

# A Simple Path to Biggins' Martingale Convergence for Branching Random Walk

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**Abstract.** We give a simple non-analytic proof of Biggins' theorem on martingale convergence for branching random walks.

Let  $\mathcal{L} := \{X_i\}_{i=1}^L$  be a random  $L$ -tuple of real numbers, where  $L$  is also random and can take the values 0 and  $\infty$ . This can also be thought of as an ordered point process on  $\mathbf{R}$ . The random variable  $\mathcal{L}$  is used as the basis for construction of a branching random walk in the usual way: An initial particle at the origin of  $\mathbf{R}$  gives birth to  $L$  particles with displacements  $X_1, X_2, \dots$ . Then each of these particles gives birth to a random number of particles with random displacements from its new position according to the same law as  $\mathcal{L}$  and independently of one another and of the initial displacements. This continues in a like manner forever or until there are no more particles. For a particle  $\sigma$ , write  $|\sigma|$  for the generation in which  $\sigma$  is born,  $X(\sigma)$  for its displacement from its parent, and  $S(\sigma)$  for its position. Denote the initial particle by 0, also known as the root of the family tree. If  $\tau$  is an ancestor of  $\sigma$ , write  $\tau < \sigma$ . Thus, we have

$$S(\sigma) = \sum_{0 < \tau \leq \sigma} X(\tau).$$

Also, write  $\mathcal{L}(\sigma)$  for the copy of  $\mathcal{L}$  used to generate the children of  $\sigma$ . Let  $q$  be the extinction probability of the underlying Galton-Watson process.

For  $\alpha \in \mathbf{R}$ , define  $\langle \alpha, \mathcal{L} \rangle := \sum_{i=1}^L e^{-\alpha X_i}$  and  $m(\alpha) := \mathbf{E}[\langle \alpha, \mathcal{L} \rangle] \in (0, \infty]$ . Assume that  $m(0) > 1$ , so that  $q < 1$ . If  $m(\alpha) < \infty$  for some  $\alpha$ , then the sequence

$$W_n(\alpha) := \frac{\sum_{|\sigma|=n} e^{-\alpha S(\sigma)}}{m(\alpha)^n}$$

is a martingale with a.s. limit  $W(\alpha)$ . Write

$$m'(\alpha) := \mathbf{E} \left[ \sum_{i=1}^L X_i e^{-\alpha X_i} \right]$$

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when this exists in  $[-\infty, \infty]$  as a Lebesgue integral. Biggins (1977) has determined when  $W(\alpha)$  is nontrivial:

**BIGGINS' THEOREM.** *Suppose that  $\alpha \in \mathbf{R}$  is such that  $m(\alpha) < \infty$  and  $m'(\alpha)$  exists and is finite. Then the following are equivalent:*

- (i)  $\mathbf{P}[W(\alpha) = 0] = q$ ;
- (ii)  $\mathbf{P}[W(\alpha) = 0] < 1$ ;
- (iii)  $\mathbf{E}[W(\alpha)] = 1$ ;
- (iv)  $\mathbf{E}[\langle \alpha, \mathcal{L} \rangle \log^+ \langle \alpha, \mathcal{L} \rangle] < \infty$  and  $\alpha m'(\alpha)/m(\alpha) < \log m(\alpha)$ .

**REMARK.** In fact, the hypotheses here are very slightly weaker than those of Biggins (1977), Lemma 5. Moreover, the proof to follow works without the assumption that  $m'(\alpha)$  be finite, except for the implication (ii)  $\Rightarrow$  (iv), where it needs the assumption that  $\alpha m'(\alpha) \neq -\infty$ .

**REMARK.** The case of Biggins' Theorem where  $L$  is constant,  $X_i$  are independent and identically distributed, and  $m(\alpha) = 1$  was proved also by Kahane (see Kahane and Peyrière (1976)); the first condition in (iv) above follows from the assumptions that  $m(\alpha) < \infty$  and  $|m'(\alpha)| < \infty$  in the case that  $L$  is bounded since convexity of the function  $x \mapsto x \log x$  shows that  $\mathbf{E}[\langle \alpha, \mathcal{L} \rangle \log^+ \langle \alpha, \mathcal{L} \rangle] \leq |m'(\alpha)| + \|\log^+ L\|_\infty m(\alpha)$ . When the conclusions of Biggins' Theorem hold in Kahane's context, the measure  $\hat{\mu}$  below is introduced by Peyrière on p. 141 of Kahane and Peyrière (1976) for another purpose. It and related constructions in other situations also occur, usually including the same direct construction as ours, in Kallenberg (1977), Hawkes (1981), Rouault (1981), Joffe and Waugh (1982), Kesten (1986), Chauvin and Rouault (1988), Chauvin, Rouault and Wakolbinger (1991), and Waymire and Williams (1996).

Evidently, (iii) implies (ii). The fact that (i) and (ii) are equivalent follows from the standard "zero-one" property of Galton-Watson processes. We shall present a simple proof of the other equivalences modelled on the proof of the Kesten-Stigum theorem in Lyons, Pemantle and Peres (1995). I am grateful to Anatole Joffe for asking me for the details of how this is done. The same method was discovered independently by Waymire and Williams (1996) for the case treated by Kahane (mentioned above). In fact, Waymire and Williams relax the condition that the  $X_i$  be i.i.d. They even relax the independence of the  $\mathcal{L}(\sigma)$ , which could be done here as well.

*Proof.* Fix  $\alpha$ . If  $t$  is a rooted tree (with distinguishable vertices) and  $X$  is a real-valued function on the vertices of  $t$  other than its root, we call  $(t, X)$  a **labelled tree**. A **ray** in a tree is an infinite line of descent starting from the root. Given a ray  $\xi$ , the vertex on  $\xi$  in generation  $n$  is denoted  $\xi_n$ . In the space of labelled trees, let  $\mathcal{F}_n$  denote the  $\sigma$ -field

generated by the first  $n$  levels. We shall also work on the space of labelled trees with distinguished rays,  $(t, X, \xi)$ ; denote by  $\mathcal{F}_n^*$  the  $\sigma$ -field generated by the first  $n$  levels there. For  $\sigma \in t$ , write  $S(\sigma) := \sum_{0 < \tau \leq \sigma} X(\tau)$  and set

$$W_n(t, X) := \frac{\sum_{|\sigma|=n} e^{-\alpha S(\sigma)}}{m(\alpha)^n}.$$

Branching random walk gives a probability measure,  $\mu$ , on the set of labelled trees. We shall construct a related probability measure  $\widehat{\mu}^*$  on the set of infinite labelled trees with distinguished rays. Let  $\mu_n$  be the restriction of  $\mu$  to  $\mathcal{F}_n$ . Now any  $\mathcal{F}_n^*$ -measurable function  $f$  can be written as

$$f(t, X, \xi) = \sum_{|\sigma|=n} f_\sigma(t, X) \mathbf{1}_{\xi_n=\sigma}$$

for some  $\mathcal{F}_n$ -measurable functions  $f_\sigma$ . Let  $\mu_n^*$  be counting measure on  $\{|\sigma| = n\}$  fibered over  $\mu_n$ ; more precisely,  $\mu_n^*$  is the (non-probability) measure on  $\mathcal{F}_n^*$  such that for all nonnegative  $\mathcal{F}_n^*$ -measurable functions  $f$ ,

$$\int f(t, X, \xi) d\mu_n^* = \int \sum_{|\sigma|=n} f_\sigma(t, X) d\mu_n.$$

Then the measure  $\widehat{\mu}^*$  will satisfy

$$(1) \quad \frac{d\widehat{\mu}_n^*}{d\mu_n^*}(t, X, \xi) = \frac{e^{-\alpha S(\xi_n)}}{m(\alpha)^n}$$

for all  $n$  and all labelled trees with rays  $(t, X, \xi)$ , where  $\widehat{\mu}_n^*$  denotes the restriction of  $\widehat{\mu}^*$  to  $\mathcal{F}_n^*$ . The projection of  $\widehat{\mu}^*$  to the space of trees, denoted  $\widehat{\mu}$ , then satisfies

$$(2) \quad \frac{d\widehat{\mu}_n}{d\mu_n}(t, X) = W_n(t, X)$$

for all  $n$  and all labelled trees  $(t, X)$ , where  $\widehat{\mu}_n$  denotes the restriction of  $\widehat{\mu}$  to  $\mathcal{F}_n$ .

To define  $\widehat{\mu}^*$ , let  $\widehat{\mathcal{L}}$  be a random variable whose law has Radon-Nikodym derivative  $\langle \alpha, \mathcal{L} \rangle / m(\alpha)$  with respect to the law of  $\mathcal{L}$ . Start with one particle  $v_0$  at the origin. Generate offspring and displacements according to a copy  $\widehat{\mathcal{L}}_1$  of  $\widehat{\mathcal{L}}$ . Pick one of these children  $v_1$  at random, where a child is picked with probability proportional to  $e^{-\alpha X}$  when its displacement is  $X$ . The children other than  $v_1$  give rise to ordinary independent branching random walks, while  $v_1$  gets an independent number of offspring and displacements according to a copy  $\widehat{\mathcal{L}}_2$  of  $\widehat{\mathcal{L}}$ . Again, pick one of the children of  $v_1$  at random, call it  $v_2$ , with the others giving rise to ordinary independent branching random walks, and so on.

Define the measure  $\widehat{\mu}^*$  as the joint distribution of the random labelled tree and the random ray  $(v_0, v_1, v_2, \dots)$ . Then

$$\frac{d\widehat{\mu}_{n+1}^*}{d\mu_{n+1}^*}(t, X, \xi) = \frac{d\widehat{\mu}_n^*}{d\mu_n^*}(t, X, \xi) \cdot \frac{\langle \alpha, \mathcal{L}(\xi_n) \rangle}{m(\alpha)} \cdot \frac{e^{-\alpha X(\xi_{n+1})}}{\langle \alpha, \mathcal{L}(\xi_n) \rangle} = \frac{e^{-\alpha X(\xi_{n+1})}}{m(\alpha)} \frac{d\widehat{\mu}_n^*}{d\mu_n^*}(t, X, \xi).$$

Thus, (1) follows by induction.

Note that for any  $k \geq 0$ ,

$$(3) \quad \int X(v_k) d\widehat{\mu}^* = \mathbf{E} \left[ \sum_{i=1}^L X_i \frac{e^{-\alpha X_i} \langle \alpha, \mathcal{L} \rangle}{\langle \alpha, \mathcal{L} \rangle m(\alpha)} \right] = -m'(\alpha)/m(\alpha).$$

Thus, by the strong law of large numbers, we have

$$(4) \quad S(v_n)/n \rightarrow -m'(\alpha)/m(\alpha) \quad \widehat{\mu}^* \text{-a.s.}$$

For any labelled tree  $(t, X)$ , set  $W(t, X) := \limsup W_n(t, X)$ . Now by (2), we have the implications (Durrett (1991), p. 210, Exercise 3.6)

$$(5) \quad W(t, X) = \infty \quad \widehat{\mu} \text{-a.s.} \quad \implies \quad W(t, X) = 0 \quad \mu \text{-a.s.},$$

$$(6) \quad W(t, X) < \infty \quad \widehat{\mu} \text{-a.s.} \quad \implies \quad \int W(t, X) d\mu = 1.$$

Suppose first that (iv) fails. We have

$$(7) \quad W_{n+1}(t, X) \geq \frac{e^{-\alpha S(v_n)}}{m(\alpha)^{n+1}} \langle \alpha, \widehat{\mathcal{L}}_{n+1} \rangle,$$

with the two terms in the product being  $\widehat{\mu}^*$ -independent. Now, if  $\alpha m'(\alpha)/m(\alpha) \geq \log m(\alpha)$ , then  $\limsup e^{-\alpha S(v_n)}/m(\alpha)^n = \infty$  by (4) in case  $\alpha m'(\alpha)/m(\alpha) > \log m(\alpha)$  and by (3) and the Chung-Fuchs theorem in case  $\alpha m'(\alpha)/m(\alpha) = \log m(\alpha)$ . This implies by (7) that  $W(t, X) = \infty$   $\widehat{\mu}$ -a.s., whence by (5), (ii) fails. On the other hand, if  $\alpha m'(\alpha)/m(\alpha) < \log m(\alpha)$ , then since  $\mathbf{E}[\log^+ \langle \alpha, \widehat{\mathcal{L}} \rangle] = \mathbf{E}[\langle \alpha, \mathcal{L} \rangle \log^+ \langle \alpha, \mathcal{L} \rangle]/m(\alpha) = \infty$  by assumption, we have that

$$\limsup \frac{1}{n} \log^+ \langle \alpha, \widehat{\mathcal{L}}_n \rangle = \infty \quad \widehat{\mu}^* \text{-a.s.}$$

by virtue of the Borel-Cantelli lemma. This means that the first term in the right-hand side of (7) decays exponentially while the second has superexponential explosions. Hence, again,  $W(t, X) = \infty$   $\widehat{\mu}$ -a.s.

Conversely, suppose that (iv) holds. Let  $\mathcal{G}$  be the  $\sigma$ -field generated by  $\{\widehat{\mathcal{L}}_k\}_{k \geq 1}$ . Then

$$\begin{aligned} \mathbf{E}_{\widehat{\mu}^*}[W_n(t, X) \mid \mathcal{G}] &= \sum_{k=0}^{n-1} \frac{e^{-\alpha S(v_k)}}{m(\alpha)^{k+1}} \left( \langle \alpha, \widehat{\mathcal{L}}_{k+1} \rangle - e^{-\alpha X(v_{k+1})} \right) + \frac{e^{-\alpha S(v_n)}}{m(\alpha)^n} \\ &= \sum_{k=0}^{n-1} \frac{e^{-\alpha S(v_k)}}{m(\alpha)^{k+1}} \langle \alpha, \widehat{\mathcal{L}}_{k+1} \rangle - \sum_{k=1}^{n-1} \frac{e^{-\alpha S(v_k)}}{m(\alpha)^k}. \end{aligned}$$

By hypothesis and (4), the terms  $e^{-\alpha S(v_k)}/m(\alpha)^k$  decay exponentially while the terms  $\langle \alpha, \widehat{\mathcal{L}}_{k+1} \rangle$  grow (at most) subexponentially by the Borel-Cantelli lemma again. Therefore both series converge  $\widehat{\mu}^*$ -a.s., whence  $\liminf W_n(t, X) < \infty$   $\widehat{\mu}$ -a.s. by Fatou's lemma. In light of (2),  $\{1/W_n(t, X)\}$  is a  $\widehat{\mu}$ -supermartingale, so that  $\{W_n(t, X)\}$  converges  $\widehat{\mu}$ -a.s. Thus, we have  $W(t, X) < \infty$   $\widehat{\mu}$ -a.s. and (iii) is a consequence of (6). ■

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