, 1)$ ,

$$\varphi'(s) = f'(s) - 1 < f'(1) - 1 \leq 0,$$

i.e.  $\varphi$  is strictly decreasing. In particular,  $\varphi(s) > \varphi(1) = 0$ , so f(s) > s for all  $s \in [0, 1)$ . Therefore, the equation f(s) = s has no solution in [0, 1). Assume now that m > 1. Then,

$$\frac{1-f(s)}{1-s} \to f'(1) = m > 1, \qquad s \to 1.$$

Thus, 1 - f(s) > 1 - s, so  $\varphi(s) < 0$  for all s in a left neighborhood of 1. On the other hand,  $\varphi(0) = f(0) \ge 0$ . Thus, by the intermediate value theorem, there exists  $s \in [0,1)$  with  $\varphi(s) = 0$ , i.e. f(s) = s. To see that there is only one such s, assume that there exist  $0 \le s_1 < s_2 < 1$  with  $f(s_1) = s_1$ and  $f(s_2) = s_2$ . Then,  $\varphi(s_1) = \varphi(s_2) = 0 = \varphi(1)$ . Thus, by the theorem of Rolle, there exist a, b with  $s_1 < a < s_2 < b < 1$  and  $\varphi'(a) = \varphi'(b) = 0$ , i.e. f'(a) = f'(b), in contradiction to the fact that f' is strictly increasing<sup>1</sup> if m > 1. Thus, the assumption is wrong, so there exists exactly one  $s \in [0, 1)$ with f(s) = s.

#### 2.2.3 Theorem

If  $p_1 < 1$  then  $P(Z_n \to 0) + P(Z_n \to \infty) = 1$ .

<sup>&</sup>lt;sup>1</sup>Since m > 1 there exists  $k_0 \in \{2, 3, ...\}$  with  $p_{k_0} > 0$ . For all  $s \in (0, 1)$  it follows that  $f''(s) = \sum_{k=2}^{\infty} k(k-1)p_k s^{k-2} \ge k_0(k_0-1)p_{k_0} s^{k_0-1} > 0$ . Hence, f' is strictly increasing.

For q = 1 there is nothing to show. Thus, let q < 1. Then, by Theorem 2.2.2, m > 1 and f' is strictly increasing.

Assume that  $f'(q) \ge 1$ . Then, for all  $s \in (q, 1)$ ,  $f'(s) > f'(q) \ge 1$ , and hence  $f(1) - q = f(1) - f(q) = \int_q^1 f'(s) ds > \int_q^1 1 ds = 1 - q$ , so f(1) > 1, an obvious contradiction. Thus, f'(q) < 1.

Induction yields

$$f'_n(q) = (f'(q))^n, \qquad n \in \mathbb{N}.$$

For n = 1 this is clear, since  $f_1 = f$ . The step from n to n+1 reads  $f'_{n+1}(q) = (f_n \circ f)'(q) = f'_n(f(q))f'(q) = f'_n(q)f'(q) \stackrel{IV}{=} (f'(q))^n f'(q) = (f'(q))^{n+1}$ .

<u>Case 1:</u> Let  $q \in (0, 1)$ . Then, for all  $k, n \in \mathbb{N}$ ,

$$P(1 \le Z_n \le k) = \sum_{j=1}^k P(Z_n = j) \le \sum_{j=1}^k P(Z_n = j) \frac{jq^{j-1}}{q^k} \le \frac{f'_n(q)}{q^k} = \frac{(f'(q))^n}{q^k}$$
  
$$\Rightarrow \sum_{n=1}^\infty P(1 \le Z_n \le k) < \infty.$$

Borel–Cantelli lemma.  $\Rightarrow P(1 \leq Z_n \leq k \infty$ -often) = 0, and the assertion follows for q > 0.

<u>Case 2:</u> Assume now that q = 0. Then,  $p_0 = f(0) = f(q) = q = 0$  and hence  $Z_1 \leq Z_2 \leq \cdots$  almost surely. For each  $n \in \mathbb{N}$  it follows that

$$P\left(\bigcap_{m=n}^{\infty} \{Z_m = Z_{m+1}\}\right) = \lim_{N \to \infty} P\left(\bigcap_{m=n}^{n+N-1} \{Z_m = Z_{m+1}\}\right)$$
$$= \lim_{N \to \infty} \sum_{k=0}^{\infty} P(k = Z_n = Z_{n+1} = \dots = Z_{n+N})$$
$$= \lim_{N \to \infty} \sum_{k=0}^{\infty} P(Z_n = k) \underbrace{p_1^k \cdots p_1^k}_{N-\text{times}} \quad (\text{since } Z_1 \le Z_2 \le \dots \text{ a.s.})$$
$$= \lim_{N \to \infty} \sum_{k=0}^{\infty} P(Z_n = k) (p_1^N)^k = \lim_{N \to \infty} f_n(p_1^N) \stackrel{p_1 \le 1}{=} f_n(0).$$

Thus,

$$P(Z_n = Z_{n+1} \text{ eventually}) = P\left(\bigcup_{n \in \mathbb{N}} \bigcap_{m=n}^{\infty} \{Z_{m+1} = Z_m\}\right)$$
$$= \lim_{n \to \infty} P\left(\bigcap_{m=n}^{\infty} \{Z_m = Z_{m+1}\}\right) = \lim_{n \to \infty} f_n(0) = q = 0,$$

i.e.  $P(Z_n \to \infty) = P(Z_n < Z_{n+1} \text{ $\infty$-often$}) = 1 - P(Z_n = Z_{n+1} \text{ eventually}) = 1.$ 

#### 2.2.4 Def.

A GWP Z is called <u>subcritical</u> if m < 1, <u>critical</u> if m = 1 and <u>supercritical</u> if m > 1.

### 2.3 Critical case

Given: GWP  $Z = (Z_n)_{n \in \mathbb{N}_0}$  with  $Z_0 = 1$ It is assumed that m = 1 and  $p_1 < 1$ . Known:

- $q := P(Z_n \to 0) = 1$
- $E(Z_n) = m^n = 1$  for all  $n \in \mathbb{N}_0$
- $\operatorname{Var}(Z_n) = n\sigma^2 \to \infty$ Note that  $p_1 < 1$  is equivalent to  $\sigma^2 := \operatorname{Var}(Z_1) > 0$ .

### 2.3.1 Lemma (Basic Lemma)

If m = 1 and  $\sigma^2 \in (0, \infty)$  then

$$\lim_{n \to \infty} \frac{1}{n} \left( \frac{1}{1 - f_n(s)} - \frac{1}{1 - s} \right) = \frac{\sigma^2}{2}$$

uniformly for  $s \in [0, 1)$ .

### 2.3.2 Theorem (Yaglom Limit)

Let m = 1 and  $\sigma^2 \in (0, \infty)$ . Then

- (a)  $\lim_{n \to \infty} nP(Z_n > 0) = \frac{2}{\sigma^2}$ , (Kolmogorov, 1938)
- (b)  $\lim_{n \to \infty} \mathbb{E}\left(\frac{Z_n}{n} \mid Z_n > 0\right) = \frac{\sigma^2}{2}$ , and
- (c) Exponential limit law:

$$\lim_{n \to \infty} P\left(\frac{Z_n}{n} \le u \, \middle| \, Z_n > 0\right) = 1 - e^{-2u/\sigma^2}, \, u \ge 0.$$
 (Yaglom, 1947)

#### Rem.

Conditional on  $Z_n > 0$ , the r.v.  $Z_n/n$  converges in distribution to an exponential distribution with parameter  $2/\sigma^2$ .

(of Lemma 2.3.1) Let  $s \in [0, 1)$ .

Taylor expansion in 1:  $f(s) = s + \frac{\sigma^2}{2}(1-s)^2 + r(s)(1-s)^2$  for a continuous function r with  $\lim_{s\uparrow 1} r(s) = 0$ .  $\Rightarrow$ 

$$\frac{1}{1-f(s)} - \frac{1}{1-s} = \frac{f(s) - s}{(1-f(s))(1-s)} = \frac{\frac{\sigma^2}{2}(1-s)^2 + r(s)(1-s)^2}{(1-f(s))(1-s)}$$
$$= \frac{1-s}{1-f(s)} \left(\frac{\sigma^2}{2} + r(s)\right) = \frac{\sigma^2}{2} + \rho(s),$$

where again  $\rho$  is continuous with  $\lim_{s\uparrow 1} \rho(s) = 0$ . Iteration yields

$$\frac{1}{n} \left( \frac{1}{1 - f_n(s)} - \frac{1}{1 - s} \right) = \frac{1}{n} \sum_{j=0}^{n-1} \left( \frac{1}{1 - f(f_j(s))} - \frac{1}{1 - f_j(s)} \right)$$
$$= \frac{\sigma^2}{2} + \frac{1}{n} \sum_{j=0}^{n-1} \rho(f_j(s)).$$

The convergence  $f_n(s) \to 1$  is uniform in  $s \in [0, 1)$ , since  $f_n(0) \leq f_n(s) \leq 1$ and  $f_n(0) \to 1$ . The result follows since  $\rho$  is bounded.  $\Box$ 

### Proof.

(of Theorem 2.3.2)

(a) 
$$nP(Z_n > 0) = n(1 - f_n(0)) = \left(\frac{1}{n}\left(\frac{1}{1 - f_n(0)} - 1\right) + \frac{1}{n}\right)^{-1} \to \frac{2}{\sigma^2}$$
 by  
Lemma 2.3.1 (applied with  $s = 0$ ).  
(b)  $E\left(\frac{Z_n}{n} \mid Z_n > 0\right) = \frac{E(Z_n)}{n(1 - f_n(0))} = \frac{1}{nP(Z_n > 0)} \stackrel{(a)}{\to} \frac{\sigma^2}{2}$ .  
(c) Let  $u > 0$ . Define  $\beta := 2/\sigma^2$ .  
 $E(e^{-uZ_n/n} \mid Z_n > 0) = \frac{f_n(e^{-u/n}) - f_n(0)}{1 - f_n(0)} = 1 - \frac{1 - f_n(e^{-u/n})}{1 - f_n(0)}$   
 $= 1 - \frac{1}{nP(Z_n > 0)} \left(\frac{1}{n}\left(\frac{1}{1 - f_n(e^{-u/n})} - \frac{1}{1 - e^{-u/n}}\right) + \frac{1}{n(1 - e^{-u/n})}\right)^{-1}$ 
converges by (a) and Lemma 2.3.1 to

converges by (a) and Lemma 2.3.1 to 1 < 1 > -1 1  $\beta_{ac} = \beta$ 

$$1 - \frac{1}{\beta} \left( \frac{1}{\beta} + \frac{1}{u} \right)^{-1} = 1 - \frac{1}{\beta} \frac{\beta u}{\beta + u} = \frac{\beta}{\beta + u}$$

where the uniform convergence in Lemma 2.3.1 is essential here.

The map  $u \mapsto \frac{\beta}{\beta+u}$  is the Laplace transform (LT) of  $\text{Exp}(\beta)$ . By the continuity theorem for LT the pointwise convergence of the LTs implies the convergence in distribution.

### 2.4 Subcritical case

Given: GWP  $Z = (Z_n)_{n \in \mathbb{N}_0}$  with  $Z_0 = 1$  and  $m := \mathbb{E}(Z_1) < \infty$ 

Taylor expansion of the pgf f of  $Z_1$  at 1:

$$f(s) = 1 - m(1 - s) + r(s)(1 - s), \qquad s \in [0, 1].$$

**2.4.1 Lemma (Comparison Lemma)** For all  $\delta \in (0, 1)$ ,

$$\sum_{k=1}^{\infty} r(1-\delta^k) < \infty. \quad \Longleftrightarrow \quad \sum_{k=1}^{\infty} p_k k \log k < \infty.$$

Rem.

The condition on the right-hand side is equivalent to  $E(Z_1 \log Z_1) < \infty$ .

#### 2.4.2 Theorem (Kolmogorov, 1938)

If  $p_0 < 1$  and m < 1 then the limit

$$\varphi(0) := \lim_{n \to \infty} \frac{P(Z_n > 0)}{m^n}$$

exists with  $\varphi(0) = 0$  if  $E(Z_1 \log Z_1) = \infty$  and  $\varphi(0) > 0$  otherwise.

#### Rem.

The theorem thus states that  $P(Z_n > 0)$  behaves (up to a multiplicative constant) as  $m^n$  provided that  $E(Z_1 \log Z_1) < \infty$ .

The following result shows convergence conditional on non-extinction.

**2.4.3 Theorem (Convergence for subcritical GWPs, Yaglom, 1947)** If  $p_0 < 1$  and m < 1 then for each  $k \in \mathbb{N}$  the limit

$$b_k := \lim_{n \to \infty} P(Z_n = k \mid Z_n > 0)$$

exists and  $\sum_{k=1}^{\infty} b_k = 1$ , i.e.  $(b_k)_{k \in \mathbb{N}}$  defines a distribution on  $\mathbb{N}$ . The mean of this distribution is finite if and only if  $\mathbb{E}(Z_1 \log Z_1) < \infty$  and in this case

$$\sum_{k=1}^{\infty} k b_k = \frac{1}{\varphi(0)}.$$

The pgf  $g(s) := \sum_{k=1}^{\infty} b_k s^k$ ,  $s \in [0, 1]$ , is a solution to the equation

$$g(f(s)) = 1 - m(1 - g(s)), \quad s \in [0, 1].$$

(of Lemma 2.4.1) For all  $s \in [0, 1)$ ,

$$\begin{aligned} r(s) &= m - \frac{1 - f(s)}{1 - s} = m - \sum_{j \ge 0} s^j \left( 1 - \sum_{k \ge 0} p_k s^k \right) \\ &= m - \sum_{j \ge 0} s^j + \sum_{k \ge 0} p_k \sum_{j \ge 0} s^{j+k} = m - \sum_{n \ge 0} s^n + \sum_{k \ge 0} p_k \sum_{n \ge k} s^n \\ &= m - \sum_{n \ge 0} s^n + \sum_{n \ge 0} \left( \sum_{k=0}^n p_k \right) s^n = m - \sum_{n \ge 0} a_n s^n, \end{aligned}$$

where  $a_n := 1 - \sum_{k=0}^n p_k = \sum_{k>n} p_k$ ,  $n \in \mathbb{N}_0$ . Note that r(1) = 0 and, hence,  $\sum_{n\geq 0} a_n = m$ . In particular, r is a nonnegative nonincreasing function on [0, 1]. Define  $\alpha := -\log \delta$ . r nonincreasing.  $\Rightarrow$  For  $j \in \mathbb{N}$ 

$$r(1-\delta) + \int_{1}^{j} r(1-e^{-\alpha x}) \, \mathrm{d}x \ge \sum_{k=1}^{j} r(1-\delta^{k})$$
  
$$\ge \int_{1}^{j} r(1-e^{-\alpha x}) \, \mathrm{d}x = \frac{1}{\alpha} \int_{1-\delta}^{1-\delta^{j}} \frac{r(s)}{1-s} \, \mathrm{d}s,$$

where the last equality follows from the substitution  $s := 1 - e^{-\alpha x}$ . Thus,

$$\sum_{k \ge 1} r(1 - \delta^k) < \infty \quad \Longleftrightarrow \quad \int_0^1 \frac{r(s)}{1 - s} \, \mathrm{d}s < \infty$$

But, for all  $s \in [0, 1)$ ,

$$\frac{r(s)}{1-s} = \sum_{j\geq 0} s^{j} \left( m - \sum_{n\geq 0} a_{n} s^{n} \right) = m \sum_{j\geq 0} s^{j} - \sum_{n\geq 0} a_{n} \sum_{j\geq 0} s^{j+n}$$
$$= m \sum_{k\geq 0} s^{k} - \sum_{n\geq 0} a_{n} \sum_{k\geq n} s^{k} = m \sum_{k\geq 0} s^{k} - \sum_{k\geq 0} \left( \sum_{n=0}^{k} a_{n} \right) s^{k}$$
$$= \sum_{k\geq 0} \left( m - \sum_{n=0}^{k} a_{n} \right) s^{k} = \sum_{k\geq 0} \left( \sum_{n>k} a_{n} \right) s^{k}.$$

Integration yields

$$\int_0^1 \frac{r(s)}{1-s} \, \mathrm{d}s = \int_0^1 \sum_{k\ge 0} \left(\sum_{n>k} a_n\right) s^k \, \mathrm{d}s = \sum_{k\ge 0} \sum_{n>k} \frac{a_n}{k+1} = \sum_{n\ge 1} a_n \sum_{k=0}^{n-1} \frac{1}{k+1}.$$

Since  $\sum_{k=0}^{n-1} \frac{1}{k+1} \sim \log n$  as  $n \to \infty$ , this series converges if and only if the series  $\sum_{n\geq 1} a_n \log n$  converges. Now,

$$\sum_{n \ge 1} a_n \log n = \sum_{n \ge 1} \left( \sum_{k > n} p_k \right) \log n = \sum_{k \ge 2} p_k \sum_{n=1}^{k-1} \log n.$$

#### 2BRANCHING PROCESSES

Since  $\sum_{n=1}^{k-1} \log n \sim \int_1^k \log x \, dx = [x \log x - x]_1^k \sim k \log k$  as  $k \to \infty$ , this series converges if and only if  $\sum_{k>1} p_k k \log k < \infty$ . 

Proof.

(of Theorem 2.4.2) We have  $\frac{1-f(s)}{1-s} = m-r(s)$ . Replacing s by  $f_k(s)$  yields

$$\frac{1 - f_{k+1}(s)}{1 - f_k(s)} = m \left( 1 - \frac{r(f_k(s))}{m} \right)$$

and hence (taking products)  $\frac{1-f_n(s)}{1-s} = m^n \prod_{k=0}^{n-1} \left(1 - \frac{r(f_k(s))}{m}\right).$ 

 $0 \leq r/m \leq 1. \Rightarrow m^{-n}(1 - f_n(s))/(1 - s)$  is nonincreasing in n and hence converges to a limit  $\varphi(s) \geq 0$ . In particular (choose s = 0)  $P(Z_n > 0) =$  $1 - f_n(0) \sim m^n \varphi(0)$ . The well-known relation between convergence of sums and products shows that  $\varphi(0) > 0$  if and only if  $\sum_{k \ge 1} r(f_k(0)) < \infty$ . Now,  $1 - f(s) \le m(1 - s)$  and, by induction,  $1 - f_k(s) \le m^k(1 - s)$  for all  $k \in \mathbb{N}$ . Similarly it follows that  $1 - f_k(s) \ge (f'(s_0))^k (1-s)$  for  $s \ge s_0$  and with  $s_0 = p_0 > 0$  it follows with the notation  $a := f'(p_0) > 0$  that

$$1 - m^k \leq f_k(0) = f_{k-1}(p_0) \leq 1 - a^{k-1}(1 - p_0) \leq 1 - b^k,$$

where  $b := a \wedge (1 - p_0)$ . From Lemma 2.4.1 it follows that

$$\sum_{k\geq 1} r(f_k(0)) < \infty \quad \Longleftrightarrow \quad \sum_{k\geq 1} p_k k \log k < \infty. \qquad \Box$$

Proof.

(of Theorem 2.4.3) Define

$$g_n(s) := E(s^{Z_n} | Z_n > 0) = \frac{f_n(s) - f_n(0)}{1 - f_n(0)} = 1 - \frac{1 - f_n(s)}{1 - f_n(0)}$$
$$= 1 - (1 - s) \prod_{k=0}^{n-1} \frac{m - r(f_k(s))}{m - r(f_k(0))}.$$

We have  $f_k(s) \ge f_k(0)$  and r is nonincreasing.  $\Rightarrow$  The fraction in the product is greater than or equal to 1. Thus,  $g_n(s)$  is nonincreasing in n and, hence, converges to some g(s). Obviously,  $g_n(0) = 0$  and  $g_n(1) = 1$  and, hence, g(0) = 0 and g(1) = 1. In order to verify that g is continuous at 1, it suffices (by the monotonicity of q and since  $f_k(0) \to q = 1$ ) to verify that  $\lim_{k\to\infty} g(f_k(0)) = 1$ . We have

$$g_n(f_k(0)) = 1 - \frac{1 - f_n(f_k(0))}{1 - f_n(0)} = 1 - \frac{1 - f_k(f_n(0))}{1 - f_n(0)} \to 1 - m^k, \quad n \to \infty.$$

#### 2 BRANCHING PROCESSES

Thus,  $g(f_k(0)) = 1 - m^k$  and hence  $\lim_{k\to\infty} g(f_k(0)) = \lim_{k\to\infty} (1 - m^k) = 1$ . Therefore, g(1-) = 1, so g is continuous at 1. By the continuity theorem for pgf's, all the limits

$$b_k := \lim_{n \to \infty} P(Z_n = k \mid Z_n > 0), \qquad k \in \mathbb{N},$$

exist and g is as well a pgf of the form  $g(s) = \sum_{k \ge 1} b_k s^k$ . It follows that

$$\sum_{k \ge 1} kb_k = g'(1-) = \lim_{k \to \infty} \frac{1 - g(f_k(0))}{1 - f_k(0)} = \lim_{k \to \infty} \frac{m^k}{1 - f_k(0)} \stackrel{\text{Thm. 2.4.2}}{=} \frac{1}{\varphi(0)}$$

and  $g_n \circ f = 1 - \frac{1 - f_{n+1}}{1 - f_n(0)} = 1 - \frac{1 - f_{n+1}}{1 - f_{n+1}(0)} \frac{1 - f(f_n(0))}{1 - f_n(0)}$ . Letting  $n \to \infty$  yields  $g \circ f = 1 - (1 - g)m$ .

### 2.5 Supercritical case

**Given.**  $(Z_n)_{n \in \mathbb{N}_0}$  GWP with reproduction r.v.  $Z_1$ , where  $m := \mathrm{E}(Z_1) \in (1, \infty)$  and  $\sigma^2 := \mathrm{Var}(Z_1) \in (0, \infty]$ .  $q := P(Z_n \to 0)$  (extinction probability)

**2.5.1 Theorem (Convergence Theorem for Supercritical GWPs)** Under the above assumptions there exist positive numbers  $k_1, k_2, \ldots$  such that  $W_n := k_n Z_n$  converges as  $n \to \infty$  almost surely to a non-degenerate nonnegative real r.v. W. Moreover, P(W = 0) = q.

If  $a \in [0, 1/m)$  then  $a^n Z_n \to 0$  a.s.. If  $a \in (1/m, \infty)$  then  $a^n Z_n \to Z_\infty$  a.s., where  $Z_\infty(\omega) := 0$  for  $\omega \in \{Z_n \to 0\}$ and  $Z_\infty(\omega) := \infty$  for  $\omega \in \{Z_n \to \infty\}$ . If  $E(Z_1 \log Z_1) < \infty$  then one can choose  $k_n := m^{-n}$ .

If  $E(Z_1 \log Z_1) = \infty$  then  $m^{-n}Z_n \to 0$  a.s. as  $n \to \infty$ .

#### Rem.

The numbers  $k_n$ ,  $n \in \mathbb{N}$ , are called <u>Seneta constants</u> (Seneta, 1968). In particular, for  $E(Z_1 \log Z_1) < \infty$  and  $k_n = m^{-n}$ , one speaks of the <u>Theorem of Kesten and Stigum</u> (1966). Heyde (1970) has also provided important contributions to the convergence properties of supercritical GWPs.

#### 2.5.2 Theorem (Characterization of the limit W)

The LT  $\psi$  of W is a solution to the equation

$$\psi(mu) = (f \circ \psi)(u), \quad u \ge 0.$$

The mean  $E(W) = -\psi'(0)$  is finite if and only if  $E(Z_1 \log Z_1) < \infty$  and in this case there exists exactly one solution  $\psi$  of the above equation, which satisfies  $\psi(0) = 1$  and whose derivative at 0 exists and is equal to a given value.

### Recapitulation.

 $p_k := P(Z_1 = k), \ k \in \mathbb{N}_0.$ pgf f of  $Z_1, \ f(s) := \mathbb{E}(s^{Z_1}) = \sum_{k=0}^{\infty} p_k s^k, \ s \in [0, 1].$ Extinction probability  $q := \lim_{n \to \infty} P(Z_n = 0) < 1.$  q = smallest fixed point of f in the interval [0, 1].

### Proof.

(of Theorem 2.5.1) f strictly increasing.  $\Rightarrow g := f^{-1}$  exists. Define  $g_0 := \text{id}, g_n := \underbrace{g \circ \cdots \circ g}_{n-\text{times}}, n \in \mathbb{N}.$ 

g is non-decreasing, concave, differentiable and maps [q, 1] to [q, 1]. Define  $X_n(s) := (g_n(s))^{Z_n}$  and  $\mathcal{F}_n := \mathcal{F}(Z_1, \ldots, Z_n), s \in [q, 1], n \in \mathbb{N}_0$ .  $X_n(s)$  is  $\mathcal{F}_n$ -measurable. Moreover, for  $k \in \mathbb{N}_0$ ,

$$E(X_{n+1}(s) | Z_n = k) = E((g_{n+1}(s))^{Z_{n+1}} | Z_n = k)$$
  

$$= E((g_{n+1}(s))^{Y_{n1}+\dots+Y_{nk}} | Z_n = k)$$
  

$$= E((g_{n+1}(s))^{Y_{n1}+\dots+Y_{nk}})$$
  

$$= E((g_{n+1}(s))^{Y_{n1}}) \cdots E((g_{n+1}(s))^{Y_{nk}})$$
  

$$= (f(g_{n+1}(s)))^k = (g_n(s))^k.$$

 $\Rightarrow \operatorname{E}(X_{n+1}(s) \mid Z_n) = (g_n(s))^{Z_n} = X_n(s) \text{ a.s.}$  $\Rightarrow \operatorname{E}(X_{n+1}(s) \mid \mathcal{F}_n) = X_n(s) \text{ a.s.}$ 

 $\Rightarrow (X_n(s))_{n \in \mathbb{N}_0} \text{ is a nonnegative martingale w.r.t. } F := (\mathcal{F}_n)_{n \in \mathbb{N}_0}.$   $\Rightarrow X_{\infty}(s) := \lim_{n \to \infty} X_n(s) \text{ exists a.s. (martingale convergence theorem)}$ Clear:  $0 \le X_{\infty}(s) \le 1$ , since  $0 \le X_n(s) \le 1 \forall n \in \mathbb{N}_0.$ dominated convergence.  $\Rightarrow E(X_{\infty}(s)) = E(X_1(s)) = E(X_0(s)) = s$  a.s.  $E(X_{n+1}^2(s) \mid \mathcal{F}_n) \ge (E(X_{n+1}(s) \mid \mathcal{F}_n))^2 = X_n^2(s)$  a.s.  $\Rightarrow (X_n^2(s))_{n \in \mathbb{N}_0}$  submartingale w.r.t. F (again with values in [0, 1]).  $\Rightarrow E(X_{\infty}^2(s)) \ge E(X_1^2(s)) > (E(X_1(s)))^2$ , since  $Z_1$  is non-degenerate.  $\Rightarrow \operatorname{Var}(X_{\infty}(s)) \ge \operatorname{Var}(X_1(s)) > 0.$  Define  $c_n(s) := -\log g_n(s), Y(s) := -\log X_{\infty}(s).$   $\Rightarrow \boxed{c_n(s)Z_n \stackrel{a.s.}{\rightarrow} Y(s)}$  and Y(s) is non-degenerate, which proves the first part of Theorem 2.5.1, except that it remains to verify that Y(s) is a.s. finite.  $f(s) \leq s$  for  $s \in [q, 1]. \Rightarrow g(s) \geq s$  for  $s \in [q, 1]. \Rightarrow g_n \nearrow g_{\infty}$  for some  $g_{\infty}$ .  $s = f_n(g_n(s)) \leq f_n(g_{\infty}(s)) \rightarrow q$ , if  $g_{\infty}(s) < 1. \Rightarrow g_{\infty}(s) = 1$  for s > q. Taylor expansion of f around 1 (as in the critical case).  $\Rightarrow 1 - f(s) = (m - r(s))(1 - s).$ 

Replacing  $s \in (q, 1)$  by g(s) yields

$$\frac{1-g(s)}{1-s} = \frac{1}{m-r(g(s))} = \frac{1}{m} \frac{1}{1-\frac{r(g(s))}{m}}$$

Repeating this and taking products.  $\Rightarrow$ 

$$m^{n}(1 - g_{n}(s)) = \frac{1 - s}{\prod_{k=1}^{n} \left(1 - \frac{r(g_{k}(s))}{m}\right)}.$$
 (\*)

This tells us something on  $c_n(s)$ , since  $-\log x \sim 1 - x$  for  $x \to 1$ . In particular

$$\frac{c_n(s)}{c_{n-1}(s)} \sim \frac{1 - g_n(s)}{1 - g_{n-1}(s)} = \frac{1}{m} \frac{m^n (1 - g_n(s))}{m^{n-1} (1 - g_{n-1}(s))} = \frac{1}{m} \frac{1}{1 - \frac{r(g_n(s))}{m}} \sim \frac{1}{m},$$

since  $g_n(s) \to g_{\infty}(s) = 1$  and r(1-) = 0. Now use this to verify that  $Y(s) := \lim_{n \to \infty} c_n(s)Z_n$  is a.s. finite.

Again, we have, with the notation  $P(A | Z_1) := E(1_A | Z_1) := E(1_A | \mathcal{F}(Z_1))$ 

$$\begin{split} P(Y(s) < \infty) &= \operatorname{E}(P(Y(s) < \infty \mid Z_1)) \\ &= \operatorname{E}\left(P\left(\lim_{n \to \infty} c_n(s)Z_n < \infty \mid Z_1\right)\right) \\ \stackrel{(+)}{=} \operatorname{E}\left(\left(P\left(\lim_{n \to \infty} c_n(s)Z_{n-1} < \infty\right)\right)^{Z_1}\right) \quad ((+) \text{ follows from the branching property, see Appendix 1}) \\ &= \operatorname{E}\left(\left(P\left(\lim_{n \to \infty} \frac{c_n(s)}{c_{n-1}(s)}c_{n-1}(s)Z_{n-1} < \infty\right)\right)^{Z_1}\right) \\ &= \operatorname{E}\left(\left(P\left(\frac{Y(s)}{m} < \infty\right)\right)^{Z_1}\right) \\ &= \operatorname{E}\left((P(Y(s) < \infty))^{Z_1}\right) = f(P(Y(s) < \infty)). \end{split}$$

Analogous: P(Y(s) = 0) = f(P(Y(s) = 0)).

#### 2 BRANCHING PROCESSES

⇒ The probabilities  $P(Y(s) < \infty)$  and P(Y(s) = 0) are both fixed points of f and can hence only be equal to q or to 1. Y(s) is non-degenerate. ⇒ P(Y(s) = 0) = q.

$$s = \operatorname{E}(X_{\infty}(s)) = \operatorname{E}(e^{-Y(s)}) \leq P(Y(s) < \infty)$$

 $\Rightarrow P(Y(s) < \infty) = 1$  for s > q. We have

$$m^n c_n(s) \not\to \infty \iff \prod_{n=1}^{\infty} \left( 1 - \frac{r(g_n(s))}{m} \right) > 0 \iff \sum_{n=1}^{\infty} r(g_n(s)) < \infty.$$

Choose  $s_0 \in (q, 1)$  such that  $m_0 := f'(s_0) > 1$  and k such that  $g_k(s) \ge s_0$ . Since

$$m_0^n(1-s) \le 1 - f_n(s) \le m^n(1-s) \ \forall \ s \in [s_0, 1],$$

it follows that

$$1 - m_0^{-(n-k)}(1 - g_k(s)) \le g_n(s) \le 1 - m^{-n}(1 - s).$$

By the comparison lemma,

$$\lim_{n \to \infty} m^n c_n(s) < \infty. \iff \mathcal{E}(Z_1 \log Z_1) < \infty. \quad (**)$$

If  $E(Z_1 \log Z_1) < \infty$ , then we can hence choose  $k_n := m^{-n}$ , since then  $k_n Z_n = \frac{c_n(s)Z_n}{m^n c_n(s)}$  converges a.s..

If  $E(Z_1 \log Z_1) = \infty$ , then  $m^{-n}Z_n = \frac{c_n(s)Z_n}{m^n c_n(s)} \to 0$  a.s.. If a < 1/m, then always  $a^n Z_n = \frac{(am)^n}{m^n c_n(s)} c_n(s) Z_n \to 0$  a.s.. Assume now that  $a \in (1/m, \infty)$ . Then,

$$\frac{c_n(s)}{a^n} \sim \frac{1 - g_n(s)}{a^n} \stackrel{(*)}{=} \frac{1 - s}{\prod_{k=1}^n a(m - r(g_k(s)))}$$

and this expression has to converge to 0 for  $s \in (q, 1)$ , since

$$1 = m - r(q) \leq m - r(g_k(s)) \rightarrow m.$$

Therefore,

$$a^n Z_n = \frac{a^n}{c_n(s)} c_n(s) Z_n \rightarrow \begin{cases} 0 & \text{if } Z_n \to 0, \\ \infty & \text{if } Z_n \to \infty. \end{cases}$$

#### Appendix 1 to (+): We have

$$\{\lim_{n \to \infty} c_n(s) Z_n < \infty\} = \{ (Z_n)_{n > 1} \in A_s \},\$$

where

 $A_s := \{a = (a_n)_{n>1} \in \mathbb{N}_0^{\infty} \mid \text{The sequence } (c_n(s)a_n)_{n>1} \text{ converges in } \mathbb{R} \}.$ 

For n > 1 let  $\pi_n : \mathbb{N}_0^{\infty} \to \mathbb{N}_0$  denote the projection to the *n*-th component, i.e.  $\pi_n(a) = a_n$  for all n > 1 and  $a = (a_n)_{n>1} \in \mathbb{N}_0^{\infty}$ . Further, let  $\mathcal{G} := \mathcal{F}(\pi_n, n > 1)$  denote the product- $\sigma$ -algebra. Then,

$$A_{s} = \{a = (a_{n})_{n>1} \in \mathbb{N}_{0}^{\infty} \mid (c_{n}(s)a_{n})_{n>1} \text{ is a Cauchy-sequence in } \mathbb{R} \}$$
  
$$= \bigcap_{N \in \mathbb{N}} \bigcup_{n_{0} \in \mathbb{N}} \bigcap_{i,j>n_{0}} \{a = (a_{n})_{n>1} \in \mathbb{N}_{0}^{\infty} \mid |c_{i}(s)\pi_{i}(a) - c_{j}(s)\pi_{j}(a)| < \frac{1}{N} \}$$
  
$$= \bigcap_{N \in \mathbb{N}} \bigcup_{n_{0} \in \mathbb{N}} \bigcap_{i,j>n_{0}} (c_{i}(s)\pi_{i} - c_{j}(s)\pi_{j})^{-1}((-\frac{1}{N}, \frac{1}{N})) \in \mathcal{G},$$

since, with  $\pi_i$  and  $\pi_j$ , also the map  $c_i(s)\pi_i - c_j(s)\pi_j : \mathbb{N}_0^{\infty} \to \mathbb{R}$  is  $\mathcal{G}$ - $\mathcal{B}$ -measurable.

Now, let  $k \in \mathbb{N}_0$  and let  $(Z_n^{(j)})_{n \in \mathbb{N}_0}$ ,  $j \in \{1, \ldots, k\}$ , be independent GWPs, each distributed as  $(Z_n)_{n \in \mathbb{N}_0}$ . Define  $\widetilde{Z}_n := \sum_{j=1}^k Z_n^{(j)}$ . Then,

$$P(\lim_{n \to \infty} c_n(s)Z_n < \infty \mid Z_1 = k) = P((Z_n)_{n>1} \in A_s \mid Z_1 = k)$$
  
=  $P((\widetilde{Z}_n)_{n \in \mathbb{N}} \in A_s)$  (branching property, see before)  
=  $P(\lim_{n \to \infty} c_n(s)\widetilde{Z}_{n-1} < \infty) = P(\sum_{j=1}^k \lim_{n \to \infty} c_n(s)Z_{n-1}^{(j)} < \infty)$   
=  $P(\bigcap_{j=1}^k \{\lim_{n \to \infty} c_n(s)Z_{n-1}^{(j)} < \infty\}) = (P(\lim_{n \to \infty} c_n(s)Z_{n-1} < \infty))^k.$ 

Therefore,

$$P(\lim_{n \to \infty} c_n(s)Z_n < \infty \mid Z_1) = (P(\lim_{n \to \infty} c_n(s)Z_{n-1} < \infty))^{Z_1} \quad \text{a.s}$$

and taking the mean yields (+).

#### Proof.

(of Theorem 2.5.2) Let  $s \in (q, 1)$ . Known (from the previous proof):

$$\frac{c_{n+1}(s)}{c_n(s)} \to \frac{1}{m}. \text{ With } n \to \infty \text{ it follows for } u \ge 0$$

$$\psi(mu) = \mathbb{E}(e^{-muY(s)})$$

$$\leftarrow \mathbb{E}(e^{-muc_{n+1}(s)Z_{n+1}})$$

$$= \mathbb{E}(\mathbb{E}(e^{-muc_{n+1}(s)Z_{n+1}} | Z_1))$$

$$= \mathbb{E}\left(\left(\mathbb{E}\left(e^{-um(\frac{c_{n+1}(s)}{c_n(s)}c_n(s)Z_n}\right)\right)^{Z_1}\right)$$

(follows from the branching property,

see Theorem 2.1.6 and Corollary 2.1.7)

$$\rightarrow f(\psi(u)),$$

where the theorem of dominated convergence was used several times.

The substitution  $u \mapsto u/m$  and an application of  $g := f^{-1}$  yields  $\psi(u/m) = g(\psi(u))$ . Since  $\psi(u) \ge \lim_{u \to \infty} \psi(u) = \lim_{u \to \infty} \mathbb{E}(e^{-uY(s)}) = P(Y(s) = 0) = q$  one can iterate this to

$$1 - \psi(u/m^n) = 1 - g_n(\psi(u)) = O(m^{-n}) \stackrel{(**)}{\longleftrightarrow} \operatorname{E}(Z_1 \log Z_1) < \infty.$$

 $\psi$  convex.  $\Rightarrow$  The map  $h \mapsto \frac{1-\psi(h)}{h}$  is non-increasing on  $(0, \infty)$ .  $\Rightarrow$  The lefthand side above is equivalent to  $\frac{1-\psi(h)}{h} = O(1)$ , i.e. equivalent to the existence of the limit  $\lim_{h\to 0} \frac{1-\psi(h)}{h} < \infty$ , i.e. equivalent to the property, that  $-\psi'(0)$ exists and is finite. This is well-known (see Appendix 2) to be equivalent to  $E(Y(s)) < \infty$  and in this case the equality  $E(Y(s)) = -\psi'(0)$  holds. Therefore, the second assertion follows.

To prove the uniqueness statement let  $\psi$  and  $\phi$  be two solutions with  $\psi(0) = \phi(0)$  finite and  $\psi'(0) = \phi'(0)$  finite. Then, for any u > 0

$$\begin{aligned} |\psi(u) - \phi(u)| &= |f(\psi(u/m)) - f(\phi(u/m)| \leq m |\psi(u/m) - \phi(u/m)| \\ &\leq \cdots \leq m^n |\psi(u/m^n) - \phi(u/m^n)| \\ &= u \left| \frac{\psi(u/m^n) - \psi(0) - (\phi(u/m^n) - \phi(0))}{u/m^n} \right| \\ &\to u |\psi'(0) - \phi'(0)| = 0. \end{aligned}$$

#### Appendix 2

#### 2.5.3 Lemma

Let X be a nonnegative real r.v. and  $\psi : [0, \infty) \to (0, 1]$  the LT of X. Then, the mean E(X) is finite if and only if the derivative  $\psi'(0)$  of  $\psi$  at 0 (in  $\mathbb{R}$ ) exists. In this case,  $E(X) = -\psi'(0)$ .

'⇒': Let  $E(X) < \infty$ . Define  $f : [0, \infty) \to [0, \infty)$  via  $f(u) := E(Xe^{-uX})$  for all  $u \in [0, \infty)$ . Then, obviously,  $-\psi$  is a antiderivative of f. By the mean value theorem there exists for each h > 0 a mean-value  $\xi \in [0, h]$  with  $1 - \psi(h) = -\psi(h) - (-\psi(0)) = \int_0^h f(u) du = f(\xi)h$ . Thus,

$$\frac{1 - \psi(h)}{h} = f(\xi) = \mathcal{E}(X e^{-\xi X}).$$

Letting  $h \to 0$  (and hence also  $\xi \to 0$ ) yields (on the left-hand side by the def. of the derivative of  $\psi$  in 0 and on the right-hand side by dominated convergence)  $-\psi'(0) = E(X) < \infty$ .

' $\Leftarrow$ ': Conversely, assume that  $-\psi'(0) < \infty$ . Then,  $\psi'$  is defined on the full interval  $[0,\infty)$  and

$$-\psi'(u) = \mathcal{E}(Xe^{-uX}), \qquad u \in [0,\infty).$$

Letting  $u \to 0$  yields (on the left-hand side by the def. of the right-sided limit and on the right-hand side by dominated convergence)

$$-\psi'(0+) = \mathcal{E}(X).$$

On the other hand, the map  $u \mapsto -\psi'(u) = \mathbb{E}(Xe^{-uX})$  is non-increasing on  $[0, \infty)$ . Thus,

$$-\psi'(u) \leq -\psi'(0)$$

Letting  $u \to 0$  yields  $E(X) = -\psi'(0+) \le -\psi'(0) < \infty$ .