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Einstein relation for biased random walk on Galton–Watson trees

Gerard Ben Arous∗ Yueyun Hu† Stefano Olla‡ Ofer Zeitouni§


Abstract

We prove the Einstein relation, relating the velocity under a small perturbation to the diffusivity in equilibrium, for certain biased random walks on Galton–Watson trees. This provides the first example where the Einstein relation is proved for motion in random media with arbitrary deep traps.

1 Introduction

Let ω be a rooted Galton–Watson tree with offspring distribution \{p_k\}, where \( p_0 = 0, \) \( m = \sum k p_k > 1 \) and \( \sum b^k p_k < \infty \) for some \( b > 1 \). For a vertex \( v \in \omega \), let \(|v|\) denote the distance of \( v \) from the root of \( \omega \). Consider a (continuous–time) nearest-neighbor random walk \( \{Y_{t}^{\alpha}\}_{t \geq 0} \) on \( \omega \), which when at a vertex \( v \), jumps with rate 1 toward each child of \( v \) and at rate \( \lambda = \lambda_{\alpha} = me^{-\alpha} \), \( \alpha \in \mathbb{R} \), toward the parent of \( v \).

It follows from [14] that if \( \alpha = 0 \), the random walk \( \{Y_{t}^{0}\}_{t \geq 0} \) is, for almost every tree \( \omega \), null recurrent (positive recurrent for \( \alpha < 0 \), transient for \( \alpha > 0 \)). Further, an easy adaptation of [18] shows that \( |Y_{nt}^{0}|/\sqrt{n} \) satisfies a (quenched, and hence also annealed) invariance principle (i.e., converges weakly to a multiple of the absolute value of a Brownian motion), with diffusivity

\[ D_0 = \frac{2m^2(m-1)}{\sum k^2 p_k - m}. \]

(Compare with [18, Corollary 1], and note that the factor 2 is due to the speed up of the continuous–time walk relative to the discrete–time walk considered there. See (2.10) below and also the derivation in [4].) On the other hand, see [16], when \( \alpha > 0 \), \( |Y_{t}^{\alpha}|/t \to_{t \to \infty} v_{\alpha} > 0 \), almost surely, with \( v_{\alpha} \) deterministic. A consequence of our main result, Theorem 1.2 below, is the following.

∗Courant institute, New York University, 251 Mercer St., New York, NY 10012, U.S.A. Email: be- narous@cims.nyu.edu.
†Département de Mathématiques, LAGA, Université Paris 13, 99 Av. J-B Clément, 93430 Villetaneuse, FRANCE. Email: yueyun@math.univ-paris13.fr
‡CEREMADE, Place du Maréchal de Lattre de TASSIGNY, F-75775 Paris Cedex 16, FRANCE. and INRIA, Projet MICMAC, Ecole des Ponts, 6 & 8 Av. Pascal, 77455 Marne-la-Vallée Cedex 2, France. Email: olla@ceremade.dauphine.fr . The work of this author was partially supported by the European Advanced Grant Macroscopic Laws and Dynamical Systems (MALADY) (ERC AdG 246953) and by grant ANR-2010-BLAN-0108 (SHEPI).
§School of Mathematics, University of Minnesota, 206 Church St. SE, Minneapolis, MN 55455, USA and Faculty of Mathematics, Weizmann Institute, POB 26, Rehovot 76100, Israel. Email: zeitouni@math.umn.edu. The work of this author was partially supported by NSF grant DMS-0804133 and by a grant from the Israel Science Foundation.
Theorem 1.1 (Einstein relation) With notation and assumptions as above,

(1.2) \[
\lim_{\alpha \downarrow 0} \frac{\tilde{\alpha}_n}{\alpha} = \mathcal{P}^0. 
\]

The relation (1.2) is known as an Einstein relation. It is straightforward to verify that for homogeneous random walks on \( \mathbb{Z}^d \) (corresponding to deterministic Galton–Watson trees, that is, those with \( p_k = 1 \) for some \( k \geq 1 \)), the Einstein relation holds.

In a weak limit (velocity rescaled with time) the Einstein relation is proved in a very general setup by Lebowitz and Rost (cf. [11]). See also [3] for general fluctuation-dissipation relations.

For the tagged particle in the symmetric exclusion process, the Einstein relation has been proved by Loulakis in \( d \geq 3 \) [12]. The approach of [12], based on perturbation theory and transient estimates, was adapted for bond diffusion in \( \mathbb{Z}^d \) for special environment distributions (cf. [9]). For mixing dynamical random environments with spectral gap, a full perturbation expansion can also be proved (cf. [10]).

For a diffusion in random potential, the recent [8] proves the Einstein relation by following the strategy of [11], adding to it a good control (uniform in the environment) of suitably defined regeneration times in the transient regime. A major difference in our setup is the possibility of having “traps” of arbitrary strength in the environment; in particular, the presence of such traps does not allow one to obtain estimates on regeneration times that are uniform in the environment, and we have been unable to obtain sharp enough estimates on regeneration times that would allow us to mimic the strategy in [8]. On the other hand, the tree structure allows us to develop some estimates directly for hitting times via recursions, see Section 3. We emphasize that our work is (to the best of our knowledge) the first in which an Einstein relation is rigorously proved for motion in random environments with arbitrary strong traps.

In order to explore the full range of parameters \( \alpha \), we will work in a more general context than that described above, following [18]. This is described next.

Consider infinite trees \( \mathcal{T} \) with no leaves, equipped with one (semi)-infinite directed path, denoted \( \text{Ray} \), starting from a distinguished vertex called the root and denoted \( o \). We call such a tree a marked tree. Using \( \text{Ray} \), we define in a natural way the offsprings of a vertex \( v \in \mathcal{T} \), and denote by \( D_n(v) \) the collection of vertices that are descendants of \( v \) at distance \( n \) from \( v \), with \( Z_n(v) = |D_n(v)| \). See [18, Section 4] for precise definitions. For any vertex \( v \in \mathcal{T} \), we let \( d_v \) denote the number of offspring of \( v \), and write \( \tilde{v} \) for the parent of \( v \). Finally, we write \( \rho(v) \) for the horocycle distance of \( v \) from the root \( o \). Note that \( \rho(v) \) is positive if \( v \) is a descendant of \( o \) and negative if it is an ancestor of \( o \).

Let \( \Omega_\mathcal{T} \) denote the space of marked trees. As in [18] and motivated by [16], given an offspring distribution \( \{p_k\}_{k \geq 0} \) satisfying our general assumptions, we introduce a reference probability \( IGW \) on \( \Omega_\mathcal{T} \), as follows. Fix the root \( o \) and a semi-infinite ray, denoted \( \text{Ray} \), emanating from it. Each vertex \( v \in \text{Ray} \) with \( v \neq o \) is assigned independently a size-biased number of offspring, that is \( P_{IGW}(d_v = k) = kp_v/m \), one of which is identified with the descendant of \( v \) on \( \text{Ray} \). To each offspring of \( v \neq o \) not on \( \text{Ray} \), and to \( o \), one attaches an independent Galton-Watson tree of offspring distribution \( \{p_k\}_{k \geq 0} \). Note that \( IGW \) makes the collection \( \{d_v\}_{v \in \mathcal{T}} \) independent. We denote expectations with respect to \( IGW \) by \( \langle \cdot \rangle_0 \) (the reason for the notation will become apparent in Section 2.1 below).

As mentioned above and in contrast with [16] and [18], it will be convenient to work in continuous time, because it slightly simplifies the formulas (the adaptation needed to transfer the results to the discrete time setup of [16] are straightforward). For background, we refer to [4], where the results in [16] and [18] are transferred to continuous time, in the more general setup of multi-type Galton–Watson trees. Given a marked tree \( \mathcal{T} \) and \( \alpha \in \mathbb{R} \), we define an \( \alpha \)-biased random walk \( \{X^\alpha_t\}_{t \geq 0} \) on \( \mathcal{T} \) as the continuous time Markov process with state space the vertices of \( \mathcal{T} \), \( X^\alpha_0 = o \), and so that when at \( v \), the jump rate is 1 toward each of the descendants.
of $v$, and the jump rate is $e^{-\alpha m}$ toward the parent $\bar{v}$. More explicitly, the generator of the random walk $\{X^n_t\}_{t \geq 0}$ can be written as

\[
L_\alpha,TF(v) = \sum_{x \in D_1(v)} (F(x) - F(v)) + e^{-\alpha m} \left( F(\bar{v}) - F(v) \right).
\]

Alternatively,

\[
L_\alpha,TF(v) = -m\partial^*_+ F(v) + (e^{-\alpha} - 1)m\partial_- F(v)
\]

where $\partial_- F(v) = F(\bar{v}) - F(v)$ and

\[
\partial^*_+ F(v) = \frac{1}{m} \sum_{x \in D_1(v)} F(x) - F(v)
\]

Note that if $\alpha < 0$ the (average) drift is towards the ancestors, whereas if $\alpha > 0$ the (average) drift is towards the children. As in [16] and [18], we have that

\[
\lim_{t \to \infty} \rho(X^n_t)_{t \to \infty} v_\alpha, \quad \text{IGW - a.s.}
\]

It is easy to verify that when $\alpha > 0$, then $v_\alpha = \bar{v}_\alpha$, and that $\text{sign}(v_\alpha) = \text{sign}(\alpha)$. Further, we have, again from [18], that $\rho(X^n_{[nt]})/\sqrt{n}$ satisfies the invariance principle (that is, converges weakly to a Brownian motion), with diffusivity constant $D^0$ as in (1.1).

Our main result concerning walks on IGW-trees is the following.

**Theorem 1.2** With assumptions as above,

\[
\lim_{\alpha \to 0} \frac{v_\alpha}{\alpha} = \frac{D^0}{2}.
\]

**Remark 1.3** It is natural to expect that the Einstein relation holds in many related models, including Galton–Watson trees with only moment bounds on the offspring distribution, multi-type Galton–Watson trees as in [4], and walks in random environments on Galton–Watson trees, at least in the regime where a CLT with non-zero variance holds, see [5]. We do not explore these extensions here.

The structure of the paper is as follows. In the next section, we consider the case of $\alpha < 0$, exhibit an invariant measure for the environment viewed from the point of view of the particle, and use it to prove the Einstein relation when $\alpha \nearrow 0$. Section 3 deals with the harder case of $\alpha \searrow 0$. We first prove an easier Einstein relation (or linear response) concerning escape probabilities of the walk, exploiting the tree structure to introduce certain recursions. Using that, we relate the Einstein relation for velocities to estimates on hitting times. A crucial role in obtaining these estimates, and an alternative formula for the velocity (Theorem 3.7), is obtained by the introduction, after [6], of a spine random walk, see Lemma 3.3.

## 2 The environment process, and proof of Theorem 1.2 for $\alpha \nearrow 0$.

As is often the case when motion in random media is concerned, it is advantageous to consider the evolution, in $\Omega_T$, of the environment from the point of view of the particle. One of the reasons for our opting to work in continuous time is that when $\alpha = 0$, the invariant measure for that (Markov) process is simply $\text{IGW}$, in contrast with the more complicated measure $\text{IGWR}$ of [18]. We will see that when $\alpha < 0$, an explicit invariant measure for the environment viewed from the point of view of the particle exists, and is absolutely continuous with respect to $\text{IGW}$. 
2.1 The environment process

For a given tree $\mathcal{T}$ and $x \in \mathcal{T}$, let $\tau_x$ denote the shift that moves the root of $\mathcal{T}$ to $x$, with $\text{Ray}$ shifted to start at $x$ in the unique way so that it differs from $\text{Ray}$ before the shift by only finitely many vertices. Then $\tau_x \mathcal{T}$ is rooted at $x$ and has the same (nonoriented) edges as $\mathcal{T}$. (A special role will be played by $\tau_x$ for $x \in D_1(o)$, and by $\tau_x$ with $x = o$. We use $\tau^{-1} \mathcal{T} = \tau_x^{-1} \mathcal{T}$ in the latter case.) The environment process $\{\mathcal{T}_t\}_{t \geq 0}$ is defined by $\mathcal{T}_t = \tau_x \mathcal{T}$. It is straightforward to check that the environment process is a Markov process. In fact, introducing the operators

$$Df(T) = f(\tau^{-1}T) - f(T),$$

we have that the adjoint operator (with respect to $IGW$) is

$$D^* f(T) = \frac{1}{m} \sum_{x \in D_1(o)} f(\tau_x T) - f(T)$$

since

$$\langle gDf \rangle_0 = \langle fD^*g \rangle_0.$$

Notice that $D^*1 = d_o/m - 1$.

Define $W(v,n) = Z_o(v)/m^n$. Then $W(v,n)$ is a positive martingale that converges to a random variable denoted $W_v$. Using the recursions

$$mW_v = \sum_{x \in D_1(v)} W_x, \quad mW(v,n) = \sum_{x \in D_1(v)} W(n-1,x),$$

we see that $\langle W_v \rangle_0 = 1$ for $v \notin \text{Ray}$. To simplify notation, we write $W_{v,j} = W_{v,j}$ with $v_j \in \text{Ray}$ denoting the $j$-th ancestor of $o$. Since $W_o(\tau_x \mathcal{T}) = W_x(\mathcal{T})$, we have that $D^*W_o = 0$.

The generator of the environment process is

$$L_\alpha f(T) = \sum_{x \in D_1(o)} [f(\tau_x T) - f(T)] + e^{-\alpha}m [f(\tau^{-1}T) - f(T)]$$

\begin{equation}
= -mD^*Df(T) + (e^{-\alpha} - 1)mDf(T)
\end{equation}

The adjoint operator (with respect to $IGW$) is $L^*_\alpha = -mD^*D + (e^{-\alpha} - 1)mD^*$. For any $\alpha \in \mathbb{R}$, let $\mu_\alpha$ denote any stationary probability measure for $L_\alpha$, that is $\mu_\alpha$ satisfies, for any bounded measurable $f$,

$$\langle L_\alpha f \rangle_{\mu_\alpha} = 0,$$

where $\langle g \rangle_{\mu} = \int gd\mu$.

Note that $IGW$ is stationary and reversible for $L_0$. Further, it is ergodic for the environment process. This is elementary to prove, since for any bounded function $f(T)$ such that $L_0f = 0$, we have that $\langle |Df|^2 \rangle_0 = 0$, i.e. $f$ is translation invariant for a.e. $\mathcal{T}$ with respect to $IGW$, i.e. constant a.e. Thus, necessarily, $\mu_0 = IGW$, justifying our notation $\langle \cdot \rangle_0 = \langle \cdot \rangle_{IGW}$.

In our setup, due to the existence of regeneration times for $\alpha \neq 0$ with bounded expectation, a general ergodic argument ensures the existence of a stationary measure $\mu_\alpha$, which however may fail in general to be absolutely continuous with respect to $IGW$, see [16]. Further, because $IGW$ is ergodic and the random walk is elliptic, there is at most one $\mu_\alpha$ which is absolutely continuous with respect to $IGW$, since under any such $\mu_\alpha$, the process $\mathcal{T}_t^\alpha$ must be ergodic, see e.g. [19, Corollary 2.1.25] for a similar argument. As we now show, when $\alpha < 0$, this stationary measure $\mu_\alpha$ with density with respect to $IGW$ can be constructed explicitly.
Lemma 2.1 For $\alpha < 0$, the probability measure $\mu_\alpha = \psi_\alpha \mu_0$ where

$$\psi_\alpha(T) = C_\alpha^{-1} Z_\alpha,$$

$$Z_\alpha = \sum_{j=0}^{\infty} e^{j\alpha} W_{-j}(T),$$

$$C_\alpha = \frac{(1-b)e^\alpha m^{-1}}{1-e^\alpha} + \frac{b e^\alpha}{1-e^\alpha} + 1,$$

is stationary for $L_\alpha$. Furthermore

$$\lim_{\alpha \rightarrow 0} \psi_\alpha(T) = 1 \mu_0 \text{ a.e.}$$

Proof of Lemma 2.1: We show first that $C_\alpha$ provides the correct normalization. In fact, from the relation

$$W_{-j} = m^{-1} W_{-j+1} + m^{-1} \sum_{s \in D_1(v_{-j}) \neq \text{Ray}} W_s =: m^{-1}(W_{-j+1} + L_j),$$

and since $\langle W_s \rangle_0 = 1$ if $s \notin \text{Ray}$, we obtain

$$\langle W_{-j} \rangle_0 = m^{-1}\langle W_{-j+1} \rangle_0 + m^{-1} b(m-1), \ j \geq 1.$$

Since $\langle W_0 \rangle_0 = 1$, we deduce that

$$\langle W_{-j} \rangle_0 = (1-b)m^{-j} + b, \ j \geq 0.$$

Thus,

$$\langle \sum_{j=0}^{\infty} e^{j\alpha} W_{-j} \rangle_0 = 1 + b \sum_{j=1}^{\infty} e^{j\alpha} + (1-b) \sum_{j=1}^{\infty} e^{j\alpha} m^{-j}$$

$$= 1 + \frac{b e^\alpha}{1-e^\alpha} + \frac{(1-b)e^\alpha m^{-1}}{1-e^\alpha m^{-1}}$$

$$= \frac{b}{1-e^\alpha} + \frac{1-b}{1-e^\alpha m^{-1}} = C_\alpha,$$

as needed.

Note that the terms $L_j$ appearing in the right side of (2.5) are i.i.d. Substituting and iterating, we get

$$W_{-k} = \frac{W_0}{m^k} + \frac{L_1}{m^{k-1}} + \frac{L_2}{m^{k-2}} + \cdots + \frac{L_k}{m}.$$

Therefore,

$$\left(1 - \frac{e^\alpha}{m}\right) Z_\alpha = W_0 + \frac{1}{m} \sum_{j=1}^{\infty} e^{\alpha j} L_j =: W_0 + M_\alpha.$$

Note that $M_\alpha$ is a weighted sum of i.i.d. random variables. Further, because $\langle |D_1(v_{-j})| \rangle_0 = \sum k^2 p_k / m$ and $\langle W_s \rangle_0 = 1$, we have that $\lim_{\alpha \rightarrow 0} \alpha \langle M_\alpha \rangle_0 = (\sum k^2 p_k / m^2) := \bar{C}$, and that $\text{Var}_{\text{IGW}}(M_\alpha) = O(1/m)$. It then follows (by an interpolation argument) that that

$$\lim_{\alpha \rightarrow 0} \alpha \langle M_\alpha \rangle = \bar{C}, \ \text{IGW-a.s.}$$
Substituting in (2.8), this yields
\[
\lim_{\alpha \to 0} \alpha Z_\alpha = b \quad \text{IGW - a.s.}
\]
and (2.4) follows.

We next verify that \( L^*_\alpha \psi_\alpha = 0 \). Since \( W_j(\tau^{-1} T) = W_{j-1}(T) \), we have
\[
D \psi_\alpha = C^{-1}_\alpha \sum_{j=0}^\infty e^{j\alpha} (W_{j-1}(T) - W_j(T))
\]
\[
= C^{-1}_\alpha \left( \sum_{j=1}^\infty e^{(j-1)\alpha} W_{j-1}(T) - \sum_{j=0}^\infty e^{j\alpha} W_j(T) \right)
\]
\[
= C^{-1}_\alpha \sum_{j=0}^\infty (e^{(j-1)\alpha} - e^{j\alpha}) W_{j-1}(T) - C^{-1}_\alpha e^{-\alpha} W_o
\]
\[
= (e^{-\alpha} - 1) C^{-1}_\alpha \sum_{j=0}^\infty e^{j\alpha} W_{j-1}(T) - C^{-1}_\alpha e^{-\alpha} W_o
\]
\[
= (e^{-\alpha} - 1) \psi_\alpha - C^{-1}_\alpha e^{-\alpha} W_o.
\]
Since \( D^* W_o = 0 \), we have
\[
D^* (D \psi_\alpha - (e^{-\alpha} - 1) \psi_\alpha) = 0,
\]
i.e.
\[
L^*_\alpha \psi_\alpha = 0.
\]

We can now provide the proof of Theorem 1.2 in case \( \alpha \not\to 0 \).

**Proof of Theorem 1.2 when \( \alpha \not\to 0 \):** We begin with the computation of \( v_\alpha \). Because \( \mu_\alpha \) is ergodic and absolutely continuous with respect to IGW, we have that \( v_\alpha \) equals the average drift (under \( \mu_\alpha \)) at 0, that is
\[
v_\alpha = m(\frac{d_\alpha}{m} - e^{-\alpha})_\alpha = m(D^* \psi_\alpha)_0 - m(e^{-\alpha} - 1)
\]
\[
= m(D \psi_\alpha)_0 - m(e^{-\alpha} - 1)
\]
\[
= m((e^{-\alpha} - 1) \psi_\alpha - C^{-1}_\alpha e^{-\alpha} W_o)_0 - m(e^{-\alpha} - 1) = -mC^{-1}_\alpha e^{-\alpha}.
\]

Thus,
\[
\lim_{\alpha \not\to 0} \frac{v_\alpha}{|\alpha|} = -\frac{m^2(m-1)}{\sum k^2 p_k - m}.
\]

It remains to compute the diffusivity \( D^0 \) when \( \alpha = 0 \). Toward this end, one simply repeats the computation in [18, Corollary 1]. One obtains that the diffusivity is
\[
D^0 = \frac{\langle m W_o^2 + \sum_{s \in D_1} W_s^2 \rangle_0}{\langle W_o^2 \rangle_0^2}.
\]
From the definitions we have that \( \langle W_o^2 \rangle_0 = (\sum k^2 p_k - m)/m(m-1) \) (see [18, (2)]), and thus
\[
D^0 = \frac{2m^2(m-1)}{\sum k^2 p_k - m}.
\]
Together with (2.9), this completes the proof of Theorem 1.2 when \( \alpha \not\to 0 \).
Remark 2.2 Note that the construction above fails for $\alpha > 0$, because then $Z_\alpha$ is not defined. The case $\alpha = \infty$ is however special. In that case, the generator is

$$L_\infty f(T) = \sum_{x \in D_\alpha} [f(\tau_x T) - f(T)].$$

In particular, one can verify that the measure defined by $d\mu_\infty / d\mu_{GW} = 1/(Cd_\alpha)$ with $C = \sum_k k^{-1}p_k$ and $\mu_{GW}$ the ordinary Galton–Watson measure $GW$ (defined as $IGW$ but with the standard Galton–Watson measure also for vertices on Ray), is a stationary measure, and that $v_\infty = \langle d_\alpha \rangle_\infty = C$. It follows that the natural invariant measure is not absolutely continuous with respect to $IGW$.

Remark 2.3 For $\alpha < 0$ one can construct other invariant measures, that of course are singular with respect to $IGW$. A particular family of such measures is absolutely continuous with respect to the ordinary Galton–Watson measure $GW$. Indeed, one can verify that the positive function

$$\psi(T) = \frac{C^\infty}{\sum_{j=1}^\infty \prod_{i=1}^{j-1} d_i^{-\alpha}} \sum_{j=1}^\infty (me^{-\alpha})^{-j+1} \prod_{i=1}^{j-1} d_i^{-1},$$

with $C = 1 - e^\alpha$, satisfies $\int \psi dGW = 1$ and $\psi dGW$ is an invariant measure for $L_\alpha$. One can also check that the Einstein relation (1.2) is not satisfied under this measure, emphasizing the role that the measure $IGW$ plays in our setup.

3 Drift towards descendants: proof of Theorem 1.2 for $\alpha \searrow 0$

In the case $\alpha > 0$ we cannot find an explicit expression for the stationary measure so we have to proceed in a different way. We first prove another form of the Einstein relation in terms of the escape probabilities (probability of never returning to the origin).

Because we consider the case $\alpha > 0$, there is no difference between considering the walk under the Galton–Watson tree or under $IGW$ – the limiting velocity is the same, i.e. $v_\alpha = \bar{v}_\alpha$. Thus, we only consider the walk $\{Y_\alpha_t\}_{t \geq 0}$ below.

Our approach is to provide an alternative formula for the speed $v_\alpha$, see Theorem 3.7 below, which is valid for all $\alpha > 0$ small enough. In doing so, we will take advantage of certain recursions, and of the spine random walk associated with the walk on the Galton–Watson tree, see Lemma 3.3.

We recall our standing assumptions: $p_0 = 0$, $m > 1$, and $\sum b^k p_k < \infty$ for some $b > 1$. We will throughout drop the superscript $\alpha$ from the notation when it is clear from the context, writing e.g. $Y_t$ for $Y_t^\alpha$. To introduce our recursions, define $T(x) := \inf\{t \geq 0 : Y_t = x\}$ and $\tau_n := \inf\{t \geq 0 : |Y_t| = n\}$. For a given tree $\omega$, we write $P_{x,\omega}$ for the law of $Y_t$ with $Y_0 = x$. For $0 < |x| \leq n$, define

$$\beta(x) := P_{x,\omega}(T(x) > \tau_n), \quad \beta_n(x) := P_{x,\omega}(T(x) = \tau_n),$$

$$\gamma_n(x) := E_{x,\omega}(\tau_n \wedge T(x)).$$
We study the recursions for $\beta_n$ and $\gamma_n$. By the Markov property of $P_{x,\omega}$, for $|x| < n$,

$$
\gamma_n(x) = \frac{1}{d_x + \lambda} + \sum_{i=1}^{d_x} \frac{1}{\lambda + d_x} E_{x,i,\omega} \left( \tau_n \wedge T(\tau_x) \right)
$$

$$
= \frac{1}{d_x + \lambda} + \sum_{i=1}^{d_x} \frac{1}{\lambda + d_x} \left( E_{x,i,\omega} \left( \tau_n \wedge T(x) \right) + P_{x,i,\omega}(T(x) < \tau_n) E_{x,\omega} \left( \tau_n \wedge T(\tau_x) \right) \right),
$$

which implies that

$$
\gamma_n(x) = \frac{1}{d_x + \lambda} + \sum_{i=1}^{d_x} \frac{1}{\lambda + d_x} \left( \gamma_n(x_i) + (1 - \beta_n(x_i)) \gamma_n(x) \right).
$$

Hence for any $0 < |x| < n$, 

$$
\gamma_n(x) = \frac{1 + \sum_{i=1}^{d_x} \gamma_n(x_i)}{\lambda + \sum_{i=1}^{d_x} \beta_n(x_i)},
$$

with boundary condition $\gamma_n(x) = 0$ for any $|x| = n$. We take the above equality as the definition of $\gamma_n(o)$. Similarly, we have

$$
\beta_n(x) = \frac{\sum_{i=1}^{d_x} \beta_n(x_i)}{\lambda + \sum_{i=1}^{d_x} \beta_n(x_i)}, \quad 0 < |x| < n,
$$

with $\beta_n(x) = 1$ if $|x| = n$, and we define $\beta_n(o)$ so that the above equality holds for $x = o$. Finally, we let $\beta(o) = \lim_{n \to \infty} \beta_n(o)$ (the limit of the monotone sequence $\beta_n(o)$).

**Proposition 3.1** As $\alpha \searrow 0$, $\alpha^{-1} \beta(o)$ converges in law and in expectation to a random variable $Y$ such that

$$
E(Y) = \frac{m(m-1)}{2m^3} = \frac{D^0}{2m}
$$

This is a form of Einstein relation, as linear response for the escape probability. The law of $Y$ can be identified, see the end of the proof of Proposition 3.1.

**Proof:** We clearly have that with $B(x) := \frac{1}{\lambda} \sum_{i=1}^{d_x} \beta_n(x_i)$, it holds that

$$
\beta(x) = \frac{B(x)}{1 + B(x)}, \quad \forall x \neq o,
$$

and

$$
B(x) = \frac{1}{\lambda} \sum_{i=1}^{d_x} \frac{B(x_i)}{1 + B(x_i)}, \quad \forall x \in T.
$$

Notice that all $B(x)$ are distributed as some random variable, say $B$, and conditionally on $d_x$ and on the tree up to generation $|x|$, the variables $B(x_i), 1 \leq i \leq d_x$ are i.i.d. and distributed as $B$. It follows that

$$
E(B) = e^\alpha E \frac{B}{1 + B},
$$

and

$$
E(B^2) = \frac{1}{\lambda^2} \left( mE \left( \frac{B}{1 + B} \right)^2 + E(d_o(d_o - 1)) \left( E \left( \frac{B}{1 + B} \right) \right)^2 \right).
$$
For any nonnegative r.v. $Z \in L^2$, let us denote by $\{Z\} := \frac{Z}{E(Z)}$. By concavity, the following inequality holds (see e.g. [17], Lemma 6.4):

$$\mathbb{E}\left(\frac{1}{1+Z}^2\right) \leq \mathbb{E}\left(\{Z\}^2\right).$$

By (3.3) and (3.4), we get that

$$\mathbb{E}\left(\{B\}^2\right) = \frac{1}{m} \mathbb{E}\left(\left(\frac{B}{1+B}\right)^2\right) + \frac{\mathbb{E}(d(o)(d_o - 1))}{m^2} \leq \frac{1}{m} \mathbb{E}\left(\{B\}^2\right) + \frac{\mathbb{E}(d(o)(d_o - 1))}{m^2},$$

which yields that the second moment of $B$ is uniformly bounded by the square of $\mathbb{E}(B)$: for any $0 < \alpha$,

$$\mathbb{E}(B^2) \leq \frac{\mathbb{E}(d(o)(d_o - 1))}{m - 1} (\mathbb{E}(B))^2.$$

By (3.3), $e^{-\alpha}\mathbb{E}(B) = \mathbb{E}(B) - \mathbb{E}(\frac{B^2}{1+B})$, hence $(1-e^{-\alpha})\mathbb{E}(B) = \mathbb{E}(\frac{B^2}{1+B}) \leq \mathbb{E}(B^2) \leq \frac{\mathbb{E}(d(o)(d_o - 1))}{m^2 - 1} (\mathbb{E}(B))^2$.

It follows that

$$\mathbb{E}(B) \geq \frac{m(m-1)}{\mathbb{E}(d(o)(d_o - 1))} (1-e^{-\alpha}).$$

On the other hand, by Jensen’s inequality, $\mathbb{E}(B) = e^\alpha \mathbb{E}\frac{B}{1+e^\alpha B} \leq e^\alpha \frac{\mathbb{E}(B)}{1+\mathbb{E}(B)}$, which implies that

$$\mathbb{E}(B) \leq (e^\alpha - 1).$$

Therefore, $B/\alpha$ is tight as $\alpha \searrow 0$. In particular, for some sub-sequence $\alpha \searrow 0$, $B(x)/\alpha$ converges in law to some $Y(x)$. Since $B/\alpha$ is bounded in $L^2$ uniformly in $\alpha > 0$ in a neighborhood of 0, we deduce from (3.2) that

$$Y := \frac{1}{m} \sum_{i=1}^{N} Y_i,$$

where $N$ is distributed like $d_o$ and, conditionally on $N$, $(Y_i)$ are i.i.d and distributed as $Y$; moreover $\mathbb{E}(Y) = \lim_{\alpha \searrow 0} \mathbb{E}(\frac{B}{m}) > 0$ (the limit along the same sub-sequence).

Dividing

$$(1-e^{-\alpha})\mathbb{E}(B) = \mathbb{E}(\frac{B^2}{1+B}) \leq \mathbb{E}(B^2)$$

by $\alpha^2$, we get that $\mathbb{E}(Y) = \mathbb{E}(Y^2)$. The same operation in (3.4) gives

$$\mathbb{E}(Y)^2 = \frac{m(m-1)}{\mathbb{E}(d(o)(d_o - 1))} \mathbb{E}(Y^2).$$

Putting these together we obtain $E(Y) = \frac{\alpha_o}{2\alpha_o}.$

On the other hand, it is known (see e.g. [1, Theorem 16]) that the law of $Y$ satisfying (3.5) is determined up to a multiplicative constant, and therefore $Y$ equals in distribution $aW_o$ for some constant $a$. The equality $\mathbb{E}Y = \mathbb{E}Y^2$ then implies that $Y$ equals in distribution $W_o/E(W_o)$. Since all possible limits in law are the same, we get that $\beta_o/\alpha$ converges in law to $W_o/E(W_o)$.

$\square$

We return to the proof of the Einstein relation concerning velocities. Recall that a level regeneration time is a time for which the random walk hits a fresh level and never backtracks, see e.g. [2] for the definition and basic properties. (Level regeneration times are related to, but
Different from the regeneration times introduced in [16], the differences of adjacent regeneration times form an i.i.d. sequence, with all moments bounded. Since $\gamma_n(x)$ is smaller than the $n$-th level regeneration time (started at $x$), it follows that the sequence $\gamma_n(x)/n$ is uniformly integrable (under the measure $GW \times P_{x,\omega}$), and therefore, the convergence in the forthcoming (3.8) holds also in expectation:

$$\lim_{n \to \infty} \frac{\mathbb{E}[\gamma_n(o)]}{n} = \frac{\mathbb{E}(\beta(o))}{v_0}.$$  

Since $\gamma_n(x) = E_{x,\omega} \left( \tau_n 1_{(\tau_n < T(x))} \right) + O(1)$ and $\frac{\lambda}{n} \to \frac{1}{v_0}$, $P_{x,\omega}$ a.s. and in $L^1$ (the latter follows at once from the integrability of regeneration times mentioned above, as $\tau_n$ is bounded above by the $n$th regeneration time), we get that for $x$ fixed,

$$\frac{\gamma_n(x)}{n} \to_{n \to \infty} \frac{1}{v_0} P_{x,\omega} \left( T(x) = \infty \right) = \frac{\beta(x)}{v_0}, \quad GW \text{ a.s.} \quad (3.8)$$

So all we need to prove in order to have the Einstein relation for velocities, is that

$$\lim_{\alpha \to 0} \lim_{n \to \infty} \frac{1}{n} \mathbb{E}(\gamma_n(o)) = \frac{1}{m}.$$  

To this end, define

$$B_n(x) := \frac{1}{\lambda} \sum_{i=1}^{d_x} \beta_n(x_i), \quad \Gamma_n(x) := \sum_{i=1}^{d_x} \gamma_n(x_i), \quad |x| < n.$$

Note that showing (3.9) is equivalent to proving that

$$\lim_{\alpha \to 0} \lim_{n \to \infty} \frac{1}{n} \mathbb{E}(\Gamma_n(o)) = 1.$$  

For $|x| < n - 1$,

$$B_n(x) = \frac{1}{\lambda} \sum_{i=1}^{d_x} \frac{B_n(x_i)}{1 + B_n(x_i)}, \quad \Gamma_n(x) = \frac{1}{\lambda} \sum_{i=1}^{d_x} \frac{1 + \Gamma_n(x_i)}{1 + B_n(x_i)}.$$  

Notice that we could define $\Gamma(x) := \lim_{n \to \infty} \frac{\Gamma_n(x)}{n}$, such that

$$\Gamma(x) := \frac{1}{\lambda} \sum_{i=1}^{d_x} \frac{\Gamma_n(x_i)}{1 + B(x_i)}.$$  

As $\alpha \to 0$ we can show that $\Gamma(x) \to aY(x)$ for some constant $a > 0$. The problem is that in the limit as $n \to \infty$ we lose information on the value of $a$ (that should be $\mathbb{E}(d_\alpha(d_\alpha - 1))/[m(m-1)]$).

In order to determine this constant we have to make a step back and iterate the equations (3.11) and, noticing that $\Gamma_n(x) = 0$ for all $|x| = n - 1$, we get that

$$\Gamma_n(o) = \sum_{r=1}^{n-1} \frac{1}{\lambda^r} \sum_{|u_r| = r} \frac{1}{1 + B_n(u_1)} \cdots \frac{1}{1 + B_n(u_{r-1})} 1 + B_n(u_r) := \sum_{r=1}^{n-1} \Phi_n(r),$$

where $\{u_0, ..., u_r\}$ is the shortest path relating the root $o$ to $u$ if $|u_0 = o, |u_1| = 1, ..., |u_r| = r$. Note that $B_n(u_1), ..., B_n(u_r)$ are correlated.

Observe that $\Phi_n(r) \leq e^{\alpha r} W(o, r)$, consequently $\mathbb{E}(\Phi_n(r)) \leq e^{\alpha r}$. Since $\alpha > 0$ it is hard to control the limit of $\Gamma_n(o)$. The aim is to analyze the asymptotic behavior of $\mathbb{E}(\Phi_n(r))$ as $n \to \infty$ and $r \leq n$, which will be done in the following two subsections: in the next first subsection we will give a useful representation of $\mathbb{E}(\Phi_n(r))$ based on a spine random walk, whereas in the second subsection we make use of an argument from renewal theory and establish the limit of $\mathbb{E}(\Phi_n(r))$ when $r, n \to \infty$ in an appropriate way.
3.1 Spine random walk representation of $\mathbb{E}(\Phi_n(r))$

Let $\Omega$ denote the space of rooted trees with no leaves. Denote by $\Omega_T$ the space of trees with a marked infinite ray $\text{Ray} = (u^*_n)_{n \geq 0}$, with $u^*_0 = o$ [$\Omega_T$ is topologically equal to $\Omega_T$]. Unlike the setup used in Section 2, where e.g. $u^*_1$ was considered a parent of $o$, we now redefine the notion of descendant in $\Omega_T$. Namely, for $x \in T$, $x \neq o$, the parent of $x$, denoted $\overleftarrow{x}$, is the unique vertex on the geodesic connecting $x$ and $o$ with $|\overleftarrow{x}| = |x| - 1$. In this section, for any $|v| < n$, we define the normalized progeny of $v$ at level $n$ as $M_n(v) := |\{w : |w| = n, w \text{ descendant of } v\}|/m^{n-|v|}$, and $M_n(v) = 1$ if $|v| = n$. We also write $M_n = M_n(o)$.

According to [13], on the space $\Omega_T$ we may construct a probability $Q$ such that the marginal of $Q$ on the space of trees $\Omega$ satisfies
\[
\frac{dQ}{dP}|_{F_n} := M_n, \quad n \geq 1,
\]
where $F_n$ is the $\sigma$-field generated by the first $n$ generations of $\omega$ and $P$ denotes the Galton–Watson law. Due to $p_0 = 0$ and our tail assumptions, we have that $M_{\infty} > 0$ a.s. and moreover,
\[
Q\left(u^*_n = u | F_n\right) = \frac{1}{M_n m^n}, \quad \forall |u| = n.
\]

Under $Q$, $d_{u^*_n}$ has the size-biased distribution associated with $\{p_k\}$, that is $Q(d_{u^*_n} = k) = kp_k/m$, $u^*_{n+1}$ is uniformly chosen among the children of $u^*_n$ and, for $v \neq u^*_{n+1}$ with $\overleftarrow{v} = u^*_n$, the sub-trees $T(v)$ rooted at $v$, are i.i.d. and have a Galton–Watson law.

For any $0 \leq j < n$, we define $a^{(n)}_j := \frac{1}{k} \sum_{v \neq u^*_{j+1}, \overleftarrow{v} = u^*_j} \beta_n(v)$. Note that under $Q$, the family $\{a^{(n)}_j\}_{0 \leq j < n}$ is independent and each $a^{(n)}_j$ is distributed as $\frac{1}{k} \sum_{k=1}^{d^*_j} \beta^{(k)}_{n-j-1}$, where $d^*$ has the size-biased distribution associated with $\{p_k\}$, $(\beta^{(k)}_l, l \geq 1)$ are i.i.d copies of $(\beta_1(o), l \geq 1)$ and independent of $d^*$. We extend $a^{(n)}_j$ to all $j \in \mathbb{Z} \cap (-\infty, n-1]$ by letting the family $\{a^{(n)}_j, n > j \}_{j \in \mathbb{Z}}$ be independent (under $Q$) and such that for each $j$, $\{a^{(n)}_j, n > j \}$ is distributed as $\{\frac{1}{k} \sum_{k=1}^{d^*_j} \beta^{(k)}_{n-j-1}, n > j\}$. We naturally define $a^{(\infty)}_j$ as the limit of $a^{(n)}_j$ as $n \to \infty$. In particular for $j \geq 0$, $a^{(\infty)}_j := \frac{1}{k} \sum_{v \neq u^*_{j+1}, \overleftarrow{v} = u^*_j} \beta_j(v)$, and each $a^{(\infty)}_j$ is distributed as $\frac{1}{k} \sum_{k=1}^{d^*_j} \beta^{(k)}_j$ with $(\beta^{(k)}_j)_{k \geq 1}$ i.i.d copies of $\beta_j(o)$, independent of $d^*$.

The main result of this subsection is the following representation for $\mathbb{E}(\Phi_n(r))$:

**Proposition 3.2** We may define a random walk $(S, P)$ on $\mathbb{Z}$, independent of the Galton-Watson tree $\omega$ and of the family $\{a^{(n)}_j\}_{j < n}$, with step distribution $P(S_i - S_{i-1} = 1) = \frac{\nu^2}{\lambda + m^2}$ and $P(S_i - S_{i-1} = -1) = \frac{\nu^2}{\lambda + m^2}$, $\forall i \geq 1$, such that for any $1 \leq r \leq n$,
\[
\mathbb{E}(\Phi_n(r)) = Q\left[\frac{Z_n(r)}{M_{n-r}}\right],
\]

where
\[
Z_n(r) := E_{0,\omega}\left(1_{\{|\tau_o(-r) < \tau_o(n-r)\}} \prod_{i=0}^{\tau_o(-r)-1} f_{n-r}(S_i)\right), \quad 1 \leq r \leq n,
\]
with $E_{0,\omega}$ the expectation with respect to the random walk $S$ starting from 0, and
\[
f_n(x) := \frac{m^2 + \lambda}{m(1 + \lambda + \lambda a^{(n)}_x)}, \quad x < n.
\]
Before entering into the proof of Proposition 3.2, we mention that the random walk \((S, P)\) may find its root in the following lemma:

**Lemma 3.3 (Spine random walk)** Let \(n > k \geq 2\). Let \(b_{j+1} > 0\) and \(a_j \geq 0\) for all \(0 \leq j < n\). Define \((z_j)_{0 \leq j \leq n}\) by \(z_n = 0\) and \(z_j := \frac{1}{1 + a_j + b_{j+1}(1 - z_{j+1})}, \quad 0 \leq j \leq n - 1.\)

Let \((S_m)\) be a Markov chain on \(\{0, 1, \ldots, n\}\) with probability transition \(\tilde{P}(S_m = j + 1|S_{m-1} = j) = \frac{b_{j+1}}{1 + b_{j+1}}, \text{ and } \tilde{P}(S_m = j - 1|S_{m-1} = j) = \frac{1}{1 + b_{j+1}}\), and denote by \(\tilde{P}_r\) the law of the chain \((S_m)\) with \(S_0 = r\). Then, for any \(1 \leq r < n\),

\[
\prod_{j=1}^{r} z_j = E_r \left(1_{\tau_S(0) < \tau_S(n)} \prod_{j=1}^{n-1} \left(1 + \frac{b_{j+1}}{1 + b_{j+1} + a_j}\right)^{L_x^r(\omega)} \right),
\]

with \(\tau_S(x) := \inf\{j \geq 1 : S_j = x\}\) the first hitting time of \(S\) at \(x\) and \(L_x^r := \sum_{i=0}^{n-1} 1_{(S_i = x)}\) is the local time at \(x\).

Lemma 3.3 can be proved exactly as in [6, Appendix], by using (A.3) and the construction of the random walk therein. We omit the details. Thanks to the spine random walk, studying the local time at \(S\) and the renewal theorem.

**Proof of Proposition 3.2:** Observe that \(\beta_n(r) = \frac{\beta_n(x)}{1 + \beta_n(x)}\) and

\[
\Phi_n(r) = \frac{1}{\lambda} \sum_{|u| = r} (1 - \beta_n(u_1)) \cdots (1 - \beta_n(u_r)).
\]

By the change of measure, we have for any \(F \geq 0\),

\[
E \left[ \sum_{|u| = n} F(\beta_n(u_1), d_{u_1}, \ldots, \beta_n(u_n), d_{u_n}) \right] = m^n Q \left[ F(\beta_n(u^*_1), d_{u^*_1}, \ldots, \beta_n(u^*_n), d_{u^*_n}) \right].
\]

It follows that for \(r < n\),

\[
m^r Q \left( (1 - \beta_n(u^*_1)) \cdots (1 - \beta_n(u^*_r)) \frac{1}{M_n(u^*_1)} \right)
\]

\[
= m^{-(n-r)} E \left[ \sum_{|v| = n} (1 - \beta_n(v_1)) \cdots (1 - \beta_n(v_r)) \frac{1}{M_n(v_r)} \right]
\]

\[
= E \left[ \sum_{|v| = r} (1 - \beta_n(v_1)) \cdots (1 - \beta_n(v_r)) \right],
\]

where the term \(m^{-(n-r)} \frac{1}{M_n(v_r)}\) disappears when one takes the sum over \(|v| = n\) by keeping \(v_r = u\). It follows that

\[
E(\Phi_n(r)) = \frac{m^r}{\lambda^r} Q \left( (1 - \beta_n(u^*_1)) \cdots (1 - \beta_n(u^*_r)) \frac{1}{M_n(u^*_1)} \right),
\]

(3.16)
with the notation of (3.18), we get that for any \( r \leq n - 1 \),
\[
\frac{B_n(o)}{\lambda^r} \prod_{|u|=r} \frac{1}{1 + B_n(u_1)} \cdots \frac{1}{1 + B_n(u_{r-1})} \frac{B_n(u_r)}{1 + B_n(u_r)} = \frac{1}{M_n(u_r)}.
\]
Hence,
\[
\mathbb{E}(B_n(o)) = \frac{m^r}{\lambda^r \mathbb{Q}} \left( 1 - \beta_n(u_n^*) \cdots (1 - \beta_n(u_{r-1}^*)) \beta_n(u_r^*) \frac{1}{M_n(u_r^*)} \right).
\]
Note that
\[
\beta_n(u_j^*) = \frac{\beta_n(u_{j+1}^*) + \sum_{v \neq u_{j+1}^*, \bar{v} = u_j^*} \beta_n(v)}{\lambda + \beta_n(u_{j+1}^*) + \sum_{v \neq u_{j+1}^*, \bar{v} = u_j^*} \beta_n(v)} = \frac{\beta_n(u_{j+1}^*) + m a_j^{(n)}}{\lambda + \beta_n(u_{j+1}^*) + m a_j^{(n)}}, \quad \forall j < n,
\]
with \( \beta_n(u_n^*) = 1 \), and
\[
M_n(u_j^*) = \frac{1}{m} \sum_{v \neq u_{j+1}^*, \bar{v} = u_j^*} M_n(v) + \frac{1}{m} M_n(u_{j+1}^*), \quad \forall j < n,
\]
with \( M_n(u_n^*) = 1 \). Under \( \mathbb{Q} \), for such \(|v| = j + 1\), \((\beta_n(v), M_n(v))\) are i.i.d. and distributed as \((\beta_{n-j-1}(o), M_{n-j-1}(o))\) (under \( P \)).

We can represent \( 1 - \beta_n(u_j^*) \) as the probability for a one-dimensional random walk in a random environment (RWRE) with cemetery point, starting from \( j \), to hit \( j - 1 \) before \( n \). In fact, applying Lemma 3.3 to \( a_j = a_j^{(n)} \) and \( b_{j+1} = \frac{1}{\lambda} \), we see that
\[
\prod_{j=1}^r (1 - \beta_n(u_j^*)) = \tilde{E}_{r, \omega} \left( 1_{(\tau_S(0) < \tau_S(n))} \prod_{j=1}^{n-1} \left( \frac{1 + \lambda}{1 + \lambda + \beta_n(u_j^*)} \right)^{L_{r,\omega}^{(n)}(j)} \right)
\]
\[
= \tilde{E}_{r, \omega} \left( 1_{(\tau_S(0) < \tau_S(n))} \prod_{i=0}^{\tau_S(0)-1} \frac{1 + \lambda}{1 + \lambda + \beta_n(u_i^*)} \right),
\]
where \( (S_i)_{i \geq 0} \) is a random walk on \( \mathbb{Z} \) with step distribution \( \tilde{P}(S_i - S_{i-1} = 1) = \frac{1}{1 + \lambda} \) and \( \tilde{P}(S_i - S_{i-1} = -1) = \frac{\lambda}{1 + \lambda} \) for \( i \geq 1 \), and the expectation \( \tilde{E}_{r, \omega} \) is taken with respect to \( (S_m) \) with \( S_0 = r \).

Define the probability \( P \) with
\[
\frac{dP}{dP_{\sigma(S_0, \ldots, S_n)}} = \left( \frac{\lambda}{m} \right)^{S_n - S_0} \left( \frac{m(1 + \lambda)}{m^2 + \lambda} \right)^n, \quad n \geq 0.
\]
Under \( P \), the random walk \( \{S_m\} \) has the properties stated in the statement of Proposition 3.2.
Further,
\[
\frac{m^r}{\lambda^r} \prod_{j=1}^r (1 - \beta_n(u_j^*)) = \tilde{E}_{r, \omega} \left( 1_{(\tau_S(0) < \tau_S(n))} \prod_{i=0}^{\tau_S(0)-1} f_n(S_i) \right) := \tilde{Z}_n(r).
\]
With the notation of \( \tilde{Z}_n(r) \), we get that
\[
\mathbb{E}(\Phi_n(r)) = \mathbb{Q} \left[ \frac{\tilde{Z}_n(r)}{M_n(u_r^*)} \right].
\]
Observe that for any \( r < n \), under \( \mathbb{Q} \), \((f_n(x+r), M_n(u_r^*))_{x \leq n-r} \) has the law as \((f_{n-r}(x), M_{n-r})_{x \leq n-r} \). This invariance by linear shift and (3.18) yield Proposition 3.2.\ \Box

We end this subsection by the following remark:
Remark 3.4 With the same notations as in Proposition 3.2, we have

\begin{align}(3.19) \quad \mathbb{E}(B_n(o)) & \leq \frac{m}{\lambda} \mathbb{Q} \left[ \frac{Z_n(r-1)}{M_{n-\left(\frac{r}{1}\right)}(u^*_1)} \right], \\
(3.20) \quad \mathbb{E}(B_n(o)) & \geq \frac{m}{\lambda} \mathbb{Q} \left[ \frac{Z_n(r-1)}{M_{n-\left(\frac{r}{1}\right)}(u^*_1)} \frac{a^{(n-r+1)}_1}{1 + a^{(n-r+1)}_1} \right]. \end{align}

Proof of Remark 3.4: In the same way which leads to (3.18), we get from (3.17) that

\begin{align}(3.21) \quad \mathbb{E}(B_n(o)) = \frac{m^r}{\lambda} \mathbb{Q} \left( (1 - \beta_n(u^*_1)) \cdots (1 - \beta_n(u^*_{r-1})) \beta_n(u^*_r) \frac{1}{M_n(u^*_r)} \right) \\
= \frac{m^r}{\lambda} \mathbb{Q} \left( \prod_{i=1}^{r-1} (1 - \beta_n(u^*_i)) - \prod_{i=1}^{r} (1 - \beta_n(u^*_i)) \right) \frac{1}{M_n(u^*_r)} \\
= \frac{m}{\lambda} \mathbb{Q} \left( \frac{Z_n(r-1)}{M_n(u^*_r)} \right) = \mathbb{Q} \left( \frac{Z_n(r)}{M_n(u^*_r)} \right), \end{align}

giving the upper bound in (3.19) after a linear shift at \( r - 1 \) for the above term with \( \Phi \). On the other hand,

\[ \beta_n(u^*_r) = \frac{\beta_n(u^*_{r+1}) + \lambda a^{(n)}_r}{\lambda + \beta_n(u^*_{r+1}) + \lambda a^{(n)}_r} \geq \frac{a^{(n)}_r}{1 + a^{(n)}_r}, \]
hence

\begin{align}(3.22) \quad \mathbb{E}(B_n(o)) & \geq \frac{m^r}{\lambda} \mathbb{Q} \left( (1 - \beta_n(u^*_1)) \cdots (1 - \beta_n(u^*_{r-1})) \frac{a^{(n)}_r}{1 + a^{(n)}_r} \frac{1}{M_n(u^*_r)} \right) \\
= \frac{m}{\lambda} \mathbb{Q} \left( \frac{Z_n(r-1)}{M_n(u^*_r)} \frac{a^{(n)}_r}{1 + a^{(n)}_r} \frac{1}{M_n(u^*_r)} \right), \end{align}

yielding the assertions in Remark 3.4 after the shit at \( r - 1 \). □

3.2 An argument based on renewal theory

The main result is Lemma 3.6 which evaluates the limit of \( \mathbb{E}(\Phi_n(r)) \) and in turn gives the velocity representation in Theorem 3.7. The analysis is based on the use of a renewal structure in the representation of Proposition 3.2. Under \( P, (S_n) \) drifts to \(-\infty\). Denote by \( (R_0 := 0) < R_1 < R_2 < \ldots \) the regeneration times for \( (S_n) \), that is \( R_i = \min\{n > R_{i-1} : \{S_j\}_{j=0}^n \cap \{S_j\}_{j=n} = \emptyset\} \). The sequence \( \{S_{j+R_i - R_i}, 0 \leq j \leq R_{i+1} - R_i\}_{i \geq 1} \) is clearly i.i.d and has as common distribution that of \( \{S_j, 0 \leq j \leq R_1\} \) conditioned on \( \{\tau_R(1) = \infty\} \). Further, because

\[ E(S_{i+1} - S_i) = \frac{\lambda - m^2}{\lambda + m^2} \leq -\frac{m^2}{m + 1}, \]

it is straightforward to check that there exists a constant \( \kappa > 0 \), independent of \( \alpha \), so that

\[ E(e^{\kappa R_i}) < \infty, \quad E(e^{\kappa (R_j - R_i)}) < \infty. \]

Define

\[ \zeta_j := \prod_{i=R_{j-1}}^{R_{j-1} - 1} f_\infty(S_i), \quad j \geq 1, \]

\[ 14 \]
where
\[
(3.24) \quad f_\infty(x) := \frac{m^2 + \lambda}{m(1 + \lambda + \lambda a(x))}, \quad x \in \mathbb{Z},
\]
and for \( x \in \mathbb{Z}, a(x)^{\infty} \) are i.i.d. and are distributed as \( \frac{1}{k} \sum_{k=1}^{d^*} \beta^{(k)} \) with \((\beta^{(k)})_{k \geq 1}\) i.i.d copies of \( \beta(o) \), independent of \( d^* \).

An important observation is that under \( Q \otimes P \), \((\zeta_j, j \geq 2)\) are i.i.d. and independent of \( \zeta_1 \).

Define further
\[
(3.25) \quad h(y) := Q \otimes P \left[ \prod_{i=0}^{\tau(y)-1} f_\infty(S_i) 1(\tau(y) \leq R_1) \right], \quad y \geq 1.
\]
We extend the definition of \( h \) to \( \mathbb{Z} \) by letting \( h(y) := 0 \) if \( y \leq 0 \).

**Lemma 3.5** Assume that
\[
(3.26) \quad \sum_{y=1}^{\infty} h(y) < \infty,
\]
\[
(3.27) \quad Q \otimes P \left[ \frac{\zeta_1}{M_\infty(u_1^{\infty})} + \zeta_1 + |S_{R_2} - S_{R_1}| \right] < \infty.
\]

Then,
\[
Q \otimes P[\zeta_2] = 1.
\]
We will see below, see Lemma 3.8, that (3.26) and (3.27) both hold for all \( \alpha > 0 \) small enough.

**Proof:** Almost surely, \( \beta_n(x) \downarrow \beta(x) \). Then for a fixed \( r \), almost surely,
\[
Z_n(r) \rightarrow Z_\infty(r) := E_0, \omega \left( \prod_{i=0}^{\tau(y)-1} f_\infty(S_i) \right),
\]
and \( M_n(u_1^1) \rightarrow M_\infty(u_1^1) \). Applying Fatou’s lemma in the expectation under \( Q \) in (3.20), we get that for any \( r \),
\[
\frac{\lambda}{m} E(B) \geq Q \left[ Z_\infty(r-1) \frac{a_1(\infty)}{M_\infty(u_1^1)} \right] = Q \otimes P \left[ \prod_{i=0}^{\tau(y)-1} f_\infty(S_i) \frac{1}{M_\infty(u_1^1)} \right].
\]

We can not directly let \( r \rightarrow \infty \) inside the above expectation, so we decompose this expectation by the regeneration times \( 0 < R_1 < R_2 < ... \). Write
\[
\zeta_1' := \frac{\zeta_1}{M_\infty(u_1^1)} \frac{a_1(\infty)}{1 + a_1(\infty)}.
\]

Then
\[
\frac{\lambda}{m} E(B) \geq \sum_{k=2}^{\infty} Q \otimes P \left[ 1(\tau_k \leq R_k) \frac{\zeta_1'}{\prod_{i=R_k-1}^{R_k-1} f_\infty(S_i)} \right] = \sum_{k=2}^{\infty} Q \otimes P \left[ 1(\tau_k \leq R_k) \frac{\zeta_1'}{\prod_{i=R_k-1}^{R_k-1} f_\infty(S_i)} \frac{f_\infty(S_k)}{h(r - 1 + S_{R_k})} \right].
\]
by using the Markov property of $S$ at $R_k$. Observe that $(\zeta_j, S_{R_j} - S_{R_{j-1}})_{j \geq 2}$ are i.i.d. under the annealed measure $Q \otimes P$, and are independent of $(\zeta'_1, S_{R_1})$. By replacing $r - 1$ by $r$, we get that for any $r$,

$$\sum_{k=2}^{\infty} Q \otimes P \left[ 1_{(S_{R_k} > r)} \zeta'_1 \prod_{j=2}^{k} \xi_j (r + S_{R_k}) \right] \leq \frac{\lambda}{m} E(B). \leqno{(3.28)}$$

Now, we claim that

$$Q \otimes P [\zeta_2] \leq 1. \leqno{(3.29)}$$

To prove (3.29), we assume that $a := Q \otimes P [\zeta_2] > 1$ and show that it leads to a contradiction with (3.28). Toward this end, define a distribution $U$ on $\mathbb{Z}_+$ by

$$U(x) := \frac{Q \otimes P \left[ 1_{(S_{R_k} - S_{R_{k-1}} = x)} \zeta_2 \right]}{Q \otimes P [\zeta_2]}, \quad x \geq 0.$$  

Then (3.28) becomes

$$\frac{\lambda}{m} E(B) \geq \sum_{k=2}^{\infty} a^{k-1} Q \otimes P \left[ 1_{(S_{R_k} > r)} \zeta'_1 \prod_{x=0}^{r+S_{R_k}} h(r + S_{R_k} - x)U^{\otimes(k-1)}(x) \right]$$

$$\geq a^{l-1} \sum_{k=l}^{\infty} Q \otimes P \left[ 1_{(S_{R_k} > r)} \zeta'_1 \prod_{x=0}^{r+S_{R_k}} h(r + S_{R_k} - x)U^{\otimes(k-1)}(x) \right]$$

definite integral for any $l \geq 2$.

Since

$$\sum_{k=1}^{l-1} Q \otimes P \left[ 1_{(S_{R_k} > r)} \zeta'_1 \prod_{x=0}^{r+S_{R_k}} h(r + S_{R_k} - x)U^{\otimes(k-1)}(x) \right] \rightarrow 0 \text{ as } r \rightarrow \infty \text{ [by the dominated convergence under (3.26)] and the integrability of } \zeta'_1 \leq \frac{1}{M_{\infty}(u_t)} \text{ under (3.27)},$$

we get that for any fixed $l$,

$$a^{l-1} \frac{\lambda}{m} E(B) \geq \sum_{k=1}^{\infty} Q \otimes P \left[ 1_{(S_{R_k} > r)} \zeta'_1 \prod_{x=0}^{r+S_{R_k}} h(r + S_{R_k} - x)U^{\otimes(k-1)}(x) \right] + o(1)$$

$$= \frac{Q \otimes P [\zeta'_1]}{\sum_{x=0}^{\infty} \lambda x} \sum_{x=0}^{\infty} \frac{h(x)}{xU(x)} + o(1), \quad r \rightarrow \infty,$$

by applying the renewal theorem [7, pg. 362], using (3.27). Thus we get some constant $C > 0$ such that $\frac{\lambda}{m} E(B) \geq a^{l-1} C$ for any $l \geq 2$, which is impossible if $a > 1$. Hence we proved (3.29).

It remains to show

$$Q \otimes P [\zeta_2] \geq 1. \leqno{(3.30)}$$

The proof of this part is similar, we shall use (3.19) instead of (3.20). Set

$$\bar{f}_S(r) := \prod_{i=0}^{\tau_S(r) - 1} f_\infty(S_i).$$

Since $f_{\infty}(x) \geq f_\ell(x)$ for any $\ell$, we get that

$$\frac{\lambda}{m} E(B_n(o)) \leq Q \otimes P \left[ 1_{(\tau_S(1-r) < \tau_S(n-(r-1)))} \bar{f}_S(r-1) \frac{1}{M_{n-r+1}(u_t)} \right].$$

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Taking \( r = n \) gives that
\[
\frac{\lambda}{m} \mathbb{E}(B_n(o)) \leq \mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) < \tau_S(1))} \tilde{f}_S(n - 1) \right].
\]
Since \( \mathbb{E}(B) \leq \mathbb{E}(B_n(o)) \), we obtain that for any \( n \),
\[
\frac{\lambda}{m} \mathbb{E}(B) \leq \mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) < \tau_S(1))} \tilde{f}_S(n) \right] + \sum_{k=1}^{\infty} \mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) < \tau_S(1))} \tilde{f}_S(n) \right] + \sum_{k=1}^{\infty} \mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) < \tau_S(1))} \tilde{f}_S(n) \right].
\]
By the Markov property at \( \tau_S(1) \), the first term equals
\[
\frac{\mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) < \tau_S(1))} \tilde{f}_S(n) \right]}{P_{-n}(\tau_S(1) = \infty)} = \frac{P(\tau_S(1) = \infty)}{P_{-n}(\tau_S(1) = \infty)} h(n) \to 0,
\]
since \( P_{-n}(\tau_S(1) = \infty) \geq c \) for some constant \( c > 0 \) and \( h(n) \to 0 \). Then, recalling that \( h \) vanishes at \( \mathbb{Z}_- \), we get
\[
\frac{\lambda}{m} \mathbb{E}(B) \leq o(1) + \sum_{k=1}^{\infty} \mathbb{Q} \otimes P \left[ \zeta_1 \prod_{j=2}^{k} \zeta_j h(n + S_{R_k}) \right],
\]
with \( \zeta_j \) and \( h \) defined as before. If \( a := \mathbb{Q} \otimes P [\zeta_2] < 1 \), then with the distribution \( U(\cdot) \) introduced before,
\[
\sum_{k=1}^{\infty} \mathbb{Q} \otimes P \left[ \zeta_1 \prod_{j=2}^{k} \zeta_j h(n + S_{R_k}) \right] = \sum_{k=1}^{\infty} a^{k-1} \mathbb{Q} \otimes P \left[ \zeta_1 \sum_{x=0}^{n+S_{R_k}} h(n + S_{R_k} - x) U^{\otimes(k-1)}(x) \right]
\approx \sum_{k=1}^{\infty} a^{k-1} b_k^{(n)}.
\]
Note that \( \max_{n} b_k^{(n)} \leq \mathbb{Q} \otimes \tilde{P} [\zeta_1] \sum_{x=0}^{\infty} h(x) U^{\otimes(k-1)}(x) \), and that, due to (3.26), \( \lim_{n \to \infty} b_k^{(n)} = 0 \). The dominated convergence theorem then implies that \( \sum_{k=1}^{\infty} a^{k-1} b_k^{(n)} \to 0 \) which in turn yields \( \frac{\lambda}{m} \mathbb{E}(B) \leq o(1) \), a contradiction. Thus \( \mathbb{Q} \otimes P [\zeta_2] \geq 1 \). This completes the proof of the lemma. \( \square \)

**Lemma 3.6** Assume (3.26), (3.27) and that for some \( p > 1 \),
\[
(3.31) \quad \mathbb{Q} \otimes P((\zeta_2)^p) < \infty.
\]
Furthermore, we assume that
\[
(3.32) \quad \text{under } \mathbb{Q} \otimes P, \text{ the family } \{ \frac{1}{M_n} \}_{k \geq 1} \text{ is uniformly integrable},
\]
and that
\[
(3.33) \quad \lim_{r \to \infty} \sup_{n \geq r} \mathbb{Q} \otimes P \left[ 1_{(\tau_S(1) \leq R_1)} \prod_{i=0}^{\tau_S(1) - 1} f_{\infty}(S_i) \frac{1}{M_{n-r}} \right] = 0,
\]

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where as before, \( R_1 \) is the first regeneration time for \( S \) under \( P \). Then, for any \( \varepsilon > 0 \),

\[
\lim_{n \to \infty} \max_{\varepsilon n \leq y \leq (1-\varepsilon)n} \left| \mathbb{E}(\Phi_n(r)) - \frac{Q \otimes P \left( \frac{\zeta_r}{M_{\infty}} \right) \sum_{y \geq 1} h(y)}{Q \otimes P (\zeta_2 | S_{R_2} - S_{R_1})} \right| = 0.
\]

Moreover,

\[
\sup_{r \geq 1, n \geq r} \mathbb{E}(\Phi_n(r)) < \infty.
\]

**Proof:** We split the proof of (3.34) into the following upper and lower bounds:

\[
\limsup_{r \to \infty, n-r \to \infty} \mathbb{E}(\Phi_n(r)) \leq \frac{Q \otimes P \left( \frac{\zeta_r}{M_{\infty}} \right) \sum_{y \geq 1} h(y)}{Q \otimes P (\zeta_2 | S_{R_2} - S_{R_1})},
\]

\[
\liminf_{n \to \infty} \min_{\varepsilon n \leq y \leq (1-\varepsilon)n} \mathbb{E}(\Phi_n(r)) \geq \frac{Q \otimes P \left( \frac{\zeta_r}{M_{\infty}} \right) \sum_{y \geq 1} h(y)}{Q \otimes P (\zeta_2 | S_{R_2} - S_{R_1})}.
\]

**Proofs of (3.35) and (3.36)**: Let us introduce the notation: for \( \ell \geq 1 \),

\[
\zeta_j(\ell) := \prod_{i=R_{j-1}}^{R_{j-1}} f\ell(S_i), \quad j \geq 1.
\]

By (3.14),

\[
\mathbb{E}(\Phi_n(r)) = Q \otimes P \left[ \prod_{i=0}^{\tau_S(-r) - 1} f_{n-r}(S_i) \right].
\]

Noticing that \( 1_{(\tau_S(-r) < \tau_S(n-r))} = 1_{(\tau_S(-r) \leq R_1 \wedge \tau_S(n-r))} + \sum_{k=1}^{\infty} 1_{(R_1 < \tau_S(n-r), R_k < \tau_S(-r) \leq R_{k+1})} \), we get

\[
\mathbb{E}(\Phi_n(r)) = I_{(3.38)}(0) + \sum_{k=1}^{\infty} I_{(3.38)}(k),
\]

where

\[
I_{(3.38)}(0) := \mathbb{E}(\Phi_n(r)) = \prod_{i=0}^{\tau_S(-r) - 1} f_{n-r}(S_i) \left( \frac{1}{M_{n-r}} \right),
\]

\[
I_{(3.38)}(k) := \mathbb{E}(\Phi_n(r)) = \prod_{i=0}^{\tau_S(-r) - 1} f_{n-r}(S_i) \left( \frac{1}{M_{n-r}} \right),
\]

by the Markov property at \( R_k \) (with convention \( \prod_0^0 = 1 \)) and with

\[
h_i(y) := \mathbb{E}(\Phi_n(r)) \left( \prod_{i=0}^{\tau_S(-y) - 1} f(S_i) \left( \tau_S(-y) \leq R_1 \right) \left( \tau_S(1) = \infty \right) \right), \quad y \geq 1, \ell \geq 1.
\]
We also define $h_L(y) := 0$ for all $y \leq 0$. Since $f_L(x) \leq f_\infty(x)$, $h_L(x) \leq h(x)$, $\zeta_j(n - r) \leq \zeta_j$ for any $j \geq 1$, we have

$$I_{(3.38)}(0) \leq Q \otimes P \left[ \prod_{i=0}^{\tau_n(-r)-1} f_\infty(S_i) \frac{1}{M_{n-r}} \right] \to 0, \quad as \ n > r \to \infty,$$

by (3.33). Moreover, $I_{(3.38)}(0)$ is uniformly bounded over all $n \geq r \geq 1$, again by (3.33). Further,

$$\sum_{k=1}^{\infty} I_{(3.38)}(k) \leq \sum_{k=1}^{\infty} Q \otimes P \left[ \prod_{j=2}^{k} \frac{\zeta_j}{M_{n-r}} h(r + S_{R_1}) \right]$$

$$= \sum_{k=1}^{\infty} Q \otimes P \left[ \prod_{j=2}^{k} \frac{\zeta_j}{M_{n-r}} h(r + S_{R_1} - x) U^{(k-1)}(x) \right]$$

$$= \sum_{k=1}^{\infty} Q \otimes P \left[ \prod_{j=2}^{k} \frac{\zeta_j}{M_{n-r}} h(r + S_{R_1} - x) U(x) \right],$$

(3.39)

where $u(x) := \sum_{y=1}^{\infty} U^{(k-1)}(x), x \geq 0$, and

$$U(x) := Q \otimes P \left[ 1_{(S_{R_2} - S_{R_1}) \geq -x} \right], \quad x \geq 0,$$

is a distribution by Lemma 3.6. By the renewal theorem, $u(x) \to \frac{1}{\sum_{y=1}^{\infty} h(y)} \frac{1}{Q \otimes P (\zeta_2|S_{R_2} - S_{R_1})} := u(\infty), \quad x \to \infty.$

Recall (3.26) that $\sum h(y) < \infty$. It follows that as $r \to \infty$ and $n - r \to \infty$, the term $[\cdots]$ in (3.39) converges almost surely to $\frac{\zeta_j}{M_{n-r}} \sum_{y=1}^{\infty} h(y) u(\infty)$. This in view of the uniform integrability (3.32), yield that (3.39) converges to

$$\frac{Q \otimes P \left( \frac{\zeta_j}{M_{n-r}} \right)}{Q \otimes P (\zeta_2|S_{R_2} - S_{R_1})} \sum_{y=1}^{\infty} h(y) \text{ as } r \to \infty \text{ and } n - r \to \infty.$$

The estimate (3.36) follows. Finally, note that (3.39) is bounded by

$$\max_{x \geq 0} u(x) \sum_{y=1}^{\infty} h(y) Q \otimes P \left[ \frac{\zeta_j}{M_{n-r}} \right] \leq c,$$

for some constant $c > 0$, uniformly over $n \geq r \geq 1$, again by (3.32). Hence $E(\Phi_n(r)) \leq I_{(3.38)}(0) + c$, implying (3.35).

**Proof of (3.37) :** The idea is to replace $\zeta_2(n - r)$ by $\zeta_2(\infty)$ in (3.38). Let $\ell = n - r$ and recall that $f_\ell(x) := \frac{m^2 + \lambda}{m(1 + \lambda + \lambda a_{\ell}^{(s)})} = \frac{m^2 + \lambda}{m(1 + \lambda + \lambda a_{\ell}^{(s)})} = f_\infty(x) \frac{\lambda(a_0^{(l)} - a_0^{(\infty)})}{1 + \lambda + \lambda a_{\ell}^{(s)}}$, for $x < \ell$. Then

$$0 \leq f_\infty(x) - f_\ell(x) = \frac{m^2 + \lambda}{m(1 + \lambda + \lambda a_{\ell}^{(s)})} = f_\infty(x) \frac{\lambda(a_0^{(l)} - a_0^{(\infty)})}{1 + \lambda + \lambda a_{\ell}^{(s)}}$$

It follows that for any $j \geq 2$,

$$\zeta_j(\ell) = \zeta_j \prod_{i=R_1-1}^{R_j-1} \left[ 1 - \frac{\lambda(a_0^{(l)} - a_0^{(\infty)})}{1 + \lambda + \lambda a_{\ell}^{(s)}} \right] := \zeta_j \times A_j(\ell).$$
Fix a large integer $L$. Using (3.38) and the fact that $h(y)$ is nondecreasing for any $y$, we deduce that for all $n - r \geq L$ and any large constant $C > 0$ [the constant $C$ will be chosen later on],

$$\mathbb{E}(\Phi_n(r)) \geq \sum_{k=1}^{C_n} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(L), R_2 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j \prod_{j=2}^{k} \Lambda_j(n-r) h_L(r + S_{R_k}) \right].$$

The first step is to replace $\Lambda_j(n-r)$ by 1, then we have to check that the error term is uniformly small:

(3.40)

$$I_{(3.40)} := \sum_{k=1}^{C_n} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(L), R_2 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j h_L(r + S_{R_k}) \right] \to 0,$$

as $r \to \infty$ and $\varepsilon n \leq r \leq (1 - \varepsilon)n$. Let us postpone for the moment the proof of (3.40). Going back to $\mathbb{E}(\Phi_n(r))$, we obtain that

$$\mathbb{E}(\Phi_n(r)) \geq \sum_{k=1}^{C_n} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(L), R_2 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j h_L(r + S_{R_k}) \right] - I_{(3.40)}$$

(3.41)

$$\geq \sum_{k=1}^{\infty} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(L), R_2 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j h_L(r + S_{R_k}) \right] - I_{(3.40)} - I_{(3.41)},$$

with

$$I_{(3.41)} := \sum_{k=C_n+1}^{\infty} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(L), R_2 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j h_L(r + S_{R_k}) \right].$$

If we can prove that for a well-chosen $C$, $I_{(3.41)}$ goes to zero uniformly for $r \to \infty$ and $\varepsilon n \leq r \leq (1 - \varepsilon)n$, then by applying the renewal theorem ($L$ fixed, $r \to \infty$ and $n - r \to \infty$) to the sum in (3.41), under the uniform integrability (3.32), we get that

$$\liminf_{n \to \infty} \min_{n-r \geq \varepsilon n} \mathbb{E}(\Phi_n(r)) \geq \frac{\mathbb{Q} \otimes P (\zeta_1(L) 1_{\{R_1 < \tau_{\xi}(L)\}} \sum_{y \geq 1} h_L(y))}{\mathbb{Q} \otimes P \left( \zeta_1 |S_{R_2} - S_{R_1}| \right)}.$$

Letting $L \to \infty$ gives the lower bound (3.37).

It remains to show that $I_{(3.41)}$ and $I_{(3.40)}$ go to zero uniformly for $r \to \infty$ and $\varepsilon n \leq r \leq (1 - \varepsilon)n$. We first deal with $I_{(3.41)}$. Let $h^* := \max_{x \geq 0} h(x)$. We have

$$I_{(3.41)} \leq h^* \sum_{k=C_n+1}^{\infty} \mathbb{Q} \otimes P \left[ 1_{\{R_1 < \tau_{\xi}(r)\}} \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j \right]$$

$$\leq h^* \sum_{k=C_n+1}^{\infty} \mathbb{Q} \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} 1_{\{S_{R_k} - S_{R_1} > r\}} \prod_{j=2}^{k} \zeta_j \right]$$

$$= h^* \sum_{k=C_n+1}^{\infty} \mathbb{Q} \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \right] \mathbb{Q} \otimes P \left[ 1_{\{S_{R_k} - S_{R_1} > r\}} \prod_{j=2}^{k} \zeta_j \right].$$
by the independence between \((\zeta_1(L), M_{n-r})\) and \((S_{R_k} - S_{R_i}, \zeta_j, j \geq 2\). Recall that \(Q \otimes P(\zeta_2) = 1\). Let \(\hat{P}\) be a new probability measure defined by \(d\hat{P} = P(\zeta_2, \xi_2)\), then under \(\hat{P}\), \(S_{R_k} - S_{R_i}\) is the sum of \(k-1\) positive i.i.d. variables with mean \(Q \otimes P(\zeta_2(S_{R_k} - S_{R_i})) = \alpha \in (0, \infty)\) by (3.27).

Taking \(C := \frac{2}{\alpha}\). Then by Cramer’s bound, there exists some \(c_0 > 0\) such that

\[
Q \otimes P \left[ \frac{1}{(S_{R_k} - S_{R_i} > r)} \prod_{j=2}^{k} \zeta_j \right] = \hat{P} \left( S_{R_k} - S_{R_i} < r \right) \leq e^{-c_0 k},
\]

for any \(k > Cn\) and \(r \leq n\). It follows

\[
I_{(3.41)} \leq h^* Q \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j \right] \sum_{k=Cn+1}^{\infty} e^{-c_0 k} \to 0,
\]

uniformly as \(r \leq n\) and \(r \rightarrow \infty\), since \(Q \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \right] \leq Q \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \right] \leq C'\), with some constant \(C' > 0\), thanks to the uniform integrability (3.32).

It remains to check (3.40) [with \(C := \frac{2}{\alpha}\) chosen before]. We first observe that \(h_1(x) \leq h(x) \leq h^*\) and that

\[
I_{(3.40)} \leq h^* \sum_{k=1}^{Cn} Q \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \prod_{j=2}^{k} \zeta_j \left( 1 - \prod_{j=2}^{k} \Lambda_j(n-r) \right) \right]
\]

\[
= h^* \sum_{k=1}^{Cn} \hat{E} \left[ \frac{\zeta_1(L)}{M_{n-r}} \left( 1 - \prod_{j=2}^{k} \Lambda_j(n-r) \right) \right]
\]

\[
= h^* \sum_{k=1}^{Cn} \hat{E} \left[ \frac{\zeta_1(L)}{M_{n-r}} \left( 1 - \hat{E}(A_2(n-r)) \right)^k \right],
\]

where the annealed expectation \(\hat{E}\) has the density \(\zeta_2\) with respect to \(Q \otimes P\) and under \(\hat{E}\), \(\Lambda_j\) are i.i.d and independent of \(\zeta_1\). Note that by the independence of \(\zeta_2\) and \((\zeta_1, M_{n-r})\) under \(Q \otimes P\), we have \(\hat{E} \left[ \frac{\zeta_1(L)}{M_{n-r}} \right] = Q \otimes P \left[ \frac{\zeta_1(L)}{M_{n-r}} \right] \leq C'\).

To proceed, we employ the following estimate, which will be proved below: there exists a constant \(c_1\) (that may depend on \(\alpha\)) so that \(\forall \ell \geq \ell_0,\)

\[
1 - \hat{E}(A_2(\ell)) \leq e^{-c_1 \ell}.
\]

Since \(n - r \geq \epsilon n\), (3.42) yields that \(I_{(3.40)} \to 0\) as stated in (3.40).

It remains to check (3.42). \(\beta_2(\alpha) - \beta(\alpha)\) corresponds to the probability that an excursion of the tree-valued walk is higher than \(\ell\); the latter is dominated by the probability that a level regeneration distance is larger than \(\ell\), which decays exponentially by [2, Lemma 4.2(i)]. It follows that

\[
\mathbb{P}(\beta_2(\alpha) - \beta(\alpha) > e^{-c_2 \ell}) \leq e^{-c_2 \ell}, \quad \forall \ell \geq \ell_0,
\]

where \(c_2\) may depend on \(\alpha\). Then \(\mathbb{E}(\beta_2(\alpha) - \beta(\alpha)) \leq 2e^{-c_2 \ell}\). Notice that for \(R_1 \leq i < R_2, S_i < 0\) hence

\[
\sum_{i=R_1}^{R_2-1} \left( a^{(x)}_{S_i} - a^{(\infty)}_{S_i} \right) \sum_{x \leq 0} \left( a^{(x)}_{S_i} - a^{(\infty)}_{S_i} \right) (L^x_{R_2} - L^x_{R_1}) \leq \frac{1}{\lambda} \sum_{x \leq 0} \sum_{k=1}^{d^*_x - 1} (\beta^{(x,k)} - \beta^{(x,k)}) (L^x_{R_2} - L^x_{R_1}),
\]

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implying that
\[ Q \otimes P \left[ \sum_{i=R_1}^{R_2-1} (a_{S_i}^{(f)} - a_{S_i}^{(\infty)}) \right] \leq \frac{2}{\lambda} Q(d^* - 1) E(R_2 - R_1) e^{-c_3 \ell}. \]

By Hölder’s inequality with \( \frac{1}{p} + \frac{1}{q} = 1 \),
\[ \mathbb{E} \left[ \frac{\sum_{i=R_1}^{R_2-1} (a_{S_i}^{(f)} - a_{S_i}^{(\infty)}) \geq e^{-c_3 \ell}}{1} \right] = Q \otimes P \left[ \sum_{i=R_1}^{R_2-1} (a_{S_i}^{(f)} - a_{S_i}^{(\infty)}) \right]^{1/p} \left( Q \otimes P \left( \sum_{i=R_1}^{R_2-1} (a_{S_i}^{(f)} - a_{S_i}^{(\infty)}) > e^{-c_3 \ell} \right) \right)^{1/q} \leq e^{-c_3 \ell}, \]
for some constant \( c_3 = c_3(\alpha, p, q, c_2) > 0 \) and for all large \( \ell \). Now, using the elementary inequality: for any \( j \geq 1 \) and \( x_1, \ldots, x_j \in [0, 1] \), \( 1 - \prod_{i=1}^{j} (1 - x_i) \leq \sum_{i=1}^{j} x_i \), we get
\[ 1 - \Lambda_2(\ell) \leq \sum_{i=R_1}^{R_2-1} \frac{\lambda (a_{S_i}^{(f)} - a_{S_i}^{(\infty)})^{(f)}}{1 + \lambda + \lambda a_{S_i}^{(f)}} \leq \sum_{i=R_1}^{R_2-1} (a_{S_i}^{(f)} - a_{S_i}^{(\infty)}). \]
Therefore
\[ 1 - \mathbb{E}(\Lambda_2(\ell)) \leq 2 e^{-c_3 \ell}, \]
proving (3.42). The proof of the lemma is now complete. \( \square \)

We have the following representation of the velocity \( v_\alpha \).

**Theorem 3.7 (Velocity representation)** Assume (3.26), (3.27), (3.31) and (3.33). Recall the function \( f_\infty \), see (3.24). Then,
\[ v_\alpha = \frac{m}{\lambda} \otimes P \left[ \prod_{i=0}^{R_1-1} f_\infty(S_i) \frac{\beta_n^{(1)}}{M_\infty^{(1)}} \right] \]
\[ \frac{Q \otimes P \left[ \prod_{i=0}^{R_1-1} f_\infty(S_i) \frac{\beta_n^{(1)}}{M_\infty^{(1)}} \right]}{Q \otimes P \left[ \prod_{i=0}^{R_1-1} f_\infty(S_i) \frac{\beta_n^{(1)}}{M_\infty^{(1)}} \right]} \sum_{y \geq 1} h(y). \]

We recall that \( M_\infty = M_\infty(u_0^\ast) \) and \( u_0^\ast = o \).

**Proof:** Noticing that \( \mathbb{E}(\Gamma_n(o)) = m \mathbb{E}(\gamma_n(1,o)) \). By (3.7), (3.13) and Lemma 3.6, we immediately obtain a representation of the velocity \( v_\alpha \):

\[ m \mathbb{E}(\beta) \]
\[ v_\alpha = \frac{m}{\lambda} \otimes P \left[ \prod_{i=0}^{R_1-1} f_\infty(S_i) \frac{\beta_n^{(1)}}{M_\infty^{(1)}} \right] \sum_{y \geq 1} h(y). \]

Going back to (3.17), and recalling that
\[ \frac{m^r}{n^r} \prod_{j=1}^{r} (1 - \beta_n(u_j^\ast)) = E_r \omega \left( 1_{(S_0(0) < \tau_\omega(n))} \prod_{i=0}^{\tau_\omega(n)-1} f_n(S_i) \right), \]
we get that for any \( r \leq n - 1 \),
\[ \mathbb{E}(R_n(o)) = \frac{m}{\lambda} \otimes E_r \omega \left( 1_{(S_0(0) < \tau_\omega(n))} \prod_{i=0}^{\tau_\omega(n)-1} f_n(S_i) \frac{\beta_n(u_j^\ast)}{M_n(u_j^\ast)} \right). \]
By shifting $P_{r-1,ω}$ to $P_{0,ω}$, we have that for any $r ≤ n - 1$,

$$\mathbb{E}(B_n(o)) = \frac{m}{λ} Q \otimes P \left( (τ_{x}(1-r) < τ_{z}(n-r))^T \prod_{i=0}^{τ_{z}(0)-1} f_{n-r+1}(S_i) \frac{β_{n-r+1}(u_{i}^*)}{M_{n-r+1}(u_{i}^*)} \right).$$

Repeating the renewal arguments in Section 3.2 which lead to Lemma 3.6 (the difference with $Φ_n(r)$ only comes from the part before the regeneration time $R_1$), we see that

$$\lim_{n \to \infty} \mathbb{E}(B_n(o)) = \frac{m}{λ} Q \otimes P \left[ \prod_{i=0}^{R_1-1} f_{∞}(S_i) \frac{β_{∞}(u_{i}^*)}{M_{∞}(u_{i}^*)} \right] \sum_{y ≥ 1} h(y).$$

On the other hand, $\lim_{n \to \infty} \mathbb{E}(B_n(o)) = \mathbb{E}(B(o)) = \frac{m}{λ} \mathbb{E}(β(o))$. Comparing this with the velocity representation (3.44), we get the result. □

Before applying Theorem 3.7, we show that the conditions for the representation of $ν_α$ hold when $α$ is small enough. Recall our standing assumption that $p_0 = 0$, and the constant $κ$, see (3.23).

**Lemma 3.8** There exists an $α_0 = α_0(m,κ)$ such that if $0 < α < α_0$ then (3.26), (3.27), (3.31), (3.32) and (3.33) hold.

**Proof:** Note that $f_∞(x) ≤ (m^2 + λ)/(m + mλ)$ and the right side is a bounded differentiable function of $α$, which equals 1 at $α = 0$. It follows that $f_∞(x) ≤ 1 + cα$ for some constant $c = c(m)$.

In what follows, we will make sure to use constants that do not depend on $α$. Note that, since $P(τ_3(1) = ∞)$ is bounded away from 0 uniformly in $α$, $\sum_{y=1}^{∞} h(y) ≤ C \sum_{n=0}^{∞} (1 + cα)^n P(R_1 ≥ n) ≤ C' \sum_{n=0}^{∞} (1 + cα)^n e^{−kn}$,

where $C' = C'(κ, m)$ and we used (3.23). In particular, for $α < α_0(m,κ)$, we deduce (3.26).

The proof of (3.27) is similar: since $|S_{R_1} - S_{R_2}| < μ ≤ μ$, the exponential moments (3.23) imply that it is enough to check that $Q \otimes P[ζ_1] < ∞$ and $Q \otimes P[ζ_1^δ] < ∞$ for some $δ > 0$ independent of $α$. Using again the estimate $f_∞(x) ≤ 1 + cα$ and the independence between $(M_k)_{1≤k≤∞}$ and $(R_1, R_2)$, we see that (3.27), (3.31) and (3.33) follow at once from (3.23) [for (3.33), we also use the fact that $Q(\{S_{R_1} = 1\}) = 1, ∀n ≥ r$]. It remains to check the uniform integrability (3.32): Since $ζ_1 ≤ (1 + cα)R_1 := ζ^*$, we have for any $a > 0$ that

$$Q \otimes P \left[ \frac{ζ_1}{M_k} > a \right] ≤ Q \otimes P \left[ \frac{ζ^*}{M_k} > a^{1/2} \right] + Q \otimes P \left[ \frac{ζ_1}{M_k} > a^{1/2} \right] = E \left[ ζ^* \right] + E(ζ^*) Q \left[ \frac{1}{M_k} > a^{1/2} \right],$$

where we used the independence between $ζ^*$ and $M_k$, and $E$ denotes the expectation with respect to $P$. Clearly, $E[ζ^*1_{(ζ^* > a^{1/2})}] = o(1)$ as $a \to ∞$. Observe that $Q \left[ \frac{1}{M_k} > a^{1/2} \right] = \mathbb{P} \left[ \frac{1}{M_k} > a^{1/2} \right] ≤ e^1 \mathbb{E}e^{-a^{1/2} M_k} ≤ e^1 \mathbb{E}e^{-a^{1/2} M_∞}$. Since $p_0 = 0, M_∞ > 0, P$-a.s., then $e^1 \mathbb{E}e^{-a^{1/2} M_∞} \to 0$ as $a \to ∞$, hence $Q \left[ \frac{1}{M_k} > a^{1/2} \right] \to 0$ uniformly on $k$ and we get (3.32). □
Proof of Theorem 1.1 (case \(\alpha \searrow 0\)). By proposition 3.1,
\[
\lim_{\alpha \searrow 0} \frac{E(\beta(x))}{\alpha} = \frac{\mathcal{D}^0}{2m}.
\]
Then by the velocity representation for \(v_{\alpha}\) in (3.44), it is enough to prove that
\[
\lim_{\alpha \searrow 0} \frac{Q \otimes \tilde{E} \left[ \prod_{i=1}^{R_1-1} f_\infty(S_i) \prod_{i=R_1}^{R_2} f_\infty(S_i) | S_{R_2} - S_{R_1} | \right]}{Q \otimes P \left[ \prod_{i=1}^{R_1-1} f_\infty(S_i) \right]} \sum_{y \geq 1} h(y) = 1.
\]

(3.45)

Since we are interested in the limit \(\alpha \searrow 0\), we may and will assume throughout that \(\alpha < \alpha_0(m, \kappa)\) the constant appeared in Lemma 3.8. We write in this proof \(A \sim B\) if \((A - B)/\alpha \to \alpha \searrow 0\).

Note that \(f_\infty(x) \leq 1 + c\alpha\) for some constant \(c>0\). Mimicking the proof of Lemma 3.8, we therefore get that
\[
Q \otimes P \left[ \prod_{i=1}^{R_1-1} f_\infty(S_i) | S_{R_2} - S_{R_1} | \right] \sim_{\alpha} E[|S_{R_2} - S_{R_1}|] = \frac{1}{P(\tau_S(1) = \infty)} \sim_{\alpha} \frac{m}{m-1}.
\]
In the same way, \(h(y) \sim_{\alpha} (\tau_S(-y) \leq R_1 | \tau_S(1) = \infty)\) for \(y \geq 1\), hence
\[
\sum_{y \geq 1} h(y) \sim_{\alpha} E(|S_{R_1}| | \tau_S(1) = \infty) \sim_{\alpha} \frac{m}{m-1}.
\]
Finally, as \(\alpha \to 0\),
\[
Q \otimes P \left[ \prod_{i=1}^{R_1-1} f_\infty(S_i) \frac{1}{M_\infty} \right] \sim_{\alpha} Q \left[ \frac{1}{M_\infty} \right] = 1,
\]
implying (3.45) and completing the proof of Theorem 1.1. \(\square\)

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References


