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Scaling limit of the path leading to the leftmost particle in a branching random walk

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Summary. We consider a discrete-time branching random walk defined on the real line, which is assumed to be supercritical and in the boundary case. It is known that its leftmost position of the *n*-th generation behaves asymptotically like $\frac{3}{2} \ln n$, provided the non-extinction of the system. The main goal of this paper, is to prove that the path from the root to the leftmost particle, after a suitable normalization, converges weakly to a Brownian excursion in $D([0, 1], \mathbb{R})$.

Keywords. Branching random walk; spinal decomposition.

1 Introduction

We consider a branching random walk, which is constructed according to a point process \mathcal{L} on the line. Precisely speaking, the system is started with one initial particle at the origin. This particle is called the root, denoted by \emptyset . At time 1, the root dies and gives birth to some new particles, which form the first generation. Their positions constitute a point process distributed as \mathcal{L} . At time 2, each of these particles dies and gives birth to new particles whose positions – relative to that of their parent – constitute a new independent copy of \mathcal{L} . The system grows according to the same mechanism.

We denote by \mathbb{T} the genealogical tree of the system, which is clearly a Galton-Watson tree rooted at \emptyset . If a vertex $u \in \mathbb{T}$ is in the *n*-th generation, we write |u| = n and denote its position by V(u). Then $\{V(u), |u| = 1\}$ follows the same law as \mathcal{L} . The family of positions $(V(u); u \in \mathbb{T})$ is viewed as our branching random walk. Throughout the paper, the branching random walk is assumed to be in the boundary case (Biggins and Kyprianou [5]):

(1.1)
$$\mathbf{E}\Big[\sum_{|u|=1} 1\Big] > 1, \quad \mathbf{E}\Big[\sum_{|x|=1} e^{-V(x)}\Big] = 1, \quad \mathbf{E}\Big[\sum_{|x|=1} V(x)e^{-V(x)}\Big] = 0.$$

For any $y \in \mathbb{R}$, let $y_+ := \max\{y, 0\}$ and $\log_+ y := \log(\max\{y, 1\})$. We also assume the following integrability conditions:

(1.2)
$$\mathbf{E}\left[\sum_{|u|=1} V(u)^2 \mathrm{e}^{-V(u)}\right] < \infty$$

(1.3)
$$\mathbf{E}[X(\log_+ X)^2] < \infty, \qquad \mathbf{E}[\widetilde{X}\log_+ \widetilde{X}] < \infty,$$

where

$$X := \sum_{|u|=1} e^{-V(u)}, \qquad \widetilde{X} := \sum_{|u|=1} V(u)_{+} e^{-V(u)}$$

We define I_n to be the leftmost position in the *n*-th generation, i.e.

(1.4)
$$I_n := \inf\{V(u), |u| = n\},\$$

with $\inf \emptyset := \infty$. If $I_n < \infty$, we choose a vertex uniformly in the set $\{u : |u| = n, V(u) = I_n\}$ of leftmost particles at time n and denote it by $m^{(n)}$. We let $[\![\emptyset, m^{(n)}]\!] = \{\emptyset = m_0^{(n)}, m_1^{(n)}, \ldots, m_n^{(n)} := m^{(n)}\}$ be the shortest path in \mathbb{T} relating the root \emptyset to $m^{(n)}$, and introduce the path from the root to $m^{(n)}$ as follows

$$(I_n(k); \ 0 \le k \le n) := (V(m_k^{(n)}); \ 0 \le k \le n).$$

In particular, $I_n(0) = 0$ and $I_n(n) = I_n$. Let σ be the positive real number such that $\sigma^2 = \mathbf{E} \left[\sum_{|u|=1} V(u)^2 e^{-V(u)} \right]$. Our main result is as follows.

Theorem 1.1 The rescaled path $(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le 1)$ converges in law in $D([0,1],\mathbb{R})$, to a normalized Brownian excursion $(e_s; 0 \le s \le 1)$.

Remark 1.2 It has been proved in [1], [11] and [2] that I_n is around $\frac{3}{2} \ln n$. In [3], the authors proved that, for the model of branching Brownian motion, the time reversed path followed by the leftmost particle converges in law to a certain stochastic process.

Let us say a few words about the proof of Theorem 1.1. We first consider the path leading to $m^{(n)}$, by conditioning that its ending point I_n is located atypically below $\frac{3}{2} \ln n - z$

with large z. Then we apply the well-known spinal decomposition to show that this path, conditioned to $\{I_n \leq \frac{3}{2} \ln n - z\}$, behaves like a simple random walk staying positive but tied down at the end. Such a random walk, being rescaled, converges in law to the Brownian excursion (see [9]). We then prove our main result by removing the condition of I_n . The main strategy is borrowed from [2], but with appropriate refinements.

The rest of the paper is organized as follows. In Section 2, we recall the spinal decomposition by a change of measures, which implies the useful many-to-one lemma. We prove a conditioned version of Theorem 1.1 in Section 3. In Section 4, we remove the conditioning and prove the theorem.

Throughout the paper, we use $a_n \sim b_n$ $(n \to \infty)$ to denote $\lim_{n\to\infty} \frac{a_n}{b_n} = 1$; and let $(c_i)_{i\geq 0}$ denote finite and positive constants. We write $\mathbf{E}[f; A]$ for $\mathbf{E}[f\mathbf{1}_A]$. Moreover, $\sum_{\varnothing} := 0$ and $\prod_{\varnothing} := 1$.

2 Lyons' change of measures and spinal decomposition

For any $a \in \mathbb{R}$, let \mathbf{P}_a be the probability measure such that $\mathbf{P}_a((V(u), u \in \mathbb{T}) \in \cdot) = \mathbf{P}((V(u) + a, u \in \mathbb{T}) \in \cdot)$. The corresponding expectation is denoted by \mathbf{E}_a . Let $(\mathcal{F}_n, n \ge 0)$ be the natural filtration generated by the branching random walk and let $\mathcal{F}_{\infty} := \bigvee_{n \ge 0} \mathcal{F}_n$. We introduce the following random variables:

(2.1)
$$W_n := \sum_{|u|=n} e^{-V(u)}, \qquad n \ge 0.$$

It follows immediately from (1.1) that $(W_n, n \ge 0)$ is a non-negative martingale with respect to (\mathcal{F}_n) . It is usually referred as the additive martingale. We define a probability measure \mathbf{Q}_a on \mathcal{F}_∞ such that for any $n \ge 0$,

(2.2)
$$\frac{d\mathbf{Q}_a}{d\mathbf{P}_a}\Big|_{\mathcal{F}_n} := e^a W_n.$$

For convenience, we write \mathbf{Q} for \mathbf{Q}_0 .

Let us give the description of the branching random walk under \mathbf{Q}_a in an intuitive way, which is known as the spinal decomposition. We introduce another point process $\widehat{\mathcal{L}}$ with Radon-Nykodin derivative $\sum_{x \in \mathcal{L}} e^{-x}$ with respect to the law of \mathcal{L} . Under \mathbf{Q}_a , the branching random walk evolves as follows. Initially, there is one particle w_0 located at $V(w_0) = a$. At each step n, particles at generation n die and give birth to new particles independently according to the law of \mathcal{L} , except for the particle w_n which generates its children according to the law of $\widehat{\mathcal{L}}$. The particle w_{n+1} is chosen proportionally to $e^{-V(u)}$ among the children u of w_n . We still call \mathbb{T} the genealogical tree of the process, so that $(w_n)_{n\geq 0}$ is a ray in \mathbb{T} , which is called the spine. This change of probabilities was presented in various forms; see, for example [15], [11] and [8].

It is convenient to use the following notation. For any $u \in \mathbb{T} \setminus \{\emptyset\}$, let \overleftarrow{u} be the parent of u, and

$$\Delta V(u) := V(u) - V(\overleftarrow{u}).$$

Let $\Omega(u)$ be the set of brothers of u, i.e. $\Omega(u) := \{v \in \mathbb{T} : \forall v = \forall u, v \neq u\}$. Let δ denote the Dirac measure. Then under \mathbf{Q}_a , $\sum_{|u|=1} \delta_{\Delta V(u)}$ follows the law of $\widehat{\mathcal{L}}$. Further, We recall the following proposition, from [11] and [15].

Proposition 2.1 (1) For any |u| = n, we have

(2.3)
$$\mathbf{Q}_a[w_n = u | \mathcal{F}_n] = \frac{e^{-V(u)}}{W_n}$$

(2) Under \mathbf{Q}_a , the random variables $\left(\sum_{v \in \Omega(w_n)} \delta_{\Delta V(v)}, \Delta V(w_n)\right)$, $n \ge 1$ are *i.i.d.*.

As a consequence of this proposition, we get the many-to-one lemma as follows:

Lemma 2.2 There exists a centered random walk $(S_n; n \ge 0)$ with $\mathbf{P}_a(S_0 = a) = 1$ such that for any $n \ge 1$ and any measurable function $g : \mathbb{R}^n \to [0, \infty)$, we have

(2.4)
$$\mathbf{E}_a\bigg[\sum_{|u|=n}g(V(u_1),\ldots,V(u_n))\bigg] = \mathbf{E}_a[e^{S_n-a}g(S_1,\ldots,S_n)],$$

where we denote by $\llbracket \emptyset, u \rrbracket = \{ \emptyset =: u_0, u_1 \dots, u_{|u|} := u \}$ the ancestral line of u in \mathbb{T} .

Note that by (1.3), S_1 has the finite variance $\sigma^2 = \mathbf{E}[S_1^2] = \mathbf{E}[\sum_{|u|=1} V(u)^2 e^{-V(u)}].$

2.1 Convergence in law for the one-dimensional random walk

Let us introduce some results about the centered random walk (S_n) with finite variance, which will be used later. For any $0 \le m \le n$, we define $\underline{S}_{[m,n]} := \min_{m \le j \le n} S_j$, and $\underline{S}_n = \underline{S}_{[0,n]}$. We denote by R(x) the renewal function of (S_n) , which is defined as follows:

(2.5)
$$R(x) = \mathbf{1}_{\{x=0\}} + \mathbf{1}_{\{x>0\}} \sum_{k\geq 0} \mathbf{P}(-x \leq S_k < \underline{S}_{n-1}).$$

For the random walk $(-S_n)$, we define $\underline{S}_{[m,n]}^-$, \underline{S}_n^- and $R_-(x)$ similarly. It is known (see [10] p. 360) that there exists $c_0 > 0$ such that

(2.6)
$$\lim_{x \to \infty} \frac{R(x)}{x} = c_0.$$

Moreover, it is shown in [13] that there exist C_+ , $C_- > 0$ such that for any $a \ge 0$,

(2.7)
$$\mathbf{P}_a\left(\underline{S}_n \ge 0\right) \sim \frac{C_+}{\sqrt{n}} R(a);$$

(2.8)
$$\mathbf{P}_a\left(\underline{S}_n^- \ge 0\right) \sim \frac{C_-}{\sqrt{n}} R_-(a).$$

We also state the following inequalities (see Lemmas 2.2 and 2.4 in [4], respectively).

Fact 2.3 (i) There exists a constant $c_1 > 0$ such that for any $b \ge a \ge 0$, $x \ge 0$ and $n \ge 1$,

(2.9)
$$\mathbf{P}(\underline{S}_n \ge -x; \ S_n \in [a-x, b-x]) \le c_1(1+x)(1+b-a)(1+b)n^{-3/2}.$$

(ii) Let $0 < \lambda < 1$. There exists a constant $c_2 > 0$ such that for any $b \ge a \ge 0$, $x, y \ge 0$ and $n \ge 1$,

(2.10)
$$\mathbf{P}_x(S_n \in [y+a, y+b], \underline{S}_n \ge 0, \underline{S}_{[\lambda n, n]} \ge y) \le c_2(1+x)(1+b-a)(1+b)n^{-3/2}.$$

Before we give the next lemma, we recall the definition of lattice distribution (see [10], p. 138). The distribution of a random variable X_1 is lattice, if it is concentrated on a set of points $\alpha + \beta \mathbb{Z}$, with α arbitrary. The largest β satisfying this property is called the span of X_1 . Otherwise, the distribution of X_1 is called non-lattice.

Lemma 2.4 Let $(r_n)_{n\geq 0}$ be a sequence of real numbers such that $\lim_{n\to\infty} \frac{r_n}{\sqrt{n}} = 0$. Let $f: \mathbb{R}_+ \to \mathbb{R}$ be a Riemann integrable function. We suppose that there exists a non-increasing function $\overline{f}: \mathbb{R}_+ \to \mathbb{R}$ such that $|f(x)| \leq \overline{f}(x)$ for any $x \geq 0$ and $\int_{x\geq 0} x\overline{f}(x)dx < \infty$. For $0 < \Delta < 1$, let $F: D([0, \Delta], \mathbb{R}) \to [0, 1]$ be continuous. Let $a \geq 0$.

(I) Non-lattice case. If the distribution of $(S_1 - S_0)$ is non-lattice, then there exists a constant $C_1 > 0$ such that

(2.11)
$$\lim_{n \to \infty} n^{3/2} \mathbf{E} \Big[F\Big(\frac{S_{\lfloor sn \rfloor}}{\sigma \sqrt{n}}; 0 \le s \le \Delta \Big) f(S_n - y); \ \underline{S}_n \ge -a, \ \underline{S}_{\lfloor \Delta n, n \rfloor} \ge y \Big] \\ = C_1 R(a) \int_{x \ge 0} f(x) R_-(x) dx \mathbf{E} [F(e_s; 0 \le s \le \Delta)],$$

uniformly in $y \in [0, r_n]$.

(II) Lattice case. If the distribution of $(S_1 - S_0)$ is supported in $(\alpha + \beta \mathbb{Z})$ with span β , then for any $d \in \mathbb{R}$,

$$(2.12) \quad \lim_{n \to \infty} n^{3/2} \mathbf{E} \Big[F\Big(\frac{S_{\lfloor sn \rfloor}}{\sigma \sqrt{n}}; 0 \le s \le \Delta \Big) f(S_n - y + d); \ \underline{S}_n \ge -a, \ \underline{S}_{[\Delta n, n]} \ge y - d \Big] \\ = C_1 R(a) \beta \sum_{j \ge \lceil -\frac{d}{\beta} \rceil} f(\beta j + d) R_-(\beta j + d) \mathbf{E} [F(e_s; 0 \le s \le \Delta)].$$

uniformly in $y \in [0, r_n] \cap \{\alpha n + \beta \mathbb{Z}\}.$

Proof of Lemma 2.4. The lemma is a refinement of Lemma 2.3 in [2], which proved the convergence in the non-lattice case when a = 0 and $F \equiv 1$. We consider the non-lattice case first. We denote the expectation on the left-hand side of (2.11) by $\chi(F, f)$. Observe that for any $K \in \mathbb{N}_+$,

$$\chi(F, f) = \chi(F, f(x)1_{(0 \le x \le K)}) + \chi(F, f(x)1_{(x > K)}).$$

Since $0 \le F \le 1$, we have $\chi(F, f(x)1_{(x>K)}) \le \chi(1, f(x)1_{(x>K)})$, which is bounded by

$$\sum_{j \ge K} \mathbf{E}_a \Big[f(S_n - y - a); \ \underline{S}_n \ge 0, \ \underline{S}_{[\Delta n, n]} \ge y + a, \ S_n \in [y + a + j, y + a + j + 1] \Big]$$

Recall that $|f(x)| \leq \overline{f}(x)$ with \overline{f} non-increasing. We get that

$$\chi\big(1, f(x)1_{(x>K)}\big) \le \sum_{j\ge K} \overline{f}(j)\mathbf{P}_a\Big[\underline{S}_n \ge 0, \, \underline{S}_{[\Delta n, n]} \ge y+a, \, S_n \in [y+a+j, y+a+j+1]\Big].$$

It then follows from (2.10) that

(2.13)
$$\chi(1, f(x)1_{(x>K)}) \le 2c_2(1+a) \Big(\sum_{j\ge K} \overline{f}(j)(2+j)\Big) n^{-3/2}.$$

Since $\int_0^\infty x\overline{f}(x)dx < \infty$, the sum $\sum_{j\geq K} \overline{f}(j)(2+j)$ decreases to zero as $K \uparrow \infty$. We thus only need to estimate $\chi(F, f(x)1_{(0\leq x\leq K)})$. Note that f is Riemann integrable. It suffices to consider $\chi(F, 1_{(0\leq x\leq K)})$ with K a positive constant.

Applying the Markov property at time $\lfloor \Delta n \rfloor$ shows that

$$\chi\left(F, \mathbf{1}_{(0 \le x \le K)}\right) = \mathbf{E}_{a}\left[F\left(\frac{S_{\lfloor sn \rfloor} - a}{\sigma\sqrt{n}}; 0 \le s \le \Delta\right); \ S_{n} \le y + a + K, \underline{S}_{n} \ge 0, \underline{S}_{\lfloor\Delta n, n\rfloor} \ge y + a\right]$$

$$(2.14) = \mathbf{E}_{a}\left[F\left(\frac{S_{\lfloor sn \rfloor} - a}{\sigma\sqrt{n}}; 0 \le s \le \Delta\right)\Psi_{K}(S_{\lfloor\Delta n \rfloor}); \ \underline{S}_{\lfloor\Delta n \rfloor} \ge 0\right],$$

where $\Psi_K(x) := \mathbf{P}_x \Big[S_{n-\lfloor \Delta n \rfloor} \leq y + a + K, \ \underline{S}_{n-\lfloor \Delta n \rfloor} \geq y + a \Big]$. By reversing time, we obtain that $\Psi_K(x) = \mathbf{P} \Big[\underline{S}_m^- \geq (-S_m) + (y + a - x) \geq -K \Big]$ with $m := n - \lfloor \Delta n \rfloor$. We define τ_n as the first time when the random walk (-S) hits the minimal level during

We define τ_n as the first time when the random walk (-S) hits the minimal level during [0,n], namely, $\tau_n := \inf\{k \in [0,n] : -S_k = \underline{S}_n^-\}$. Define also $\varkappa(z,\zeta;n) := \mathbf{P}(-S_n \in [z, z+\zeta], \underline{S}_n^- \ge 0)$ for any $z, \zeta \ge 0$. Then,

(2.15)
$$\Psi_{K}(x) = \sum_{k=0}^{m} \mathbf{P} \Big[\tau_{m} = k; \ \underline{S}_{m}^{-} \ge (-S_{m}) + (y + a - x) \ge -K \Big] \\ = \sum_{k=0}^{m} \mathbf{P} \Big[-S_{k} = \underline{S}_{k}^{-} \ge -K; \ \varkappa (x - y - a, \underline{S}_{k}^{-} + K; m - k) \Big],$$

where the last equality follows from the Markov property.

Let $\psi(x) := xe^{-x^2/2} \mathbf{1}_{(x \ge 0)}$. Combining Theorem 1 of [6] with (2.7) yields that

(2.16)
$$\varkappa(z,\zeta;n) = \mathbf{P}_0 \Big[-S_n \in [z,z+\zeta]; \ \underline{S}_n \ge 0 \Big] = \frac{C_-\zeta}{\sigma n} \psi \Big(\frac{z}{\sigma\sqrt{n}}\Big) + o(n^{-1}),$$

uniformly in $z \in \mathbb{R}_+$ and ζ in compact sets of \mathbb{R}_+ . Note that ψ is bounded on \mathbb{R}_+ . Therefore, there exists a constant $c_3 > 0$ such that for any $\zeta \in [0, K]$, $z \ge 0$ and $n \ge 0$,

(2.17)
$$\varkappa(z,\zeta;n) \le c_3 \frac{(1+K)}{n+1}$$

Let $k_n := \lfloor \sqrt{n} \rfloor$. We divide the sum on the right-hand side of (2.15) into two parts:

(2.18)
$$\Psi_K(x) = \sum_{k=0}^{k_n} + \sum_{k=k_n+1}^m \mathbf{P} \Big[-S_k = \underline{S}_k^- \ge -K; \ \varkappa(x - y - a, \underline{S}_k^- + K; m - k) \Big].$$

By (2.16), under the assumption that $y = o(\sqrt{n})$, the first part becomes that

$$(2.19) \quad \frac{C_{-}}{\sigma m} \psi\left(\frac{x-a}{\sigma\sqrt{m}}\right) \sum_{k=0}^{k_n} \mathbf{E}\left[\underline{S}_k^- + K; -S_k = \underline{S}_k^- \ge -K\right] + o(n^{-1}) \sum_{k=0}^{k_n} \mathbf{P}\left[-S_k = \underline{S}_k^- \ge -K\right] \\ = \frac{C_{-}}{\sigma m} \psi\left(\frac{x-a}{\sigma\sqrt{m}}\right) \int_0^K R_-(u) du + o(n^{-1}),$$

where the last equation comes from the fact that $\sum_{k\geq 0} \mathbf{E} \left[\underline{S}_k^- + K; -S_k = \underline{S}_k^- \geq -K \right] = \int_0^K R_-(u) du$. On the other hand, using (2.17) for $\varkappa(x - y - a, \underline{S}_k^- + K; m - k)$ and then applying (i) of Fact 2.3 imply that for *n* large enough, the second part of (2.18) is bounded

by

(2.20)
$$\sum_{k=k_n+1}^{m} c_3 \frac{1+K}{m+1-k} \mathbf{P} \left(\underline{S}_k^- \ge -K, \ -S_k \in [-K, 0] \right)$$
$$\leq c_4 \sum_{k=k_n+1}^{m} \frac{(1+K)^3}{(m+1-k)k^{3/2}} = o(n^{-1}).$$

By (2.19) and (2.20), we obtain that as n goes to infinity,

(2.21)
$$\Psi_K(x) = o(n^{-1}) + \frac{C_-}{\sigma(n - \lfloor \Delta n \rfloor)} \psi\left(\frac{x - a}{\sigma\sqrt{n - \lfloor \Delta n \rfloor}}\right) \int_0^K R_-(u) du,$$

uniformly in $x \ge 0$ and $y \in [0, r_n]$. Plugging it into (2.14) and then combining with (2.7) yield that

$$\chi(F, 1_{(0 \le x \le K)}) = o(n^{-3/2}) + \frac{C_{-}}{\sigma(1 - \Delta)n} \int_{0}^{K} R_{-}(u) du \\ \times \frac{C_{+}R(a)}{\sqrt{\Delta n}} \mathbf{E}_{a} \Big[F\Big(\frac{S_{\lfloor sn \rfloor} - a}{\sigma\sqrt{n}}; 0 \le s \le \Delta\Big) \psi\Big(\frac{S_{\Delta n} - a}{\sigma\sqrt{(1 - \Delta)n}}\Big) \Big| \underline{S}_{\Delta n} \ge 0 \Big].$$

Theorem 1.1 of [7] says that under the conditioned probability $\mathbf{P}_a\left(\cdot \left| \underline{S}_{\Delta n} \geq 0 \right), \left(\frac{S_{\lfloor r \Delta n \rfloor}}{\sigma \sqrt{\Delta n}}; 0 \leq r \leq 1 \right)$ converges in law to a Brownian meander, denoted by $(\mathcal{M}_r; 0 \leq r \leq 1)$. Therefore,

$$\chi(F, 1_{(0 \le x \le K)}) \sim \frac{C_- C_+ R(a)}{\sigma n^{3/2} (1 - \Delta) \sqrt{\Delta}} \int_0^K R_-(u) du \mathbf{E} \Big[F\Big(\sqrt{\Delta} \mathcal{M}_{s/\Delta}; 0 \le s \le \Delta\Big) \psi\Big(\frac{\sqrt{\Delta} \mathcal{M}_1}{\sqrt{1 - \Delta}}\Big) \Big].$$

It remains to check that

(2.22)
$$\frac{1}{(1-\Delta)\sqrt{\Delta}}\mathbf{E}\Big[F\Big(\sqrt{\Delta}\mathcal{M}_{s/\Delta}; 0 \le s \le \Delta\Big)\psi\Big(\frac{\sqrt{\Delta}\mathcal{M}_1}{\sqrt{1-\Delta}}\Big)\Big] = \sqrt{\frac{\pi}{2}}\mathbf{E}\Big[F(e_s; 0 \le s \le \Delta)\Big].$$

Let $(R_s; 0 \le s \le 1)$ be a standard three-dimensional Bessel process. Then, as is shown in [12],

$$\frac{1}{(1-\Delta)\sqrt{\Delta}} \mathbf{E} \Big[F\Big(\sqrt{\Delta}\mathcal{M}_{s/\Delta}; 0 \le s \le \Delta\Big) \psi\Big(\frac{\sqrt{\Delta}\mathcal{M}_1}{\sqrt{1-\Delta}}\Big) \Big] \\ = \sqrt{\frac{\pi}{2}} \frac{1}{(1-\Delta)\sqrt{\Delta}} \mathbf{E} \Big[\frac{1}{R_1} F\Big(\sqrt{\Delta}R_{s/\Delta}; 0 \le s \le \Delta\Big) \psi\Big(\frac{\sqrt{\Delta}R_1}{\sqrt{1-\Delta}}\Big) \Big], \\ = \sqrt{\frac{\pi}{2}} \mathbf{E} \Big[\frac{1}{(1-\Delta)^{3/2}} e^{-\frac{R_\Delta^2}{2(1-\Delta)}} F\Big(R_s; 0 \le s \le \Delta\Big) \Big],$$

where the last equation follows from the scaling property of Bessel process. Let $(r_s; 0 \le s \le 1)$ be a standard three-dimensional Bessel bridge. Note that for any $\Delta < 1$, $(r_s; 0 \le s \le \Delta)$ is equivalent to $(R_s; 0 \le s \le \Delta)$, with density $(1 - \Delta)^{-3/2} \exp(-\frac{R_{\Delta}^2}{2(1 - \Delta)})$ (see p. 468 (3.11) of [16]). Thus,

$$\frac{1}{(1-\Delta)\sqrt{\Delta}}\mathbf{E}\Big[F\Big(\sqrt{\Delta}\mathcal{M}_{s/\Delta}; 0 \le s \le \Delta\Big)\psi\Big(\frac{\sqrt{\Delta}\mathcal{M}_1}{\sqrt{1-\Delta}}\Big)\Big] = \sqrt{\frac{\pi}{2}}\mathbf{E}\Big[F(r_s; 0 \le s \le \Delta)\Big].$$

Since a normalized Brownian excursion is exactly a standard three-dimensional Bessel bridge, this yields (2.22). Therefore, (2.11) is proved by taking $C_1 = \sqrt{\frac{\pi}{2}} \frac{C_-C_+}{\sigma}$.

The proof of the lemma in the lattice case is along the same lines, except that we use Theorem 2 (instead of Theorem 1) of [6]. \Box

3 Conditioning on the event $\{I_n \leq \frac{3}{2} \ln n - z\}$

On the event $\{I_n \leq \frac{3}{2} \ln n - z\}$, we analyze the sample path leading to a particle located at the leftmost position at the *n*th generation. For $z \geq 0$ and $n \geq 1$, let $a_n(z) := \frac{3}{2} \ln n - z$ if the distribution of \mathcal{L} is non-lattice and let $a_n(z) := \alpha n + \beta \lfloor \frac{\frac{3}{2} \ln n - \alpha n}{\beta} \rfloor - z$ if the distribution of \mathcal{L} is supported by $\alpha + \beta \mathbb{Z}$. This section is devoted to the proof of the following proposition.

Proposition 3.1 For any $\Delta \in (0,1]$ and any continuous functional $F : D([0,\Delta], \mathbb{R}) \rightarrow [0,1],$

(3.1)
$$\lim_{z \to \infty} \limsup_{n \to \infty} \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le \Delta \right) \middle| I_n \le a_n(z) \right] - \mathbf{E} \left[F(e_s; 0 \le s \le \Delta) \right] \right| = 0.$$

We begin with some preliminary results.

For any $0 < \Delta < 1$ and $L, K \ge 0$, we denote by $J^{\Delta}_{z,K,L}(n)$ the following collection of particles:

(3.2)
$$\left\{ u \in \mathbb{T} : |u| = n, \ V(u) \le a_n(z), \ \min_{0 \le k \le n} V(u_k) \ge -z + K, \ \min_{\Delta n \le k \le n} V(u_k) \ge a_n(z+L) \right\}.$$

Lemma 3.2 For any $\varepsilon > 0$, there exists $L_{\varepsilon} > 0$ such that for any $L \ge L_{\varepsilon}$, $n \ge 1$ and $z \ge K \ge 0$,

(3.3)
$$\mathbf{P}\Big(m^{(n)} \notin J^{\Delta}_{z,K,L}(n), \ I_n \le a_n(z)\Big) \le \Big(e^K + \varepsilon(1+z-K)\Big)e^{-z}.$$

Proof. It suffices to show that for any $\varepsilon \in (0, 1)$, there exists $L_{\varepsilon} \geq 1$ such that for any $L \geq L_{\varepsilon}, n \geq 1$ and $z \geq K \geq 0$,

(3.4)
$$\mathbf{P}\Big(\exists |u| = n : V(u) \le a_n(z), \ u \notin J^{\Delta}_{z,K,L}(n)\Big) \le \Big(e^K + \varepsilon(1+z-K)\Big)e^{-z}.$$

We observe that

(3.5)
$$\mathbf{P}\Big(\exists |u| = n : V(u) \le a_n(z), \ u \notin J^{\Delta}_{z,K,L}(n)\Big) \le \mathbf{P}\Big(\exists u \in \mathbb{T} : V(u) \le -z + K\Big) \\ + \mathbf{P}\Big(\exists |u| = n : V(u) \le a_n(z), \ \min_{0 \le k \le n} V(u_k) \ge -z + K, \ \min_{\Delta n \le k \le n} V(u_k) \le a_n(z+L)\Big).$$

On the one hand, by (2.4),

$$(3.6) \quad \mathbf{P}\Big(\exists u \in \mathbb{T} : V(u) \leq -z+k\Big) \leq \sum_{n\geq 0} \mathbf{E}\Big[\sum_{|u|=n} \mathbf{1}_{\{V(u)\leq -z+K<\min_{k< n} V(u_k)\}}\Big]$$
$$= \sum_{n\geq 0} \mathbf{E}[e^{S_n}; \ S_n \leq -z+K < \underline{S}_{n-1}] \leq e^{-z+K}.$$

On the other hand, denoting $A_n(z) := [a_n(z) - 1, a_n(z)]$ for any $z \ge 0$,

$$\mathbf{P}\Big(\exists |u| = n : V(u) \leq a_n(z), \min_{0 \leq k \leq n} V(u_k) \geq -z + K, \min_{\Delta n \leq k \leq n} V(u_k) \leq a_n(z+L)\Big)$$

$$= \mathbf{P}_{z-K}\Big(\exists |u| = n : V(u) \leq a_n(K), \min_{0 \leq k \leq n} V(u_k) \geq 0, \min_{\Delta n \leq k \leq n} V(u_k) \leq a_n(K+L)\Big)$$

$$\leq \sum_{\ell \geq L+K} \sum_{j=K}^{j=K+\ell} \mathbf{P}_{z-K}\Big(\exists |u| = n : V(u) \in A_n(j), \min_{0 \leq k \leq n} V(u_k) \geq 0, \min_{\Delta n \leq k \leq n} V(u_k) \in A_n(\ell)\Big).$$

According to Lemma 3.3 in [2], there exist constants $1 > c_5 > 0$ and $c_6 > 0$ such that for any $n \ge 1$, $L \ge 0$ and $x, z \ge 0$,

(3.7)
$$\mathbf{P}_{x}\Big(\exists u \in \mathbb{T} : |u| = n, \ V(u) \in A_{n}(z), \ \min_{0 \le k \le n} V(u_{k}) \ge 0, \ \min_{\Delta n \le k \le n} V(u_{k}) \in A_{n}(z+L)\Big) \\ \le \ c_{6}(1+x)e^{-c_{5}L}e^{-x-z}.$$

Hence, combining (3.6) with (3.5) yields that

$$\begin{aligned} \mathbf{P} \Big(\exists |u| &= n : V(u) \leq a_n(z), \ u \not\in J_{z,K,L}^{\Delta}(n) \Big) \\ \leq & e^{-z+K} + \sum_{\ell \geq L} \sum_{0 \leq j \leq \ell} c_6 (1+z-K) e^{-c_5(\ell-j)} e^{-z-j} \\ \leq & \Big(e^K + c_7 \sum_{\ell \geq L} e^{-c_5 \ell} (1+z-K) \Big) e^{-z}, \end{aligned}$$

where the last inequality comes from the fact that $\sum_{j\geq 0} e^{-(1-c_5)j} < \infty$. We take $L_{\varepsilon} = -c_8 \ln \varepsilon$ so that $c_7 \sum_{\ell \geq L} e^{-c_5\ell} \leq \varepsilon$ for all $L \geq L_{\varepsilon}$. Therefore, for any $L \geq L_{\varepsilon}$, $n \geq 1$ and $z \geq K \geq 0$,

(3.8)
$$\mathbf{P}\Big(\exists |u| = n : V(u) \le a_n(z), \ u \notin J^{\Delta}_{z,K,L}(n)\Big) \le \Big(e^K + \varepsilon(1+z-K)\Big)e^{-z},$$

which completes the proof. \Box

For $b \in \mathbb{Z}_+$, we define

(3.9)
$$\mathcal{E}_n = \mathcal{E}_n(z, b) := \{ \forall k \le n - b, \min_{u \ge w_k, |u| = n} V(u) > a_n(z) \}$$

We note that on the event $\mathcal{E}_n \cap \{I_n \leq a_n(z)\}$, any particle located at the leftmost position must be separated from the spine after time n - b.

Lemma 3.3 For any $\eta > 0$ and L > 0, there exist $K(\eta) > 0$, $B(L, \eta) \ge 1$ and $N(\eta) \ge 1$ such that for any $b \ge B(L, \eta)$, $n \ge N(\eta)$ and $z \ge K \ge K(\eta)$,

(3.10)
$$\mathbf{Q}\Big(\mathcal{E}_n^c, \ w_n \in J^{\Delta}_{z,K,L}(n)\Big) \le \eta (1+L)^2 (1+z-K) n^{-3/2}.$$

We feel free to omit the proof of Lemma 3.3 since it is just a slightly stronger version of Lemma 3.8 in [2]. It follows from the same arguments.

Let us turn to the proof of Proposition 3.1. We break it up into 3 steps. Step (I) (The conditioned convergence of $(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \leq s \leq \Delta)$ for $\Delta < 1$ in the non-lattice case)

Assume that the distribution of \mathcal{L} is non-lattice in this step. Recall that $a_n(z) = \frac{3}{2} \ln n - z$. The tail distribution of I_n has been given in Propositions 1.3 and 4.1 of [2], recalled as follows.

Fact 3.4 ([2]) There exists a constant C > 0 such that

(3.11)
$$\lim_{z \to \infty} \limsup_{n \to \infty} \left| \frac{e^z}{z} \mathbf{P}(I_n \le a_n(z)) - C \right| = 0.$$

Furthermore, for any $\varepsilon > 0$, there exist $N_{\varepsilon} \ge 1$ and $\Lambda_{\varepsilon} > 0$ such that for any $n \ge N_{\varepsilon}$ and $\Lambda_{\varepsilon} \le z \le \frac{3}{2} \ln n - \Lambda_{\varepsilon}$,

(3.12)
$$\left|\frac{e^z}{z}\mathbf{P}(I_n \le a_n(z)) - C\right| \le \varepsilon.$$

For any continuous functional $F: D([0, \Delta], \mathbb{R}) \to [0, 1]$, it is convenient to write that

(3.13)
$$\Sigma_n(F,z) := \mathbf{E}\left[F\left(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le \Delta\right) \mathbf{1}_{\{I_n \le a_n(z)\}}\right].$$

In particular, if $F \equiv 1$, $\Sigma_n(1, z) = \mathbf{P}(I_n \leq a_n(z))$. Thus,

(3.14)
$$\frac{\Sigma_n(F,z)}{\Sigma_n(1,z)} = \mathbf{E}\left[F\left(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le \Delta\right) \middle| I_n \le a_n(z)\right].$$

Let us prove the following convergence for $0 < \Delta < 1$,

(3.15)
$$\lim_{z \to \infty} \limsup_{n \to \infty} \left| \frac{\Sigma_n(F, z)}{\Sigma_n(1, z)} - \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| = 0.$$

Proof of (3.15). For any $n \ge 1, L \ge 0$ and $z \ge K \ge 0$, let

(3.16)
$$\Pi_n(F) = \Pi_n(F, z, K, L) := \mathbf{E}\left[F\left(\frac{I_n(sn)}{\sigma\sqrt{n}}; 0 \le s \le \Delta\right)\mathbf{1}_{\{m^{(n)} \in J^{\Delta}_{z,K,L}(n)\}}\right].$$

By Lemma 3.2, we obtain that for $L \ge L_{\varepsilon}$, $n \ge 1$ and $z \ge K \ge 0$,

(3.17)
$$\left|\Sigma_n(F,z) - \Pi_n(F)\right| \le \left(e^K + \varepsilon(1+z-K)\right)e^{-z}.$$

Note that $m^{(n)}$ is chosen uniformly among the particles located at the leftmost position. Thus,

$$\Pi_{n}(F) = \mathbf{E} \bigg[\sum_{|u|=n} \mathbf{1}_{(u=m^{(n)}, u \in J_{z,K,L}^{\Delta}(n))} F\bigg(\frac{V(u_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le \Delta \bigg) \bigg]$$
$$= \mathbf{E} \bigg[\frac{1}{\sum_{|u|=n} \mathbf{1}_{(V(u)=I_{n})}} \sum_{|u|=n} \mathbf{1}_{(V(u)=I_{n}, u \in J_{z,K,L}^{\Delta}(n))} F\bigg(\frac{V(u_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le \Delta \bigg) \bigg].$$

Applying the change of measures given in (2.2), it follows from Proposition 2.1 that

$$(3.18) \quad \Pi_n(F) = \mathbf{E}_{\mathbf{Q}} \bigg[\frac{e^{V(w_n)}}{\sum_{|u|=n} \mathbf{1}_{(V(u)=I_n)}} \mathbf{1}_{(V(w_n)=I_n, w_n \in J_{z,K,L}^{\Delta}(n))} F\bigg(\frac{V(w_{\lfloor sn \rfloor})}{\sigma \sqrt{n}}; 0 \le s \le \Delta \bigg) \bigg].$$

In order to estimate Π_n , we restrict ourselves to the event \mathcal{E}_n . Define

$$\Lambda_n(F) := \mathbf{E}_{\mathbf{Q}} \bigg[\frac{e^{V(w_n)}}{\sum_{|u|=n} \mathbf{1}_{(V(u)=I_n)}} \mathbf{1}_{(V(w_n)=I_n, w_n \in J^{\Delta}_{z,K,L}(n))} F\Big(\frac{V(w_{\lfloor sn \rfloor})}{\sigma \sqrt{n}}; 0 \le s \le \Delta \Big); \ \mathcal{E}_n \bigg].$$

In view of Lemma 3.3, for any $b \ge B(L,\eta), n \ge N(\eta)$ and $z \ge K \ge K(\eta)$,

(3.19)
$$\begin{aligned} \left|\Pi_{n}(F) - \Lambda_{n}(F)\right| &\leq \mathbf{E}_{\mathbf{Q}} \Big[e^{V(w_{n})}; \ w_{n} \in J_{z,K,L}^{\Delta}(n), \ \mathcal{E}_{n}^{c} \Big] \\ &\leq e^{-z} n^{-3/2} \mathbf{Q} \Big(\mathcal{E}_{n}^{c}, \ w_{n} \in J_{z,K,L}^{\Delta}(n) \Big) \\ &\leq \eta (1+L)^{2} (1+z-K) e^{-z}. \end{aligned}$$

On the event $\mathcal{E}_n \cap \{I_n \leq a_n(z)\}, \Lambda_n(F)$ equals

$$\mathbf{E}_{\mathbf{Q}}\bigg[\frac{e^{V(w_n)}}{\sum_{u>w_{n-b},|u|=n}\mathbf{1}_{(V(u)=I_n)}}\mathbf{1}_{(V(w_n)=I_n,\ w_n\in J^{\Delta}_{z,K,L}(n))}F\bigg(\frac{V(w_{\lfloor sn\rfloor})}{\sigma\sqrt{n}}; 0\le s\le \Delta\bigg);\ \mathcal{E}_n\bigg].$$

Let, for $x \ge 0$, L > 0, and $b \ge 1$,

(3.20)
$$f_{L,b}(x) := \mathbf{E}_{\mathbf{Q}_{x}} \left[\frac{e^{V(w_{b})-L} \mathbf{1}_{\{V(w_{b})=I_{b}\}}}{\sum_{|u|=b} \mathbf{1}_{\{V(u)=I_{b}\}}}, \min_{0 \le k \le b} V(w_{k}) \ge 0, V(w_{b}) \le L \right]$$
$$\leq \mathbf{Q}_{x} \left(\min_{0 \le k \le b} V(w_{k}) \ge 0, V(w_{b}) \le L \right).$$

We choose n large enough so that $\Delta n \leq n-b$. Thus, applying the Markov property at time n-b yields that

(3.21)
$$\Lambda_n(F) = n^{3/2} e^{-z} \mathbf{E}_{\mathbf{Q}} \Big[F\Big(\frac{V(w_{\lfloor sn \rfloor})}{\sigma \sqrt{n}}; 0 \le s \le \Delta \Big) f_{L,b}(V(w_{n-b}) - a_n(z+L)); \\ \min_{0 \le k \le n-b} V(w_k) \ge -z + K, \ \min_{\Delta n \le k \le n-b} V(w_k) \ge a_n(z+L), \ \mathcal{E}_n \Big].$$

Let us introduce the following quantity by removing the restriction to \mathcal{E}_n :

(3.22)
$$\Lambda_n^I(F) := n^{3/2} e^{-z} \mathbf{E}_{\mathbf{Q}} \Big[F\Big(\frac{V(w_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le \Delta\Big) f_{L,b}(V(w_{n-b}) - a_n(z+L)); \\ \min_{0 \le k \le n-b} V(w_k) \ge -z + K, \ \min_{\Delta n \le k \le n-b} V(w_k) \ge a_n(z+L) \Big].$$

We immediately observe that

(3.23)
$$\left| \Lambda_n(F) - \Lambda_n^I(F) \right| \leq n^{3/2} e^{-z} \mathbf{Q} \Big(f_{L,b}(V(w_{n-b}) - a_n(z+L)), \\ \min_{0 \leq k \leq n-b} V(w_k) \geq -z + K, \min_{\Delta n \leq k \leq n-b} V(w_k) \geq a_n(z+L); \ (\mathcal{E}_n)^c \Big).$$

By (3.20), we check that $|\Lambda_n(F) - \Lambda_n^I(F)| \leq n^{3/2} e^{-z} \mathbf{Q}(w_n \in J^{\Delta}_{z,K,L}(n), (\mathcal{E}_n)^c)$. Applying Lemma 3.3 again implies that

(3.24)
$$\left| \Lambda_n(F) - \Lambda_n^I(F) \right| \le \eta (1+L)^2 (1+z-K) e^{-z}.$$

Combining with (3.19), we obtain that for any $b \ge B(L,\eta)$, $z \ge K \ge K(\eta)$ and n large enough,

(3.25)
$$\left| \Pi_n(F) - \Lambda_n^I(F) \right| \le 2\eta (1+L)^2 (1+z-K)e^{-z}.$$

Note that $(V(w_k); k \ge 1)$ is a centered random walk under **Q** and that it is proved in [2] that $f_{L,b}$ satisfies the conditions of Lemma 2.4. By (I) of Lemma 2.4, we get that

(3.26)
$$\lim_{n \to \infty} \Lambda_n^I(F) = \alpha_{L,b}^I R(z - K) e^{-z} \mathbf{E}[F(e_s, 0 \le s \le \delta)],$$

where $\alpha_{L,b}^I := C_1 \int_{x \ge 0} f_{L,b}(x) R_-(x) dx \in [0, \infty)$. Thus, by (3.25), one sees that for any $b \ge B(L, \eta)$ and $z \ge K \ge K(\eta)$,

(3.27)
$$\lim_{n \to \infty} \sup_{n \to \infty} \left| \Pi_n(F) - \alpha_{L,b}^I R(z - K) e^{-z} \mathbf{E}[F(e_s, \ 0 \le s \le \Delta)] \right| \le 2\eta (1 + L)^2 (1 + z - K) e^{-z}.$$

Going back to (3.17), we deduce that for any $L \ge L_{\varepsilon}$, $b \ge B(L, \eta)$ and $z \ge K \ge K(\eta)$,

$$\lim_{n \to \infty} \sup_{n \to \infty} \left| \Sigma_n(F, z) - \alpha_{L,b}^I R(z - K) e^{-z} \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right|$$

$$\leq 2\eta (1 + L)^2 (1 + z - K) e^{-z} + \left(e^K + \varepsilon (1 + z - K) \right) e^{-z}.$$

Recall that $\lim_{z\to\infty} \frac{R(z)}{z} = c_0$. We multiply each term by $\frac{e^z}{z}$, and then let z go to infinity to conclude that

(3.28)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \left| \frac{e^z}{z} \Sigma_n(F, z) - \alpha_{L,b}^I c_0 \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| \le 2\eta (1+L)^2 + \varepsilon.$$

In particular, taking $F \equiv 1$ gives that

(3.29)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \left| \frac{e^z}{z} \mathbf{P}(I_n \le a_n(z)) - \alpha_{L,b}^I c_0 \right| \le 2\eta (1+L)^2 + \varepsilon$$

It follows from Fact 3.4 that $|C - \alpha_{L,b}^{I}c_{0}| \leq 2\eta(1+L)^{2} + \varepsilon$. We thus choose $0 < \varepsilon < C/10$ and $0 < \eta \leq \frac{\varepsilon}{2(1+L_{\varepsilon})^{2}}$ so that $2C > \alpha_{L_{\varepsilon},b}^{I}c_{0} > C/2 > 0$.

Therefore, for any $\varepsilon \in (0, C/10)$, $0 < \eta \leq \frac{\varepsilon}{2(1+L_{\varepsilon})^2}$, $L = L_{\varepsilon}$ and $b \geq B(L_{\varepsilon}, \eta)$,

(3.30)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \left| \frac{\sum_n (F, z)}{\sum_n (1, z)} - \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| \le \frac{4\varepsilon}{C/2 - 2\varepsilon},$$

which completes the proof of (3.15) in the non-lattice case. Step (II) (The conditioned convergence of $(\frac{I_n(sn)}{\sigma\sqrt{n}}; 0 \leq s \leq \Delta)$ for $\Delta < 1$ in the lattice case) Assume that the law of \mathcal{L} is supported by $\alpha + \beta \mathbb{Z}$ with span β . Recall that $a_n(0) = \alpha n + \beta \lfloor \frac{\frac{3}{2} \ln n - \alpha n}{\beta} \rfloor$ and that $a_n(z) = a_n(0) - z$. We use the same notation of Step (I). Let us prove

(3.31)
$$\lim_{\beta \mathbb{Z} \ni z \to \infty} \limsup_{n \to \infty} \left| \frac{\Sigma_n(F, z)}{\Sigma_n(1, z)} - \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| = 0.$$

Suppose that $z \in \beta \mathbb{Z}$. Whereas the arguments of Step (I), we obtain that for any $L \geq L_{\varepsilon}$, $b \geq B(L, \eta), z \geq K \geq K(\eta)$ and n sufficiently large,

(3.32)
$$\left| \Sigma_n(F,z) - \Lambda_n^{II}(F) \right| \le 2\eta (1+L)^2 (1+z-K)e^{-z} + \left(e^K + \varepsilon(1+z-K) \right) e^{-z},$$

where

$$\Lambda_n^{II}(F) = \Lambda^{II}(F, z, K, L, b) := e^{a_n(0)} e^{-z} \mathbf{E}_{\mathbf{Q}} \Big[F\Big(\frac{V(w_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le \Delta\Big) \times f_{L,b}\Big(V(w_{n-b} - a_n(z+L))\Big); \min_{0 \le k \le n-b} V(w_k) \ge -z + K, \min_{\Delta n \le k \le n-b} V(w_k) \ge a_n(z+L) \Big].$$

Under \mathbf{Q} , the distribution of $V(w_1) - V(w_0)$ is also supported by $\alpha + \beta \mathbb{Z}$. Let $d = d(L, b) := \beta \lceil \frac{\alpha b - L}{\beta} \rceil - \alpha b + L$ and $\lambda_n := n^{3/2} e^{-a_n(0)}$. Recall that $f_{L,b}$ is well defined in (3.20), it follows from (II) of Lemma 2.4 that

(3.33)
$$\lim_{n \to \infty} \lambda_n \Lambda_n^{II}(F) = \alpha_{L,b}^{II} R(z - K) e^{-z} \mathbf{E}[F(e_s, \ 0 \le s \le \Delta)]$$

where $\alpha_{L,b}^{II} := C_1 \beta \sum_{j \ge 0} f_{L,b}(\beta j + d) R_-(\beta j + d) \in [0, \infty)$. Observe that $1 \le \lambda_n \le e^{\beta}$. Combining with (3.32), we conclude that

(3.34)
$$\lim_{\beta \mathbb{Z} \ni z \to \infty} \sup_{n \to \infty} \left| \frac{e^z}{z} \lambda_n \Sigma_n(F, z) - \alpha_{L, b}^{II} c_0 \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| \le e^\beta (2\eta (1+L)^2 + \varepsilon).$$

We admit for the moment that there exist $0 < c_9 < c_{10} < \infty$ such that $\alpha_{L,b}^{II} \in [c_9, c_{10}]$ for all L, b large enough. Then take $\varepsilon < \frac{c_9c_0}{4e^{\beta}}$, $L = L_{\varepsilon}$, $\eta = \frac{\varepsilon}{2(1+L_{\varepsilon})^2}$ and $b \ge B(L_{\varepsilon}, \eta)$ so that $e^{\beta}(2\eta(1+L)^2 + \varepsilon) < c_9c_0/2 \le \alpha_{L_{\varepsilon},b}^{II}c_0/2 \le 2c_{10}c_0$. Note that $\frac{\sum_n(F,z)}{\sum_n(1,z)} = \frac{\frac{e^z}{z}\lambda_n\sum_n(F,z)}{\frac{e^z}{z}\lambda_n\sum_n(1,z)}$. We thus deduce from (3.34) that

(3.35)
$$\lim_{\beta \mathbb{Z} \ni z \to \infty} \sup_{n \to \infty} \left| \frac{\Sigma_n(F, z)}{\Sigma_n(1, z)} - \mathbf{E}[F(e_s, 0 \le s \le \Delta)] \right| \le \frac{4\varepsilon}{c_9 c_0 / e^\beta - 2\varepsilon}$$

which tends to zero as $\varepsilon \downarrow 0$.

It remains to prove that $\alpha_{L,b}^{II} \in [c_9, c_{10}]$ for all L, b large enough. Instead of investigating the entire system, we consider the branching random walk killed at 0. Define

(3.36)
$$I_n^{kill} := \inf\{V(u) : |u| = n, V(u_k) \ge 0, \ \forall 0 \le k \le n\},$$

and we get the following fact from Corollary 3.4 and Lemma 3.6 of [2].

Fact 3.5 ([2]) There exists a constant $c_{11} > 0$ such that for any $n \ge 1$ and $x, z \ge 0$,

(3.37)
$$\mathbf{P}_{x}(I_{n}^{kill} \le a_{n}(z)) \le c_{11}(1+x)e^{-x-z}$$

Moreover, there exists $c_{12} > 0$ such that for any $n \ge 1$ and $z \in [0, a_n(1)]$,

$$\mathbf{P}(I_n^{kill} \le a_n(z)) \ge c_{12}e^{-z}.$$

Even though Fact 3.5 is proved in [2] under the assumption that the distribution of \mathcal{L} is non-lattice, the lattice case is actually recovered from that proof.

Analogically, let $m^{kill,(n)}$ be the particle chosen uniformly in the set $\{u : |u| = n, V(u) = I_n^{kill}, \min_{0 \le k \le n} V(u_k) \ge 0\}$. Moreover, let $\Sigma_n^{kill}(1, z) := \mathbf{P} \Big[I_n^{kill} \le a_n(z) \Big]$ and $\Pi_n^{kill}(1, z, z, L) := \mathbf{P} \Big[I_n^{kill} \le a_n(z), m^{kill,(n)} \in J_{z,z,L}^{\Delta}(n) \Big]$. By (3.7) again, we check that for all $L \ge L_{\varepsilon}$,

$$(3.39) \qquad \left| \begin{aligned} \Sigma_n^{kill}(1,z) - \Pi_n^{kill}(1,z,z,L) \right| \\ \leq \mathbf{P} \Big[\exists |u| = n : V(u) \leq a_n(z); \min_{0 \leq k \leq n} V(u_k) \geq 0; \min_{\Delta n \leq k \leq n} V(u_k) \leq a_n(z+L) \Big] \\ \leq \varepsilon e^{-z}. \end{aligned}$$

Recounting the arguments of Step (1), one sees that for any $L \ge L_{\varepsilon}$, $b \ge B(L, \eta)$, $z \ge K(\eta)$ and *n* sufficiently large,

(3.40)
$$\left| \Pi_n^{kill}(1, z, z, L) - \Lambda_n^{kill} \right| \le 2\eta (1+L)^2 e^{-z},$$

where

(3.41)
$$\Lambda_n^{kill} := \mathbf{E}_{\mathbf{Q}} \Big[f^{kill}(V(w_{n-b})); \min_{0 \le k \le n-b} V(w_k) \ge 0, \min_{\Delta n \le k \le n-b} V(w_k) \ge a_n(z+L) \Big],$$

with $f^{kill}(x) := \mathbf{E}_{\mathbf{Q}_x} \Big[\frac{e^{V(w_b)} \mathbf{1}_{\{V(w_b) = I_b^{kill}\}}}{\sum_{|u| = b} \mathbf{1}_{\{V(u) = I_b^{kill}, \min_{0 \le j \le b} V(u_j) \ge 0\}}}; \min_{0 \le k \le b} V(w_k) \ge a_n(z+L), V(w_b) \le a_n(z) \Big].$ For $\varepsilon > 0$ and n sufficiently large, it has been proved in [2] that

(3.42)
$$\left| e^{z} \Lambda_{n}^{II}(1, z, z, L, b) - \Lambda_{n}^{kill} \right| \leq \varepsilon.$$

Recalling the convergence (3.33) with K = z and $F \equiv 1$, we deduce from (3.39), (3.40) and (3.42) that for any $L \ge L_{\varepsilon}$, $b \ge B(L, \eta)$ and $z \ge K(\eta)$,

(3.43)
$$\limsup_{n \to \infty} \left| \lambda_n \Sigma_n^{kill}(1, z) - \alpha_{L,b}^{II} e^{-z} \right| \le e^\beta \Big(2\eta (1+L)^2 + 2\varepsilon \Big) e^{-z},$$

since R(0) = 1 and $1 \leq \lambda_n \leq e^{\beta}$. Fact 3.5 implies that $c_{12} \leq e^z \lambda_n \mathbf{P}(I_n^{kill} \leq a_n(z)) \leq c_{11}e^{\beta}$. Hence, we obtain that

(3.44)
$$c_{12} - e^{\beta} \Big(2\eta (1+L)^2 + 2\varepsilon \Big) \le \alpha_{L,b}^{II} \le e^{\beta} c_{11} + e^{\beta} \Big(2\eta (1+L)^2 + 2\varepsilon \Big).$$

Let $c_{10} := c_{11}e^{\beta} + c_{12}$ and $c_9 := 3c_{12}/4 > 0$. For any $\varepsilon < e^{-\beta}c_{12}/12$, we take $L = L_{\varepsilon}$ and $\eta \le \varepsilon/2(1 + L_{\varepsilon})^2$. Then $c_{10} > \alpha_{L,b}^{II} \ge c_9 > 0$ for $b \ge B(L_{\varepsilon}, \eta)$. This completes the second step.

Step (III) (The tightness) Actually, it suffices to prove the following proposition.

Proposition 3.6 For any $\eta > 0$,

(3.45)
$$\lim_{\delta \to 0} \limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P}\Big(\sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta \sigma \sqrt{n} \Big| I_n \le a_n(z) \Big) = 0.$$

The first two steps allow us to obtain the following fact whether the distribution is lattice or non-lattice.

Fact 3.7 There exist constants $c_{13}, c_{14} \in (0, \infty)$ such that

(3.46)
$$c_{13} \le \liminf_{z \to \infty} \liminf_{n \to \infty} \frac{e^z}{z} \mathbf{P}(I_n \le a_n(z)) \le \limsup_{z \to \infty} \limsup_{n \to \infty} \frac{e^z}{z} \mathbf{P}(I_n \le a_n(z)) \le c_{14}.$$

Proof of Proposition 3.6. First, we observe that for any $M \ge 1$ and $\delta \in (0, 1/2)$,

$$\mathbf{P}\Big(\sup_{0\leq k\leq\delta n} |I_n(n-k)-I_n| \geq \delta\sigma\sqrt{n}, \ I_n\leq a_n(z)\Big) \\
\leq \mathbf{P}\Big(m_n^{(n)}\notin J_{z,0,L}^{1/2}(n), \ I_n\leq a_n(z)\Big) + \mathbf{P}\Big(I_n(n-\lfloor\delta n\rfloor)\geq M\sigma\sqrt{\delta n}, \ I_n\leq a_n(z)\Big) + \chi(\delta,z,n).$$

where $\chi(\delta, z, n) := \mathbf{P}\Big(m_n^{(n)} \in J_{z,0,L}^{1/2}(n), I_n(n - \lfloor \delta n \rfloor) \le M\sigma\sqrt{\delta n}, \sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta\sigma\sqrt{n}\Big).$

It follows from Lemma 3.2 that for any $\varepsilon > 0$, if $L \ge L_{\varepsilon}$, $n \ge 1$ and $z \ge 0$,

(3.47)
$$\mathbf{P}\Big(m_n^{(n)} \notin J_{z,0,L}^{1/2}(n), \ I_n \le a_n(z)\Big) \le (1 + \varepsilon(1+z))e^{-z}.$$

Then dividing each term of (3.47) by $\mathbf{P}(I_n \leq a_n(z))$ yields that

$$(3.48) \quad \mathbf{P}\Big(\sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta \sigma \sqrt{n} \Big| I_n \le a_n(z) \Big) \\ \le \frac{(1 + \varepsilon(1+z))e^{-z}}{\mathbf{P}(I_n \le a_n(z))} + \mathbf{P}\Big(I_n(n - \lfloor \delta n \rfloor) \ge M \sigma \sqrt{\delta n} \Big| I_n \le a_n(z) \Big) + \frac{\chi(\delta, z, n)}{\mathbf{P}(I_n \le a_n(z))}.$$

On the one hand, by Fact 3.7,

(3.49)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \frac{(1 + \varepsilon(1 + z))e^{-z}}{\mathbf{P}(I_n \le a_n(z))} \le \frac{\varepsilon}{c_{13}}$$

On the other hand, Steps (I) and (II) tell us that for any $1 > \delta > 0$ and $M \ge 1$,

(3.50)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P} \Big[I_n(n - \lfloor \delta n \rfloor) \ge M \sigma \sqrt{\delta n} \Big| I_n \le a_n(z) \Big] = \mathbf{P} [e_{1-\delta} \ge M \sqrt{\delta}],$$

which, by Chebyshev's inequality, is bounded by $\frac{\mathbf{E}[e_{1-\delta}]}{M\sqrt{\delta}} = \frac{4\sqrt{1-\delta}}{M\sqrt{2\pi}}$. Consequently,

(3.51)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P} \Big(\sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta \sigma \sqrt{n} \Big| I_n \le a_n(z) \Big)$$
$$\le \frac{\varepsilon}{c_{13}} + \frac{2}{M} + \limsup_{z \to \infty} \limsup_{n \to \infty} \frac{\chi(\delta, z, n)}{\mathbf{P}(I_n \le a_n(z))}.$$

Let us estimate $\chi(\delta, z, n)$. One sees that

$$\chi(\delta, z, n) \leq \mathbf{E} \bigg[\sum_{|u|=n} \mathbf{1}_{\{u \in J_{z,L}^{1/2}(n); \sup_{0 \le k \le \delta n} |V(u_{n-k}) - V(u)| \ge \eta \sigma \sqrt{n}; V(u_{n-\lfloor \delta n \rfloor}) \le M \sigma \sqrt{\delta n} \} \bigg].$$

By Lemma 2.4, it becomes that

$$\begin{split} \chi(\delta, z, n) &\leq \mathbf{E} \Big[e^{S_n}; S_n \leq a_n(z), \underline{S}_n \geq -z, \underline{S}_{[n/2,n]} \geq a_n(z+L), \\ S_{n-\lfloor \delta n \rfloor} \leq M \sigma \sqrt{\delta n}, \sup_{0 \leq k \leq \delta n} |S_{n-k} - S_n| \geq \eta \sigma \sqrt{n} \Big] \\ &\leq n^{3/2} e^{-z} \Upsilon(\delta, z, n), \end{split}$$

where $\Upsilon(\delta, z, n) := \mathbf{P} \Big(S_n \leq a_n(z), \ \underline{S}_n \geq -z, \ \underline{S}_{[n/2,n]} \geq a_n(z+L), \ S_{n-\lfloor\delta n\rfloor} \leq M\sigma\sqrt{\delta n},$ $\sup_{0 \leq k \leq \delta n} |S_{n-k} - S_n| \geq \eta\sigma\sqrt{n}, \ S_{n-\lfloor\delta n\rfloor} \leq M\sigma\sqrt{\delta n} \Big).$

Reversing time yields that

$$(3.52) \quad \Upsilon(\delta, z, n) \leq \mathbf{P}\Big(\underline{S}_n^- \geq -a_n(0), \, \underline{S}_{n/2}^- \geq -L, \, -S_n \in [-a_n(z), -a_n(z+L)], \\ \sup_{0 \leq k \leq \delta n} |-S_k| \geq \eta \sigma \sqrt{n}, -S_{\lfloor \delta n \rfloor} \leq M \sigma \sqrt{\delta n} - a_n(z+L)\Big).$$

Applying the Markov property at time $\lfloor \delta n \rfloor$, we obtain that

(3.53)
$$\Upsilon(\delta, z, n) = \mathbf{E}\Big[\Theta(-S_{\lfloor \delta n \rfloor}); \ \underline{S}_{\delta n}^{-} \ge -L, \sup_{0 \le k \le \delta n} |-S_k| \ge \eta \sigma \sqrt{n}\Big],$$

where $\Theta(x) := \mathbf{1}_{\{x \le M\sigma\sqrt{\delta n} - a_n(z+L)\}} \mathbf{P}_x \Big(\underline{S}^-_{(1/2-\delta)n} \ge -L, \underline{S}^-_{(1-\delta)n} \ge -a_n(0), -S_{n-\lfloor\delta n\rfloor} \in [-a_n(z), -a_n(z+L)] \Big)$. Reversing time again implies that

$$\Theta(x) \leq \mathbf{1}_{\{x \leq M\sigma\sqrt{\delta n}\}} \mathbf{P}\Big(\underline{S}_{(1-\delta)n} \geq -z - L, \\ \underline{S}_{[n/2,(1-\delta)n]} \geq a_n(z+2L), S_{n-\lfloor\delta n\rfloor} \in [x+a_n(z+L), x+a_n(z)]\Big).$$

By (2.10), $\Theta(x) \leq c_{15}(1+z+L)(1+L)(1+M\sigma\sqrt{\delta n}+2L)n^{-3/2}$. Plugging it into (3.53) and taking *n* large enough so that $1+2L < \eta\sigma\sqrt{\delta n}$, we get that

$$\Upsilon(\delta, z, n) \le c_{15}(1+z)(1+L)^2 n^{-3/2} (M+\eta) \sigma \sqrt{\delta n} \mathbf{E} \Big[\underline{S}_{\delta n}^- \ge -L, \sup_{0 \le k \le \delta n} |-S_k| \ge \eta \sigma \sqrt{n} \Big].$$

Recall that $\chi(\delta, z, n) \leq e^{-z} n^{3/2} \Upsilon(\delta, z, n)$. We check that

(3.54)
$$\chi(\delta, z, n) \leq c_{15}e^{-z}(1+z)(1+L)^2(M+\eta)\sigma$$

 $\times \mathbf{E}_L\Big[\sup_{0\leq k\leq\delta n}(-S_k)\geq \eta\sigma\sqrt{n}\Big|\underline{S}_{\delta n}\geq 0\Big]\Big(\sqrt{\delta n}\mathbf{P}_L\Big[\underline{S}_{\delta n}\geq 0\Big]\Big).$

On the one hand, by Theorem 1.1 of [7], $\mathbf{E}_L \left[\sup_{0 \le k \le \delta n} (-S_k) \ge \eta \sigma \sqrt{n} \middle| \underline{S}_{\delta n} \ge 0 \right]$ converges to $\mathbf{P}(\sup_{0 \le s \le 1} \mathcal{M}_s \ge \eta / \sqrt{\delta})$ as $n \to \infty$. On the other hand, (2.7) shows that $\sqrt{\delta n} \mathbf{P}_L \left[\underline{S}_{\delta n} \ge 0 \right]$ converges to $C_-R_-(L)$ as $n \to \infty$. Therefore,

$$\limsup_{n \to \infty} \chi(\delta, z, n) \leq c_{15} e^{-z} (1+z)(1+L)^2 (M+\eta) \sigma C_- R_-(L) \times \mathbf{P}(\sup_{0 \leq s \leq 1} \mathcal{M}_s \geq \eta/\sqrt{\delta}).$$

Going back to (3.51) and letting $z \to \infty$, we deduce from Fact 3.7 that

(3.55)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P} \Big(\sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta \sigma \sqrt{n} \Big| I_n \le a_n(z) \Big) \\ \le \frac{\varepsilon}{c_{13}} + \frac{2}{M} + \frac{c_{15}(1+L)^2(M+\eta)\sigma C_- R_-(L) \times \mathbf{P}(\sup_{0 \le s \le 1} \mathcal{M}_s \ge \eta/\sqrt{\delta})}{c_{13}}.$$

Notice that $\mathbf{P}(\sup_{0 \le s \le 1} \mathcal{M}_s \ge \eta/\sqrt{\delta})$ decreases to 0 as $\delta \downarrow 0$. Take $M \ge 2/\varepsilon$. We conclude that for any $0 < \varepsilon < c_{13}$,

(3.56)
$$\limsup_{\delta \to 0} \limsup_{z \to \infty} \limsup_{n \to \infty} \Pr\left(\sup_{0 \le k \le \delta n} |I_n(n-k) - I_n| \ge \eta \sigma \sqrt{n} \Big| I_n \le a_n(z)\right) \le \frac{\varepsilon}{c_{13}} + \varepsilon,$$

which completes the proof of Proposition 3.6. And Proposition 3.1 is thus proved. \Box

4 Proof of Theorem 1.1

Let us prove the main theorem now. It suffices to prove that for any continuous functional $F: D([0,1], \mathbb{R}) \to [0,1]$, we have

(4.1)
$$\lim_{n \to \infty} \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \right] - \mathbf{E} \left[F(e_s, \ 0 \le s \le 1) \right] \right| = 0.$$

Proof of (4.1). Define for $A \ge 0$,

(4.2)
$$\mathcal{Z}[A] := \{ u \in \mathbb{T} : V(u) \ge A > \max_{k < |u|} V(u_k) \}.$$

For any particle $u \in \mathcal{Z}[A]$, there is a subtree rooted at u. If $|u| \leq n$, let

$$I_n(u) := \min_{v \ge u, |v|=n} V(v).$$

Moreover, assume m_n^u is the particle uniformly chosen in the set $\{|v| = n : v \ge u, V(v) = I_n(u)\}$. Similarly, we write $[\![\emptyset, m_n^u]\!] := \{\emptyset =: m_0^u, m_1^u, \cdots, m_n^u\}$. The trajectory leading to m_n^u is denoted by $\{V(m_k^u); 0 \le k \le n\}$. Let ω_A be the particle uniformly chosen in $\{u \in \mathcal{Z}[A] : |u| \le n, I_n(u) = I_n\}$.

Let $\mathcal{Y}_A := \{\max_{u \in \mathcal{Z}[A]} |u| \leq M, \max_{u \in \mathcal{Z}[A]} V(u) \leq M\}$. Then for any $\varepsilon > 0$, there exist $M := M(A, \varepsilon)$ large enough such that $\mathbf{P}(\mathcal{Y}_A^c) \leq \varepsilon$. It follows that

$$(4.3) \quad \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \right] - \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2 \right] \right| \\ \le \quad \varepsilon + \mathbf{P}[|I_n - a_n(0)| \ge A/2].$$

We then check that for $n \ge M$,

(4.4)
$$\mathbf{E}\Big[F\Big(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le 1\Big); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big]$$
$$= \mathbf{E}\Big[\sum_{u \in \mathcal{Z}[A]} \mathbf{1}_{(u=\omega_A)} F\Big(\frac{V(m^u_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le 1\Big); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big].$$

Define another trajectory $\{\tilde{V}(m_k^u); 0 \le k \le n\}$ as follows.

(4.5)
$$\tilde{V}(m_k^u) := \begin{cases} V(u) & \text{if } k < |u|;\\ V(m_k^u) & \text{if } |u| \le k \le n \end{cases}$$

It follows that

(4.6)
$$\mathbf{E}\Big[F\Big(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le 1\Big); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big]$$
$$= \mathbf{E}\Big[\sum_{u \in \mathcal{Z}[A]} \mathbf{1}_{(u=\omega_A)} F\Big(\frac{\tilde{V}(m^u_{\lfloor sn \rfloor})}{\sigma\sqrt{n}}; 0 \le s \le 1\Big); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big] + o_n(1),$$

where $o_n(1) \to 0$ as n goes to infinity.

Define the sigma-field $\mathcal{G}_A := \sigma\{(u, V(u), I_n(u)); u \in \mathcal{Z}[A]\}$. Note that on \mathcal{Y}_A , $I_n = \min_{u \in \mathcal{Z}[A]} I_n(u)$ as long as $n \geq M$. One sees that $\mathcal{Y}_A \cap \{|I_n - a_n(0)| \leq A/2\}$ is \mathcal{G}_A -measurable for all n large enough. Thus,

$$(4.7) \quad \mathbf{E}\Big[\sum_{u\in\mathcal{Z}[A]}\mathbf{1}_{(u=\omega_A)}F\Big(\frac{\tilde{V}(m^u_{\lfloor sn\rfloor})}{\sigma\sqrt{n}}; 0\leq s\leq 1\Big); \mathcal{Y}_A, |I_n-a_n(0)|\leq A/2\Big] \\ = \mathbf{E}\Big[\sum_{u\in\mathcal{Z}[A]}\mathbf{1}_{(u=\omega_A)}\mathbf{E}\Big[F\Big(\frac{\tilde{V}(m^u_{\lfloor sn\rfloor})}{\sigma\sqrt{n}}; 0\leq s\leq 1\Big)\Big|\mathcal{G}_A, u=\omega_A\Big]; \mathcal{Y}_A, |I_n-a_n(0)|\leq A/2\Big].$$

Further, we notice by the branching property that conditioned on $\{(u, V(u)); u \in \mathbb{Z}[A]\}$, the subtrees generated by $u \in \mathbb{Z}[A]$ are independent copies of the original one, started from V(u), respectively. Therefore, given $\mathcal{Y}_A \cap \{|I_n - a_n(0)| \leq A/2\}$,

$$\mathbf{1}_{(u=\omega_A)} \mathbf{E} \left[F\left(\frac{V(m_{\lfloor sn \rfloor}^u)}{\sigma\sqrt{n}}; 0 \le s \le 1\right) \middle| \mathcal{G}_A, u = \omega_A \right] \\ = \mathbf{1}_{(u=\omega_A)} \mathbf{E} \left[F\left(\frac{I(\lfloor s(n-|u|) \rfloor)}{\sigma\sqrt{n-|u|}}; 0 \le s \le 1\right) \middle| I_{n-|u|} \le a_n(-r_u) \right] + o_n(1),$$

where $r_u := \min\{\min_{v \in \mathcal{Z}[A] \setminus \{u\}} I_n(v) - a_n(0), A/2\} - V(u)$ is independent of $I_{n-|u|}$. Thus, (4.6) becomes that

(4.8)
$$\mathbf{E}\Big[F\Big(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le 1\Big); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big]$$
$$= \mathbf{E}\Big[\sum_{u \in \mathcal{Z}[A]} \mathbf{1}_{(u=\omega_A)} \mathbf{E}\Big[F\Big(\frac{I(\lfloor s(n-|u|)\rfloor)}{\sigma\sqrt{n-|u|}}; 0 \le s \le 1\Big)\Big|I_{n-|u|} \le a_n(-r_u)\Big];$$
$$\mathcal{Y}_A, |I_n - a_n(0)| \le A/2\Big] + o_n(1).$$

The event $\mathcal{Y}_A \cap \{|I_n - a_n(0)| \le A/2\}$ ensures that $A/2 + M \ge -r_u \ge A/2$. The conditioned convergence has been given in Proposition 3.1. We need a slightly stronger version here.

According to Proposition 3.1, for any $\varepsilon > 0$, there exists $z_{\varepsilon} > 0$ such that for all $z \ge z_{\varepsilon}$,

(4.9)
$$\limsup_{n \to \infty} \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \right| I_n \le a_n(z) \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right| < \varepsilon.$$

Thus, for any $z \ge z_{\varepsilon}$, there exists $N_z \ge 1$ such that for any $n \ge N_z$,

(4.10)
$$\left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \middle| I_n \le a_n(z) \right] - \mathbf{E} \left[F(e_s, 0 \le s \le 1) \right] \right| < 2\varepsilon.$$

Take $A = 2z_{\varepsilon}$ and K = M. We say that for n sufficiently large,

(4.11)
$$\sup_{z \in [z_{\varepsilon}, z_{\varepsilon}+K]} \left| \mathbf{E} \left[F \left(\frac{I(\lfloor s(n) \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \middle| I_n \le a_n(z) \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right| \le 3\varepsilon.$$

In the lattice case, (4.11) follows immediately. We only need to prove it in the non-lattice case.

Recall that $\Sigma_n(F, z) = \mathbf{E} \Big[F\Big(\frac{I_n(\lfloor sn \rfloor)}{\sigma\sqrt{n}}; 0 \le s \le 1 \Big); \ I_n \le a_n(z) \Big]$ with $0 \le F \le 1$. Then, for any $\ell > 0$ and $z \ge 0$,

$$(4.12) \qquad \left| \frac{\Sigma_{n}(F,z)}{\Sigma_{n}(1,z)} - \frac{\Sigma_{n}(F,z+\ell)}{\Sigma_{n}(1,z+\ell)} \right| \\ \leq \left| \frac{\Sigma_{n}(F,z) - \Sigma_{n}(F,z+\ell)}{\Sigma_{n}(1,z)} \right| + \left| \frac{\Sigma_{n}(F,z+\ell)}{\Sigma_{n}(1,z)} - \frac{\Sigma_{n}(F,z+\ell)}{\Sigma_{n}(1,z+\ell)} \right| \\ = \frac{1}{\Sigma_{n}(1,z)} \left(\left| \Sigma_{n}(F,z) - \Sigma_{n}(F,z+\ell) \right| + \frac{\Sigma_{n}(F,z+\ell)}{\Sigma_{n}(1,z+\ell)} \right| \Sigma_{n}(1,z+\ell) - \Sigma_{n}(1,z) \right| \right).$$

Since $0 \le F \le 1$, the two following inequalities

$$\begin{aligned} \left| \Sigma_n(F,z) - \Sigma_n(F,z+\ell) \right| &= \mathbf{E} \Big[F\Big(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \Big); \ a_n(z+\ell) < I_n \le a_n(z) \Big] \\ &\le \mathbf{P}(a_n(z+\ell) < I_n \le a_n(z)), \end{aligned}$$

and $\frac{\Sigma_n(F,z+\ell)}{\Sigma_n(1,z+\ell)} \leq 1$ hold. Note also that $|\Sigma_n(1,z+\ell) - \Sigma_n(1,z)| = \mathbf{P}(a_n(z+\ell) < I_n \leq a_n(z))$. It follows that

(4.13)
$$\left|\frac{\Sigma_n(F,z)}{\Sigma_n(1,z)} - \frac{\Sigma_n(F,z+\ell)}{\Sigma_n(1,z+\ell)}\right| \leq 2\frac{\mathbf{P}(a_n(z+\ell) < I_n \leq a_n(z))}{\mathbf{P}(I_n \leq a_n(z))}$$
$$= 2 - 2\frac{\mathbf{P}(I_n \leq a_n(z+\ell))}{\mathbf{P}(I_n \leq a_n(z))}.$$

In view of Fact 3.4, we take $\frac{3}{2} \ln n - \Lambda_{\varepsilon'} \ge \ell + z > z \ge \Lambda_{\varepsilon'}$ so that for any $n \ge N_{\varepsilon'}$,

(4.14)
$$\frac{\mathbf{P}(I_n \le a_n(z+\ell))}{\mathbf{P}(I_n \le a_n(z))} \ge \frac{(C-\varepsilon')(z+\ell)e^{-z-\ell}}{(C+\varepsilon')ze^{-z}} \ge \frac{C-\varepsilon'}{C+\varepsilon'}e^{-\ell}.$$

For $\varepsilon' = C\varepsilon/8 > 0$, we choose $\zeta = \frac{\varepsilon}{4}$ so that $\frac{C-\varepsilon'}{C+\varepsilon'}e^{-\zeta} \ge 1 - \frac{\varepsilon}{2}$. As a consequence, for any $\Lambda_{\varepsilon'} \le z \le \frac{3}{2} \ln n - \Lambda_{\varepsilon'} - \zeta$, $0 \le \ell \le \zeta$ and $n \ge N_{\varepsilon'}$,

(4.15)
$$\left|\frac{\Sigma_n(F,z)}{\Sigma_n(1,z)} - \frac{\Sigma_n(F,z+\ell)}{\Sigma_n(1,z+\ell)}\right| \le 2\left(1 - \frac{C-\varepsilon'}{C+\varepsilon'}e^{-\ell}\right) \le \varepsilon.$$

For $\varepsilon > 0$, z_{ε} can be chosen so that $[z_{\varepsilon}, z_{\varepsilon} + K] \subset [\Lambda_{\varepsilon'}, \frac{3}{2} \ln n - \Lambda_{\varepsilon'}]$ for $n \ge e^K N_{\varepsilon'}$. For any integer $0 \le j \le \lceil K/\zeta \rceil$, let $z_j := z_{\varepsilon} + j\zeta$. Then $[z_{\varepsilon}, z_{\varepsilon} + K] \subset \bigcup_{0 \le j \le \lceil K/\zeta \rceil} [z_j, z_{j+1}]$. Take $N'_{\varepsilon} = \max_{0 \le j \le \lceil K/\zeta \rceil} \{N_{z_j}, e^K N_{\varepsilon'}\}$. By (4.10) and (4.15), we conclude that for any $n \ge N'_{\varepsilon}$,

$$\sup_{z \in [z_{\varepsilon}, z_{\varepsilon} + K]} \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \middle| I_n \le a_n(z) \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right|$$

$$\leq \sup_{0 \le j \le \lceil K/\zeta \rceil} \left| \frac{\Sigma_n(F, z_j)}{\Sigma_n(1, z_j)} - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right| + \sup_{0 \le j < \lceil K/\zeta \rceil} \sup_{z_j \le z \le z_{j+1}} \left| \frac{\Sigma_n(F, z)}{\Sigma_n(1, z)} - \frac{\Sigma_n(F, z_j)}{\Sigma_n(1, z_j)} \right|$$

$$\leq 3\varepsilon.$$

We continue to prove the main theorem. Since $\sum_{u \in \mathcal{Z}[A]} \mathbf{1}_{(u=\omega_A)} = 1$, we deduce from (4.8) and (4.11) that for *n* sufficiently large,

$$\left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right); \mathcal{Y}_A, |I_n - a_n(0)| \le A/2 \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right|$$

$$\le 3\varepsilon \mathbf{P}(\mathcal{Y}_A; |I_n - a_n(0)| \le A/2) + o_n(1) + \mathbf{P}(\mathcal{Y}_A^c) + \mathbf{P}(|I_n - a_n(0)| \ge A/2)$$

$$\le 4\varepsilon + o_n(1) + \mathbf{P}(|I_n - a_n(0)| \ge A/2).$$

Going back to (4.3), we conclude that for n large enough,

$$\left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right| \le 5\varepsilon + 2\mathbf{P}(|I_n - a_n(0)| \ge A/2) + o_n(1).$$

Let n go to infinity and then make $\varepsilon \downarrow 0$. Therefore,

(4.16)
$$\limsup_{n \to \infty} \left| \mathbf{E} \left[F \left(\frac{I_n(\lfloor sn \rfloor)}{\sigma \sqrt{n}}; 0 \le s \le 1 \right) \right] - \mathbf{E} [F(e_s, 0 \le s \le 1)] \right|$$
$$\leq \limsup_{z \to \infty} \limsup_{n \to \infty} 2\mathbf{P}(|I_n - a_n(0)| \ge z).$$

It remains to show that $\limsup_{z\to\infty} \limsup_{n\to\infty} \mathbf{P}(|I_n - a_n(0)| \ge z) = 0$. Because of Fact (3.7), it suffices to prove that

(4.17)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P}(I_n \ge a_n(0) + z) = 0.$$

In the non-lattice case, Theorem 1.1 of [2] implies it directly. In the lattice case, we see that for n large enough,

(4.18)
$$\mathbf{P}(I_n \ge a_n(0) + z) \le \mathbf{E}\Big[\prod_{u \in \mathcal{Z}[A]} (1 - \Phi_u(z, n)); \mathcal{Y}_A\Big] + \varepsilon,$$

with $\Phi_u(z, n) := \mathbf{P}(I_{n-|u|} \le a_n(V(u) - z))$. Take A = 2z here. Then it follows from Fact 3.7 that for n large enough and for any particle $u \in \mathcal{Z}[A]$,

(4.19)
$$\Phi_u(z,n) \ge c_{13}/2(V(u)-z)e^{z-V(u)} \ge \frac{c_{13}}{4}V(u)e^{z-V(u)}.$$

(4.18) hence becomes that

$$\limsup_{n \to \infty} \mathbf{P}(I_n \ge a_n(0) + z) \le \mathbf{E} \Big[\prod_{u \in \mathcal{Z}[A]} (1 - \frac{c_{13}}{4} V(u) e^{z - V(u)}); \mathcal{Y}_A \Big] + \varepsilon$$
$$\le \mathbf{E} \Big[\exp \Big(- \frac{c_{13}}{4} e^z \sum_{u \in \mathcal{Z}[A]} V(u) e^{-V(u)} \Big) \Big] + \varepsilon.$$

It has been proved that as A goes to infinity, $\sum_{u \in \mathcal{Z}[A]} V(u) e^{-V(u)}$ converges almost surely to some limit D_{∞} , which is strictly positive on the set of non-extinction of \mathbb{T} , (see (5.2) in [2]). We end up with

(4.20)
$$\limsup_{z \to \infty} \limsup_{n \to \infty} \mathbf{P}(I_n \ge a_n(0) + z) \le \varepsilon,$$

which completes the proof of Theorem 1.1. \Box

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References

- Addario-Berry, L. and Reed, B. (2009). Minima in branching random walks. Ann. Probab. 37, 1044–1079.
- [2] Aïdékon, E. (2011+). Weak convergence of the minimum of a branching random walk. ArXiv:1101.1810
- [3] Aïdékon, E., Berestycki, J., Brunet, É. and Shi, Z. (2011) The branching Brownian motion seen from its tip. Arxiv:1104.3738v2
- [4] Aïdékon, E. and Shi, Z. (2010+). The Seneta-Heyde scaling for the branching random walk. (preprint)

- [5] Biggins, J.D. and Kyprianou, A.E. (2005). Fixed points of the smoothing transform: the boundary case. *Electron. J. Probab.* 10, Paper no. 17, 609–631.
- [6] Caravenna, F. (2005). A local limit theorem for random walks conditioned to stay positive. Probab. Theory Related Fields 133, 508–530.
- [7] Caravenna, F. and Chaumont, L. (2008) Invariance principles for random walks conditioned to stay positive. Annales de l'Institut Henri Poincaré - Probabilités et Statistiques 44, No. 1, 170–190.
- [8] Chauvin, B. and Rouault, A. (1988). KPP equation and supercritical branching Brownian motion in the subcritical speed area. Application to spatial trees. *Probab. Theory Related Fields* 80, 299–314.
- [9] Durrett, R. Iglehart, D. and Miller, D. (1977). Weak convergence to Brownian meander and Brownian excursion. Ann. Probab. 5, No. 1, 117–129.
- [10] Feller, W. (1971). An Introduction to Probability Theory and Its Applications II, 2nd ed. Wiley, New York.
- [11] Hu, Y. and Shi, Z. (2009). Minimal position and critical martingale convergence in branching random walks, and directed polymers on disordered trees. Ann. Probab. 37, 742–789.
- [12] Imhof, J.-P. (1984) Density factorizations for Brownian motion, meander and the threedimensional Bessel process, and applications. J. Appl. Probab. 21, 500-510.
- [13] Kozlov, M.V. (1976). The asymptotic behavior of the probability of non-extinction of critical branching processes in a random environment. *Theory Probab. Appl.* 21, 791– 804.
- [14] Kyprianou, A. (2004). Travelling wave solutions to the K-P-P equation: alternatives to Simon Harris' probabilistic analysis. Ann. Inst. H. Poincaré Probab. Statist. 40, 53–72.
- [15] Lyons, R. (1997). A simple path to Biggins' martingale convergence for branching random walk. In: *Classical and Modern Branching Processes* (Eds.: K.B. Athreya and P. Jagers). *IMA Volumes in Mathematics and its Applications* 84, 217–221. Springer, New York.
- [16] Revuz, D. and Yor, M. (2005) Continuous Martingales and Brownian Motion. 3rd ed, Springer-Verlag.