# A branching random walk on $\mathbb{R}$ in random environment 

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1. Introduction and description of the model

Branching random walk (BRW) on $\mathbb{R}($ or $\mathbb{Z})$ in random environment:
a system of particles on $\mathbb{R}$ (or $\mathbb{Z}$ ), where each particle reproduces new ones moving on $\mathbb{R}$ (or $\mathbb{Z}$ ); the offspring distribution and the moving law of a particle depend on an environment in time and/or in locations. Different models have been studied according to the dependence on the environment in time and/or in locations.

Our model: Branching random walk on $\mathbb{R}$ with random environment in time:
the offspring distribution of a particle of generation $\boldsymbol{n}$, and the distributions of the displacements of their children depend on an environment $\xi_{n}$, indexed by the time $n$, supposed to be stationary and ergodic.

The difference with the classical BRW: the distributions are realizations of a stochastic process, rather than fixed ones.

Description of the model:

## Random environment (in time):

modeled as a sequence of stationary and ergodic random variables with values in some measurable space $(\Theta, F)$.

Each value of $\xi_{n}$ corresponds to point distribution $\mu_{n}=\mu\left(\xi_{n}\right)$ on $\mathbb{R}$.

At generation 0 , there is one initial particle located at $S_{\emptyset}=0 \in \mathbb{R}$.
Each particle $\boldsymbol{u}$ of gen. $\boldsymbol{n}$ located at $S_{u}$ is replaced by $\boldsymbol{N}_{\boldsymbol{u}}$ new particles $u i$ of gen. $n+1$, located at

$$
S_{u i}=S_{u}+L_{u i} \quad\left(1 \leq i \leq N_{u}\right) ;
$$

the point process formulated by the the number of offspring and their displacements, ( $N_{u} ; L_{u 1}, \ldots, L_{u N_{u}}$ ), is of distribution $\mu_{n}=\mu\left(\xi_{n}\right)$ (given the environment $\xi$ ).

All particles behave independently: given $\xi,\left(N_{u} ; L_{u 1}, \ldots, L_{u N_{u}}\right)$ are independent of each other.

Closely related topics: Classical Branching Random Walks: Biggins (1977, 1978), R. Lyons (1997), ...
Branching processes: Harris (1963), Smith, W.L. and Wilkinson, W. (1969), Athreya and Karlin (1971), ...

Mandelbrot's multiplicative cascades: J. P. Kahane and J. Peyriére (1976), Y. Guivarc'h (1990), Q. Liu (2000), ... ;

Infinite particle systems: Durrett and Liggett (1983), ...
Quicksort algorithms: Rösler (1993), ...
Random fractals: Mauldin and Williams (1986), Falconer (1986), Q. Liu (1993) ...
Weighted branching processes: D. Kuhlbusch (2004), ...
Other topics: see Q. Liu (1998), and D. J. ALDOUS AND A. BANDYOPADHYAY (2005).

## Quenched and annealed laws:

Let $\left(\boldsymbol{\Gamma}, \boldsymbol{P}_{\boldsymbol{\xi}}\right)$ be the probability space under which the process is defined when the environment $\boldsymbol{\xi}$ is fixed. As usual, $\boldsymbol{P}_{\boldsymbol{\xi}}$ is called quenched law.

The total probability space can be formulated as the product space $\left(\Theta^{\mathbb{N}} \times \Gamma, P\right)$, where $\boldsymbol{P}=\boldsymbol{P}_{\boldsymbol{\xi}} \otimes \boldsymbol{\tau}$ in the sense that for all measurable and positive $g$ we have

$$
\int g d P=\iint g(\xi, y) d P_{\xi}(y) d \tau(\xi)
$$

where $\boldsymbol{\tau}$ is the law of the environment $\boldsymbol{\xi} . \boldsymbol{P}$ is called annealed law. $\boldsymbol{P}_{\boldsymbol{\xi}}$ may be considered to be the conditional probability of $\boldsymbol{P}$ given $\boldsymbol{\xi}$.

## Related models: random walks on $\mathbb{Z}$ in random environment:

(1) Many authors considered the case where the offspring distribution of a particle situated at $z \in \mathbb{Z}$ depends on a random environment indexed by $z$, while the moving mechanism is controlled by a fixed deterministic law. See for example Greven and der Hollander (1992), Baillon, Clement, Greven and den Hollander (1992), Fleischmann and Greven (1992), and Révśz (1998).
(2) Devulder (2005) considered the case where the offspring distribution is fixed, while the moving law of a particle situated at $z$ depends on a random environment indexed by $\boldsymbol{z}$.
(3) Comets, Menshikov and Popop (1998), and Machado and Popov (2000) considered a branching random walk on $\mathbb{N}$ rather then $\mathbb{Z}$, where both offspring distributions and the moving laws depend on a random environment $\left(\omega_{z}\right)$ indexed by locations.
(4) Xu Li, Yingqiu Li and Quansheng Liu (2007) considered a branching random walk on $\mathbb{Z}$ where the offspring distribution of a particle of generation $n$ and located at $z \in \mathbb{Z}$ depend on the environment in time $\xi_{n}$, while the moving laws of its children depend on an environment in locations, $\omega_{z}$.

Main problems that we study:

Recall that $\boldsymbol{S}_{\boldsymbol{u}}$ denotes the position of a particle $\boldsymbol{u}$. For $\boldsymbol{n} \in \mathbb{N}$, let $\boldsymbol{T}_{\boldsymbol{n}}$ be the set of particles of generation $\boldsymbol{n}$, and let

$$
Z_{n}=\sum_{u \in T_{n}} \delta_{S_{u}}
$$

be the counting measure of particles of gen. $n$, so that for $A \subset \mathbb{R}$,
$Z_{n}(A)=$ number of particles of gen. n located in $A$.
Let

$$
L_{n}=\min _{u \in T_{n}} S_{u} \quad\left(\text { resp. } R_{n}=\max _{u \in T_{n}} S_{u}\right)
$$

be the position of leftmost (resp. rightmost) particles of gen. $\boldsymbol{n}$.

We consider the case where the corresponding branching process in random environment, $\left(Z_{n}(\mathbb{R})\right)$, is supercritical, and we want to know asymptotic properties of

$$
Z_{n}(A), \quad L_{n} \quad \text { and } R_{n}
$$

2. The mean of $Z_{n}(n A)$

For $n \in \mathbb{N}$ and $t \in \mathbb{R}$, let

$$
\begin{equation*}
m_{n}(t)=E_{\xi} \sum_{i=1}^{N_{u}} e^{t L_{u i}} \tag{1}
\end{equation*}
$$

where $\boldsymbol{u}$ is a sequence of length $\boldsymbol{n}$; by convention, $\emptyset$ is of length $\mathbf{0}$, and $\emptyset i=i$. Notice that the expectation does not depend on the choice of $\boldsymbol{u}$. Then

$$
m_{0}(t)=E_{\xi} \sum_{i=1}^{N} e^{t L_{i}}, \quad m_{0}(0)=E_{\xi} N
$$

Throughout the paper we will always assume that

$$
\begin{equation*}
E \log m_{0}(0)>0 \text { and } E\left|\log m_{0}(t)\right|<\infty \tag{2}
\end{equation*}
$$

for all $t \in \mathbb{R}$, so that the corresponding $\operatorname{BPRE} Z_{\boldsymbol{n}}(\mathbb{R})$ is supercritical:

$$
Z_{n}(\mathbb{R}) \rightarrow \infty
$$

with positive probability. For simplicity, assume also that a.s.

$$
N \geq 1
$$

Let

$$
\Lambda(t):=E \log m_{0}(t)
$$

and let

$$
\begin{equation*}
\Lambda^{*}(x)=\sup _{t \in \mathbb{R}}\{x t-\Lambda(t)\} \tag{3}
\end{equation*}
$$

be its Legendre transform. Then $\Lambda$ is differentiable on $\mathbb{R}$,

$$
\Lambda^{*}(x)= \begin{cases}t \Lambda^{\prime}(t)-\Lambda(t) & \text { if } x=\Lambda^{\prime}(t) \text { for some } t \in \mathbb{R} \\ +\infty & \text { if } x>\Lambda^{\prime}(+\infty) \text { or } x<\Lambda^{\prime}(-\infty)\end{cases}
$$

and

$$
\min _{x} \Lambda^{*}(x)=\Lambda^{*}\left(\Lambda^{\prime}(0)\right)=-\Lambda(t)=-E \log m_{0}(0)<0
$$

Theorem 1 (Large Deviation Principle for quenched mean). For almost every $\xi$, the sequence of finite measures $A \mapsto E_{\xi} Z_{n}(\boldsymbol{n} A)$ satisfies a principle of large deviation with rate function $\Lambda^{*}$ : for each measurable subset $\boldsymbol{A}$ of $\mathbb{R}$,

$$
\begin{aligned}
-\inf _{x \in A^{o}} \Lambda^{*}(x) & \leq \liminf _{n \rightarrow \infty} \frac{1}{n} \log E_{\xi} Z_{n}(n A) \\
& \leq \limsup _{n \rightarrow \infty} \frac{1}{n} \log E_{\xi} Z_{n}(n A) \leq-\inf _{x \in \bar{A}} \Lambda^{*}(x),
\end{aligned}
$$

where $\boldsymbol{A}^{o}$ denotes the interior of $\boldsymbol{A}$, and $\bar{A}$ its closure.

Proof. Let $q_{n}(A)=E_{\xi} Z_{n}(A)$. Show that

$$
\begin{equation*}
\tilde{q_{n}}(t):=\int e^{t x} d q_{n}(x)=E_{\xi} \sum_{u \in \mathbb{T}_{n}} e^{t S_{u}}=m_{0}(t) \ldots m_{n-1}(t), \tag{5}
\end{equation*}
$$

$$
\begin{equation*}
\lim _{n \rightarrow \infty} \frac{1}{n} \log \tilde{q_{n}}(t)=\Lambda(t):=E \log m_{0}(t) \text { a.s. } \tag{6}
\end{equation*}
$$

and apply Gärtner-Ellis' theorem to the sequence of normalized probability measures $q_{n}(n A) / q_{n}(\mathbb{R})$.

Let $\Lambda_{a}(t)=\log E m_{0}(t)$ and $\Lambda_{a}^{*}$ be its Legendre transform. Similarly, we have:

Theorem 2 (Large Deviation Principle for annealed mean). Assume that $\xi_{n}$ are iid. Then the sequence of finite measures $A \mapsto E Z_{n}(\boldsymbol{n} \boldsymbol{A})$ satisfies a principle of large deviation with rate function $\Lambda_{a}^{*}$ : for each measurable subset $\boldsymbol{A}$ of $\mathbb{R}$,

$$
\begin{aligned}
-\inf _{x \in A^{o}} \Lambda_{a}^{*}(x) & \leq \liminf _{n \rightarrow \infty} \frac{1}{n} \log E Z_{n}(n A) \\
& \leq \limsup _{n \rightarrow \infty} \frac{1}{n} \log E Z_{n}(n A) \leq-\inf _{x \in \bar{A}} \Lambda_{a}^{*}(x)
\end{aligned}
$$

where $\boldsymbol{A}^{o}$ denotes the interior of $\boldsymbol{A}$, and $\overline{\boldsymbol{A}}$ its closure.

Remark:

$$
\Lambda_{a}(t) \geq \Lambda(t) \quad \text { and } \quad \Lambda_{a}^{*}(x) \leq \Lambda^{*}(x)
$$

so that the growth rate of the annealed mean $\boldsymbol{E} Z_{\boldsymbol{n}}[a n, \infty)$ is greater than that of the quenched mean $\boldsymbol{E}_{\xi} Z_{\boldsymbol{n}}[a n, \infty)$.
3. Convergence of the free energy

Let

$$
\tilde{Z}_{n}(t):=\int e^{t x} d Z_{n}(x)=\sum_{u \in T_{n}} e^{t S_{u}}
$$

be the Laplace transform of $Z_{n}$, also called the partition function. We are interested to the convergence of the free energy $\frac{\log \tilde{Z}_{n}(t)}{n}$, and the asymptotic properties of $Z_{n}(n A)$.

## Definition of $t_{-}$and $t_{+}$

Let

$$
\rho(t)=t \Lambda^{\prime}(t)-\lambda(t), \quad t \in \mathbb{R}
$$

Then

$$
\rho^{\prime}(t)=t \Lambda^{\prime \prime}(t)
$$

$\rho(t)$ decreases on $\mathbb{R}_{-}$, increases on $\mathbb{R}_{+}$, and attains its minimum at 0 :

$$
\min _{t} \rho(t)=\rho(0)=-\lambda(0)<0
$$

Let

$$
\begin{aligned}
& t_{-}=\inf \left\{t \in \mathbb{R}: t \Lambda^{\prime}(t)-\lambda(t) \leq 0\right\} \\
& t_{+}=\sup \left\{t \in \mathbb{R}: t \Lambda^{\prime}(t)-\lambda(t) \leq 0\right\}
\end{aligned}
$$

Then $-\infty \leq t_{-}<0<t_{+} \leq \infty, t_{-}$and $t_{+}$are two solutions of $t \Lambda^{\prime}(t)-\lambda(t)=0$ if they are finite.

Theorem 3 (Convergence of the free energy). A.s.

$$
\lim _{n \rightarrow \infty} \frac{\log \tilde{Z}_{n}(t)}{n}=\tilde{\Lambda}(t):= \begin{cases}\Lambda(t) & \text { if } t \in\left(t_{-}, t_{+}\right) \\ t \Lambda^{\prime}\left(t_{+}\right) & \text {if } t \geq t_{+} \\ t \Lambda^{\prime}\left(t_{-}\right) & \text {if } t \leq t_{-}\end{cases}
$$

Deterministic environment case: B. Chauvin and A. Rouault (1997), J. Franchi (1993).
4. Almost sure asymptotic properties of $Z_{n}(n A)$ :

Let $\tilde{\Lambda}^{*}(x)$ be the Legendre transform of $\tilde{\Lambda}(t)$. By Theorem 3 and Gärtner- Ellis' theorem, we obtain:

Theorem 4 (LDP for $Z_{n}(n A)$ ). A.s. the sequence of finite measures $A \mapsto Z_{n}(n A)$ satisfies a principle of large deviation with rate function $\tilde{\Lambda}^{*}$ : for each measurable subset $\boldsymbol{A}$ of $\mathbb{R}$,

$$
\begin{aligned}
-\inf _{x \in A^{o}} \tilde{\Lambda}^{*}(x) & \leq \liminf _{n \rightarrow \infty} \frac{1}{n} \log Z_{n}(n A) \\
& \leq \limsup _{n \rightarrow \infty} \frac{1}{n} \log Z_{n}(n A) \leq-\inf _{x \in \bar{A}} \tilde{\Lambda}^{*}(x)
\end{aligned}
$$

where $\boldsymbol{A}^{\boldsymbol{o}}$ denotes the interior of $\boldsymbol{A}$, and $\overline{\boldsymbol{A}}$ its closure.

Corollary A.s.

$$
\begin{gathered}
\lim _{n} \frac{1}{n} \log Z_{n}[n x, \infty)=-\Lambda^{*}(x)>0 \text { if } x \in\left(\Lambda^{\prime}(0), \Lambda^{\prime}\left(t_{+}\right),\right. \\
\lim _{n} \frac{1}{n} \log Z_{n}(-\infty, n x]=-\Lambda^{*}(x)>0 \text { if } x \in\left(\Lambda^{\prime}\left(t_{-}\right), \Lambda^{\prime}(0)\right) .
\end{gathered}
$$

Remark.

$$
\begin{aligned}
& x \in\left(\Lambda^{\prime}(0), \Lambda^{\prime}\left(t_{+}\right) \text {iff } x>\Lambda^{\prime}(0) \text { and } \Lambda^{*}(x)<0 .\right. \\
& x \in\left(\Lambda^{\prime}\left(t_{-}\right), \Lambda^{\prime}(0) \text { iff } x<\Lambda^{\prime}(0) \text { and } \Lambda^{*}(x)<0 .\right.
\end{aligned}
$$

For deterministic branching random walk: see Biggins (1977).
5. Positions of the rightmost and leftmost particles of gen. $\boldsymbol{n}$

Theorem 5 (Asymptotic properties of $L_{n}$ and $\boldsymbol{R}_{n}$ ) It is a.s. that

$$
\begin{aligned}
& \lim _{n} \frac{L_{n}}{n}=\Lambda^{\prime}\left(t_{-}\right) \\
& \lim _{n} \frac{R_{n}}{n}=\Lambda^{\prime}\left(t_{+}\right)
\end{aligned}
$$

For deterministic branching random walk: see Biggins (1977).

## 6 Proof of theorems 3 and 5

Observation:

$$
W_{n}(t):=\frac{\tilde{Z}_{n}(t)}{E_{\xi} \tilde{Z}_{n}(t)}=\frac{\sum_{u \in T_{n}} e^{t S_{u}}}{m_{0}(t) \ldots m_{n-1}(t)}
$$

is a martingale, therefore converges a.s. to a r.v. $\boldsymbol{W}(t) \in[0, \infty)$.

Remark: In the constant environment case, this martingale has been studied by J. P. Kahane - J. Peyrière (1976), J. Biggins (1977), DurrettLiggett (1983), Y. Guivarc'h (1990), R. Lyons (1997) and Q. Liu (1997, 1998, 2000, 2001), etc. in different contexts.

Lemma 1 If $t \in\left(\boldsymbol{t}_{-}, \boldsymbol{t}_{+}\right)$and $\boldsymbol{E} \boldsymbol{W}_{\mathbf{1}}(\boldsymbol{t}) \log ^{+} \boldsymbol{W}_{\mathbf{1}}(\boldsymbol{t})<\infty$, then

$$
W(t)>0 \quad \text { a.s. }
$$

If $t \leq t_{-}$or $t \geq t_{+}$), then

$$
W(t)=0 \quad \text { a.s. }
$$

Proof. Use a result by Dirk Kuhlbusch (2004) on weighted branching processes in random environment.

## Lemma 2 A.s.

$$
\limsup _{n} \frac{R_{n}}{n} \leq \Lambda^{\prime}\left(t_{+}\right)
$$

Proof. For $a>\Lambda^{\prime}\left(t_{+}\right)$, use asymptotic properties of $\boldsymbol{E}_{\xi} Z_{n}[a n, \infty)$ to show that

$$
\sum_{n} P_{\xi}\left(Z_{n}[a n, \infty) \geq 1\right)<\infty
$$

and apply Borel-Cantelli's lemma to conclude that $\boldsymbol{P}_{\boldsymbol{\xi}}$ a.s.

$$
Z_{n}[a n, \infty)=0 \quad \text { for } n \text { large enough. }
$$

Lemma 3 If $\boldsymbol{t} \geq \boldsymbol{t}_{+}$, then

$$
\limsup _{n} \frac{\log \tilde{Z}_{n}(t)}{n} \leq t \Lambda^{\prime}\left(t_{+}\right)
$$

## Thank you!

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