

QUENCHED LIMITS FOR TRANSIENT, ZERO SPEED ONE-DIMENSIONAL RANDOM WALK IN RANDOM ENVIRONMENT

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We consider a nearest-neighbor, one dimensional random walk $\{X_n\}_{n \geq 0}$ in a random i.i.d. environment, in the regime where the walk is transient but with zero speed, so that X_n is of order n^s for some $s < 1$. Under the quenched law (i.e., conditioned on the environment), we show that no limit laws are possible: There exist sequences $\{n_k\}$ and $\{x_k\}$ depending on the environment only, such that $X_{n_k} - x_k = o(\log n_k)^2$ (a *localized regime*). On the other hand, there exist sequences $\{t_m\}$ and $\{s_m\}$ depending on the environment only, such that $\log s_m / \log t_m \rightarrow s < 1$ and $P_\omega(X_{t_m}/s_m \leq x) \rightarrow 1/2$ for all $x > 0$ and $\rightarrow 0$ for $x \leq 0$ (a *spread out regime*).

1. Introduction and statement of main results. Let $\Omega = [0, 1]^{\mathbb{Z}}$, and let \mathcal{F} be the Borel σ -algebra on Ω . A random environment is an Ω -valued random variable $\omega = \{\omega_i\}_{i \in \mathbb{Z}}$ with distribution P . In this paper we will assume that the ω_i are i.i.d.

The *quenched* law P_ω^x for a random walk X_n in the environment ω is defined by

$$P_\omega^x(X_0 = x) = 1 \quad \text{and} \quad P_\omega^x(X_{n+1} = j | X_n = i) = \begin{cases} \omega_i, & \text{if } j = i + 1, \\ 1 - \omega_i, & \text{if } j = i - 1. \end{cases}$$

$\mathbb{Z}^{\mathbb{N}}$ is the space for the paths of the random walk $\{X_n\}_{n \in \mathbb{N}}$, and \mathcal{G} denotes the σ -algebra generated by the cylinder sets. Note that for each $\omega \in \Omega$, P_ω is a probability measure on \mathcal{G} , and for each $G \in \mathcal{G}$, $P_\omega^x(G) : (\Omega, \mathcal{F}) \rightarrow [0, 1]$ is a measurable function of ω . Expectations under the law P_ω^x are denoted E_ω^x .

The *annealed* law for the random walk in random environment X_n is defined by

$$\mathbb{P}^x(F \times G) = \int_F P_\omega^x(G) P(d\omega), \quad F \in \mathcal{F}, G \in \mathcal{G}.$$

For ease of notation, we will use P_ω and \mathbb{P} in place of P_ω^0 and \mathbb{P}^0 , respectively. We will also use \mathbb{P}^x to refer to the marginal on the space of paths, that is, $\mathbb{P}^x(G) = \mathbb{P}^x(\Omega \times G) = E_P[P_\omega^x(G)]$ for $G \in \mathcal{G}$. Expectations under the law \mathbb{P} will be written \mathbb{E} .

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A simple criterion for recurrence and a formula for the speed of transience was given by Solomon in [13]. For any integers $i \leq j$, define

$$(1) \quad \rho_i := \frac{1 - \omega_i}{\omega_i} \quad \text{and} \quad \Pi_{i,j} := \prod_{k=i}^j \rho_k$$

and for $x \in \mathbb{Z}$, define the hitting times

$$T_x := \min\{n \geq 0 : X_n = x\}.$$

Then X_n is transient to the right (resp. to the left) if $E_P(\log \rho_0) < 0$ (resp. $E_P \log \rho_0 > 0$) and recurrent if $E_P(\log \rho_0) = 0$ (henceforth, we will write ρ instead of ρ_0 in expectations involving only ρ_0). In the case where $E_P \log \rho < 0$ (transience to the right), Solomon established the following law of large numbers:

$$v_P := \lim_{n \rightarrow \infty} \frac{X_n}{n} = \lim_{n \rightarrow \infty} \frac{n}{T_n} = \frac{1}{\mathbb{E}T_1}, \quad \mathbb{P}\text{-a.s.}$$

For any integers $i < j$, define

$$(2) \quad W_{i,j} := \sum_{k=i}^j \Pi_{k,j} \quad \text{and} \quad W_j := \sum_{k \leq j} \Pi_{k,j}.$$

When $E_P \log \rho < 0$, it was shown in [13] and [14] (remark following Lemma 2.1.12), that

$$(3) \quad E_\omega^j T_{j+1} = 1 + 2W_j < \infty, \quad P\text{-a.s.},$$

and thus $v_P = 1/(1 + 2E_P W_0)$. Since P is a product measure, $E_P W_0 = \sum_{k=1}^\infty (E_P \rho)^k$. In particular, $v_P = 0$ if $E_P \rho \geq 1$.

Kesten, Kozlov and Spitzer [8] determined the annealed limiting distribution of a RWRE with $E_P \log \rho < 0$, that is, transient to the right. They derived the limiting distributions for the walk by first establishing a stable limit law of index s for T_n , where s is defined by the equation

$$E_P \rho^s = 1.$$

In particular, they showed that when $s < 1$, there exists a $b > 0$ such that

$$\lim_{n \rightarrow \infty} \mathbb{P}\left(\frac{T_n}{n^{1/s}} \leq x\right) = L_{s,b}(x),$$

and

$$(4) \quad \lim_{n \rightarrow \infty} \mathbb{P}\left(\frac{X_n}{n^s} \leq x\right) = 1 - L_{s,b}(x^{-1/s}),$$

where $L_{s,b}$ is the distribution function for a stable random variable with characteristic function

$$(5) \quad \hat{L}_{s,b}(t) = \exp\left\{-b|t|^s \left(1 - i \frac{t}{|t|} \tan(\pi s/2)\right)\right\}.$$

The value of b was recently identified [2]. While the annealed limiting distributions for transient one-dimensional RWRE have been known for quite a while, the corresponding quenched limiting distributions have remained largely unstudied until recently. Goldsheid [5] and Peterson [11] independently proved that when $s > 2$, a quenched CLT holds with a random (depending on the environment) centering. A similar result was given by Rassoul-Agha and Seppäläinen in [12] under different assumptions on the environment. Previously, in [10] and [14], it was shown that the limiting statement for the quenched CLT with random centering holds in probability rather than almost surely. No other results of quenched limiting distributions are known when $s \leq 2$.

In this paper, we analyze the quenched limiting distributions of a one-dimensional transient RWRE in the case $s < 1$. One could expect that the quenched limiting distributions are of the same type as the annealed limiting distributions since annealed probabilities are averages of quenched probabilities. However, this turns out not to be the case. In fact, a consequence of our main results, Theorems 1.1, 1.2 and 1.3 below is that the annealed stable behavior of T_n comes from fluctuations in the environment.

Throughout the paper, we will make the following assumptions.

ASSUMPTION 1. P is an i.i.d. product measure on Ω such that

$$(6) \quad E_P \log \rho < 0 \quad \text{and} \quad E_P \rho^s = 1 \quad \text{for some } s > 0.$$

ASSUMPTION 2. The distribution of $\log \rho$ is nonlattice under P and $E_P \rho^s \times \log \rho < \infty$.

NOTE. Since $E_P \rho^\gamma$ is a convex function of γ , the two statements in (6) imply that $E_P \rho^\gamma < 1$ for all $\gamma < s$ and $E_P \rho^\gamma > 1$ for all $\gamma > s$. Assumption 1 contains the essential assumption necessary for the walk to be transient. The main results of this paper are for $s < 1$ (the zero-speed regime), but many statements hold for $s \in (0, 2)$ or even $s \in (0, \infty)$. If no mention is made of bounds on s , then it is assumed that the statement holds for all $s > 0$. We recall that the technical conditions contained in Assumption 2 were also invoked in [8].

Define the “ladder locations” v_i of the environment by

$$(7) \quad v_0 = 0 \quad \text{and} \quad v_i = \begin{cases} \inf\{n > v_{i-1} : \Pi_{v_{i-1}, n-1} < 1\}, & i \geq 1, \\ \sup\{j < v_{i+1} : \Pi_{k, j-1} < 1, \forall k < j\}, & i \leq -1. \end{cases}$$

Throughout the remainder of the paper, we will let $v = v_1$. We will sometimes refer to sections of the environment between v_{i-1} and $v_i - 1$ as “blocks” of the environment. Note that the block between v_{-1} and $v_0 - 1$ is different from all the other blocks between consecutive ladder locations. Define the measure Q on environments by $Q(\cdot) := P(\cdot | \mathcal{R})$, where the event

$$\mathcal{R} := \{\omega \in \Omega : \Pi_{-k, -1} < 1, \forall k \geq 1\}.$$

Note that $P(\mathcal{R}) > 0$ since $E_P \log \rho < 0$. Q is defined so that the blocks of the environment between ladder locations are i.i.d. under Q , all with distribution the same as that of the block from 0 to $\nu - 1$ under P . In Section 3, we prove the following annealed theorem.

THEOREM 1.1. *Let Assumptions 1 and 2 hold, and let $s < 1$. Then there exists a $b' > 0$ such that*

$$\lim_{n \rightarrow \infty} Q\left(\frac{E_\omega T_{v_n}}{n^{1/s}} \leq x\right) = L_{s,b'}(x).$$

We then use Theorem 1.1 to prove the following two theorems which show that P -a.s. there exist two different random sequences of times (depending on the environment) where the random walk has different limiting behavior. These are the main results of the paper.

THEOREM 1.2. *Let Assumptions 1 and 2 hold, and let $s < 1$. Then P -a.s. there exist random subsequences $t_m = t_m(\omega)$ and $u_m = u_m(\omega)$, such that for any $\delta > 0$,*

$$\lim_{m \rightarrow \infty} P_\omega\left(\frac{X_{t_m} - u_m}{(\log t_m)^2} \in [-\delta, \delta]\right) = 1.$$

THEOREM 1.3. *Let Assumptions 1 and 2 hold, and let $s < 1$. Then P -a.s. there exists a random subsequence $n_{k_m} = n_{k_m}(\omega)$ of $n_k = 2^{2^k}$ and a random sequence $t_m = t_m(\omega)$, such that*

$$\lim_{m \rightarrow \infty} \frac{\log t_m}{\log n_{k_m}} = \frac{1}{s},$$

and

$$\lim_{m \rightarrow \infty} P_\omega\left(\frac{X_{t_m}}{n_{k_m}} \leq x\right) = \begin{cases} 0, & \text{if } x \leq 0, \\ \frac{1}{2}, & \text{if } 0 < x < \infty. \end{cases}$$

Note that Theorems 1.2 and 1.3 preclude the possibility of a quenched analogue of the annealed statement (4). It should be noted that in [4], Gantert and Shi prove that when $s \leq 1$, there exists a random sequence of times t_m at which the local time of the random walk at a single site is a positive fraction of t_m . This is related to the statement of Theorem 1.2, but we do not see a simple argument which directly implies Theorem 1.2 from the results of [4].

As in [8], limiting distributions for X_n arise from first studying limiting distributions for T_n . Thus, to prove Theorem 1.3, we first prove that there exists random subsequences $x_m = x_m(\omega)$ and $v_{m,\omega}$ in which

$$\lim_{m \rightarrow \infty} P_\omega\left(\frac{T_{x_m} - E_\omega T_{x_m}}{\sqrt{v_{m,\omega}}} \leq y\right) = \int_{-\infty}^y \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt =: \Phi(y).$$

We actually prove a stronger statement than this in Theorem 5.10 below, where we prove that all x_m “near” a subsequence n_{k_m} of $n_k = 2^{2^k}$ have the same Gaussian behavior (what we mean by “near” the subsequence n_{k_m} is made precise in the statement of the theorem).

The structure of the paper is as follows. In Section 2, we prove some introductory lemmas which will be used throughout the paper. Section 3 is devoted to proving Theorem 1.1. In Section 4, we use the latter to prove Theorem 1.2. In Section 5, we prove the existence of random subsequences $\{n_k\}$ where T_{n_k} is approximately Gaussian, and use this fact to prove Theorem 1.3. Section 6 contains the proof of the following technical theorem which is used throughout the paper.

THEOREM 1.4. *Let Assumptions 1 and 2 hold. Then there exists a constant $K_\infty \in (0, \infty)$ such that*

$$Q(E_\omega T_\nu > x) \sim K_\infty x^{-s}.$$

The proof of Theorem 1.4 is based on results from [7] and mimics the proof of tail asymptotics in [8].

2. Introductory lemmas. Before proceeding with the proofs of the main theorems, we mention a few easy lemmas which will be used throughout the rest of the paper. Recall the definitions of $\Pi_{1,k}$ and W_i in (1) and (2).

LEMMA 2.1. *For any $c < -E_P \log \rho$, there exist $\delta_c, A_c > 0$ such that*

$$(8) \quad P(\Pi_{1,k} > e^{-ck}) = P\left(\frac{1}{k} \sum_{i=1}^k \log \rho_i > -c\right) \leq A_c e^{-\delta_c k}.$$

Also, there exist constant $C_1, C_2 > 0$ such that $P(\nu > x) \leq C_1 e^{-C_2 x}$ for all $x \geq 0$.

PROOF. First, note that due to Assumption 1, $\log \rho$ has negative mean and finite exponential moments in a neighborhood of zero. If $c < -E_P \log \rho$, Cramér’s theorem ([1], Theorem 2.2.3) then yields (8). By the definition of ν , we have $P(\nu > x) \leq P(\Pi_{0, \lfloor x \rfloor - 1} \geq 1)$, which together with (8), completes the proof of the lemma. □

From [7], Theorem 5, there exist constants $K, K_1 > 0$ such that for all i

$$(9) \quad P(W_i > x) \sim Kx^{-s} \quad \text{and} \quad P(W_i > x) \leq K_1 x^{-s}.$$

The tails of W_{-1} , however, are different (under the measure Q), as the following lemma shows.

LEMMA 2.2. *There exist constants $C_3, C_4 > 0$ such that $Q(W_{-1} > x) \leq C_3 e^{-C_4 x}$ for all $x \geq 0$.*

PROOF. Since $\Pi_{i,-1} < 1$, Q -a.s. we have $W_{-1} < k + \sum_{i < -k} \Pi_{i,-1}$ for any $k > 0$. Also, note that from (8), we have $Q(\Pi_{-k,-1} > e^{-ck}) \leq A_c e^{-\delta_c k} / P(\mathcal{R})$. Thus,

$$\begin{aligned} Q(W_{-1} > x) &\leq Q\left(\frac{x}{2} + \sum_{k=x/2}^{\infty} e^{-ck} > x\right) + Q\left(\Pi_{-k,-1} > e^{-ck}, \text{ for some } k \geq \frac{x}{2}\right) \\ &\leq \mathbf{1}_{x/2+1/(1-e^{-c})>x} + \sum_{k=x/2}^{\infty} Q(\Pi_{-k,-1} > e^{-ck}) \\ &\leq \mathbf{1}_{1/(1-e^{-c})>x/2} + \mathcal{O}(e^{-\delta_c x/2}). \quad \square \end{aligned}$$

We also need a few more definitions that will be used throughout the paper. For any $i \leq k$,

$$(10) \quad R_{i,k} := \sum_{j=i}^k \Pi_{i,j} \quad \text{and} \quad R_i := \sum_{j=i}^{\infty} \Pi_{i,j}.$$

Note that since P is a product measure, $R_{i,k}$ and R_i have the same distributions as $W_{i,k}$ and W_i respectively. In particular with K, K_1 , the same as in (9),

$$(11) \quad P(R_i > x) \sim Kx^{-s} \quad \text{and} \quad P(R_i > x) \leq K_1x^{-s}.$$

3. Stable behavior of expected crossing time. Recall from Theorem 1.4 that there exists $K_\infty > 0$ such that $Q(E_\omega T_\nu > x) \sim K_\infty x^{-s}$. Thus, $E_\omega T_\nu$ is in the domain of attraction of a stable distribution. Also, from the comments after the definition of Q in the Introduction, it is evident that under Q , the environment ω is stationary under shifts of the ladder times ν_i . Thus, under Q , $\{E_\omega^{\nu_i-1} T_{\nu_i}\}_{i \in \mathbb{Z}}$ is a stationary sequence of random variables. Therefore, it is reasonable to expect that $n^{-1/s} E_\omega T_{\nu_n} = n^{-1/s} \sum_{i=1}^n E_\omega^{\nu_i-1} T_{\nu_i}$ converge in distribution to a stable distribution of index s . The main obstacle to proving this is that the random variables $E_\omega^{\nu_i-1} T_{\nu_i}$ are not independent. This dependence, however, is rather weak. The strategy of the proof of Theorem 1.1 is to first show that we need only consider the blocks where the expected crossing time $E_\omega^{\nu_i-1} T_{\nu_i}$ is relatively large. These blocks will then be separated enough to make the expected crossing times essentially independent.

For every $k \in \mathbb{Z}$, define

$$(12) \quad M_k := \max\{\Pi_{\nu_{k-1},j} : \nu_{k-1} \leq j < \nu_k\}.$$

Theorem 1 in [6] gives that there exists a constant $C_5 > 0$ such that

$$(13) \quad Q(M_1 > x) \sim C_5 x^{-s}.$$

Thus, M_1 and $E_\omega T_\nu$ have similar tails under Q . We will now show that $E_\omega T_\nu$ cannot be too much larger than M_1 . From (3), we have that

$$(14) \quad E_\omega T_\nu = \nu + 2 \sum_{j=0}^{\nu-1} W_j = \nu + 2W_{-1}R_{0,\nu-1} + 2 \sum_{i=0}^{\nu-1} R_{i,\nu-1}.$$

From the definitions of ν and M_1 , we have that $R_{i,\nu-1} \leq (\nu-i)M_1 \leq \nu M_1$ for any $0 \leq i < \nu$. Therefore, $E_\omega T_\nu \leq \nu + 2W_{-1}\nu M_1 + 2\nu^2 M_1$. Thus, given any $0 < \alpha < \beta$ and $\delta > 0$, we have

$$(15) \quad \begin{aligned} Q(E_\omega T_\nu > \delta n^\beta, M_1 \leq n^\alpha) &\leq Q(\nu + 2W_{-1}\nu n^\alpha + 2\nu^2 n^\alpha > \delta n^\beta) \\ &\leq Q(W_{-1} > n^{(\beta-\alpha)/2}) + Q(\nu^2 > n^{(\beta-\alpha)/2}) \\ &= o(e^{-n^{(\beta-\alpha)/5}}), \end{aligned}$$

where the second inequality holds for all n large enough and the last equality is a result of Lemmas 2.1 and 2.2. We now show that only the ladder times with $M_k > n^{(1-\varepsilon)/s}$ contribute to the limiting distribution of $n^{-1/s} E_\omega T_{\nu_n}$.

LEMMA 3.1. *Assume $s < 1$. Then for any $\varepsilon > 0$ and any $\delta > 0$, there exists an $\eta > 0$ such that*

$$\lim_{n \rightarrow \infty} Q\left(\sum_{i=1}^n (E_\omega^{v_{i-1}} T_{v_i}) \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} > \delta n^{1/s}\right) = o(n^{-\eta}).$$

PROOF. First note that

$$\begin{aligned} &Q\left(\sum_{i=1}^n (E_\omega^{v_{i-1}} T_{v_i}) \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} > \delta n^{1/s}\right) \\ &\leq Q\left(\sum_{i=1}^n (E_\omega^{v_{i-1}} T_{v_i}) \mathbf{1}_{E_\omega^{v_{i-1}} T_{v_i} \leq n^{(1-\varepsilon/2)/s}} > \delta n^{1/s}\right) \\ &\quad + nQ(E_\omega T_\nu > n^{(1-\varepsilon/2)/s}, M_1 \leq n^{(1-\varepsilon)/s}). \end{aligned}$$

By (15), the last term above decreases faster than any power of n . Thus, it is enough to prove that for any $\delta, \varepsilon > 0$, there exists an $\eta > 0$ such that

$$Q\left(\sum_{i=1}^n (E_\omega^{v_{i-1}} T_{v_i}) \mathbf{1}_{E_\omega^{v_{i-1}} T_{v_i} \leq n^{(1-\varepsilon)/s}} > \delta n^{1/s}\right) = o(n^{-\eta}).$$

Next, pick $C \in (1, \frac{1}{s})$ and let $J_{C,\varepsilon,k,n} := \{i \leq n : n^{(1-C^k\varepsilon)/s} < E_\omega^{v_{i-1}} T_{v_i} \leq n^{(1-C^{k-1}\varepsilon)/s}\}$. Let $k_0 = k_0(C, \varepsilon)$ be the smallest integer such that $(1 - C^k\varepsilon) \leq 0$. Then for any $k < k_0$, we have

$$\begin{aligned} Q\left(\sum_{i \in J_{C,\varepsilon,k,n}} E_\omega^{v_{i-1}} T_{v_i} > \delta n^{1/s}\right) &\leq Q(\#J_{C,\varepsilon,k,n} > \delta n^{1/s - (1-C^{k-1}\varepsilon)/s}) \\ &\leq \frac{nQ(E_\omega T_\nu > n^{(1-C^k\varepsilon)/s})}{\delta n^{C^{k-1}\varepsilon/s}} \sim \frac{K_\infty}{\delta} n^{-C^{k-1}\varepsilon(1/s-C)}, \end{aligned}$$

where the asymptotics in the last line above is from Theorem 1.4. Letting $\eta = \frac{\varepsilon}{2}(\frac{1}{s} - C)$, we have for any $k < k_0$ that

$$(16) \quad \mathcal{Q}\left(\sum_{i \in J_{C,\varepsilon,k,n}} E_\omega^{v_{i-1} T_{v_i}} > \delta n^{1/s}\right) = o(n^{-\eta}).$$

Finally, note that

$$(17) \quad \mathcal{Q}\left(\sum_{i=1}^n (E_\omega^{v_{i-1} T_{v_i}}) \mathbf{1}_{E_\omega^{v_{i-1} T_{v_i}} \leq n^{(1-Ck_0-1\varepsilon)/s}} \geq \delta n^{1/s}\right) \leq \mathbf{1}_{n^{1+(1-Ck_0-1\varepsilon)/s} \geq \delta n^{1/s}}.$$

However, since $C^{k_0}\varepsilon \geq 1 > Cs$, we have $C^{k_0-1}\varepsilon > s$, which implies that the right side of (17) vanishes for all n large enough. Therefore, combining (16) and (17), we have

$$\begin{aligned} & \mathcal{Q}\left(\sum_{i=1}^n (E_\omega^{v_{i-1} T_{v_i}}) \mathbf{1}_{E_\omega^{v_{i-1} T_{v_i}} \leq n^{(1-\varepsilon)/s}} > \delta n^{1/s}\right) \\ & \leq \sum_{k=1}^{k_0-1} \mathcal{Q}\left(\sum_{i \in J_{C,\varepsilon,k,n}} E_\omega^{v_{i-1} T_{v_i}} > \frac{\delta}{k_0} n^{1/s}\right) \\ & \quad + \mathcal{Q}\left(\sum_{i=1}^n (E_\omega^{v_{i-1} T_{v_i}}) \mathbf{1}_{E_\omega^{v_{i-1} T_{v_i}} \leq n^{(1-Ck_0-1\varepsilon)/s}} \geq \frac{\delta}{k_0} n^{1/s}\right) = o(n^{-\eta}). \quad \square \end{aligned}$$

In order to make the crossing times of the significant blocks essentially independent, we introduce some reflections to the RWRE. For $n = 1, 2, \dots$, define

$$(18) \quad b_n := \lfloor \log^2(n) \rfloor.$$

Let $\bar{X}_t^{(n)}$ be the random walk that is the same as X_t with the added condition that after reaching v_k the environment is modified by setting $\omega_{v_k-b_n} = 1$, that is, never allow the walk to backtrack more than $\log^2(n)$ ladder times. We couple $\bar{X}_t^{(n)}$ with the random walk X_t in such a way that $\bar{X}_t^{(n)} \geq X_t$ with equality holding until the first time t when the walk $\bar{X}_t^{(n)}$ reaches a modified environment location. Denote by $\bar{T}_x^{(n)}$ the corresponding hitting times for the walk $\bar{X}_t^{(n)}$. The following lemmas show that we can add reflections to the random walk without changing the expected crossing time by very much.

LEMMA 3.2. *There exist $B, \delta' > 0$ such that for any $x > 0$*

$$\mathcal{Q}(E_\omega T_v - E_\omega \bar{T}_v^{(n)} > x) \leq B(x^{-s} \vee 1)e^{-\delta' b_n}.$$

PROOF. First, note that for any n the formula for $E_\omega \bar{T}_v^{(n)}$ is the same as for $E_\omega T_v$ in (14) except with $\rho_{v-b_n} = 0$. Thus, $E_\omega T_v$ can be written as

$$(19) \quad E_\omega T_v = E_\omega \bar{T}_v^{(n)} + 2(1 + W_{v-b_n-1})\Pi_{v-b_n, -1}R_{0, v-1}.$$

Now, since $v-b_n \leq -b_n$, we have

$$\begin{aligned} Q(\Pi_{v-b_n, -1} > e^{-cb_n}) &\leq \sum_{k=b_n}^{\infty} Q(\Pi_{-k, -1} > e^{-ck}) \\ &\leq \sum_{k=b_n}^{\infty} \frac{1}{P(\mathcal{R})} P(\Pi_{-k, -1} > e^{-ck}). \end{aligned}$$

Applying (8), we have that for any $0 < c < -E_P \log \rho$, there exist $A', \delta_c > 0$ such that

$$Q(\Pi_{v-b_n, -1} > e^{-cb_n}) \leq A' e^{-\delta_c b_n}.$$

Therefore, for any $x > 0$,

$$\begin{aligned} (20) \quad Q(E_\omega T_v - E_\omega \bar{T}_v^{(n)} > x) &\leq Q(2(1 + W_{v-b_n-1})\Pi_{v-b_n, -1}R_{0, v-1} > x) \\ &\leq Q(2(1 + W_{v-b_n-1})R_{0, v-1} > x e^{cb_n}) + A' e^{-\delta_c b_n} \\ &= Q(2(1 + W_{-1})R_{0, v-1} > x e^{cb_n}) + A' e^{-\delta_c b_n}, \end{aligned}$$

where the equality in the second line is due to the fact that the blocks of the environment are i.i.d. under Q . Also, from (14) and Theorem 1.4, we have

$$(21) \quad Q(2(1 + W_{-1})R_{0, v-1} > x e^{cb_n}) \leq Q(E_\omega T_v > x e^{cb_n}) \sim K_\infty x^{-s} e^{-c s b_n}.$$

Combining (20) and (21) completes the proof. \square

LEMMA 3.3. *For any $x > 0$ and $\varepsilon > 0$, we have that*

$$(22) \quad \lim_{n \rightarrow \infty} n Q(E_\omega \bar{T}_v^{(n)} > x n^{1/s}, M_1 > n^{(1-\varepsilon)/s}) = K_\infty x^{-s}.$$

PROOF. Since adding reflections only decreases the crossing times, we can get an upper bound using Theorem 1.4, that is,

$$\begin{aligned} (23) \quad \limsup_{n \rightarrow \infty} n Q(E_\omega \bar{T}_v^{(n)} > x n^{1/s}, M_1 > n^{(1-\varepsilon)/s}) \\ \leq \limsup_{n \rightarrow \infty} n Q(E_\omega T_v > x n^{1/s}) = K_\infty x^{-s}. \end{aligned}$$

To get a lower bound, we first note that for any $\delta > 0$,

$$\begin{aligned} (24) \quad Q(E_\omega T_v > (1 + \delta)x n^{1/s}) &\leq Q(E_\omega \bar{T}_v^{(n)} > x n^{1/s}, M_1 > n^{(1-\varepsilon)/s}) \\ &\quad + Q(E_\omega T_v - E_\omega \bar{T}_v^{(n)} > \delta x n^{1/s}) \\ &\quad + Q(E_\omega T_v > (1 + \delta)x n^{1/s}, M_1 \leq n^{(1-\varepsilon)/s}) \\ &\leq Q(E_\omega \bar{T}_v^{(n)} > x n^{1/s}, M_1 > n^{(1-\varepsilon)/s}) + o(1/n), \end{aligned}$$

where the second inequality is from (15) and Lemma 3.2. Again, using Theorem 1.4, we have

$$\begin{aligned}
 (25) \quad & \liminf_{n \rightarrow \infty} n Q(E_\omega \bar{T}_v^{(n)} > xn^{1/s}, M_1 > n^{(1-\varepsilon)/s}) \\
 & \geq \liminf_{n \rightarrow \infty} n Q(E_\omega T_v > (1 + \delta)xn^{1/s}) - o(1) \\
 & = K_\infty (1 + \delta)^{-s} x^{-s}.
 \end{aligned}$$

Thus, by applying (23) and (25) and then letting $\delta \rightarrow 0$, we get (22). \square

Our general strategy is to show that the partial sums

$$\frac{1}{n^{1/s}} \sum_{k=1}^n E_\omega^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}$$

converge in distribution to a stable law of parameter s . To establish this, we will need bounds on the mixing properties of the sequence $E_\omega^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}$. As in [9], we say that an array $\{\xi_{n,k} : k \in \mathbb{Z}, n \in \mathbb{N}\}$ which is stationary in rows is α -mixing if $\lim_{k \rightarrow \infty} \limsup_{n \rightarrow \infty} \alpha_n(k) = 0$, where

$$\begin{aligned}
 \alpha_n(k) := \sup\{ & |P(A \cap B) - P(A)P(B)| : A \in \sigma(\dots, \xi_{n,-1}, \xi_{n,0}), \\
 & B \in \sigma(\xi_{n,k}, \xi_{n,k+1}, \dots)\}.
 \end{aligned}$$

LEMMA 3.4. *For any $0 < \varepsilon < \frac{1}{2}$, under the measure Q , the array of random variables $\{E_\omega^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}\}_{k \in \mathbb{Z}, n \in \mathbb{N}}$ is α -mixing with*

$$\sup_{k \in [1, \log^2 n]} \alpha_n(k) = o(n^{-1+2\varepsilon}), \quad \alpha_n(k) = 0 \quad \forall k > \log^2 n.$$

PROOF. Fix $\varepsilon \in (0, \frac{1}{2})$. For ease of notation, define $\xi_{n,k} := E_\omega^{v_{k-1}} \bar{T}_{v_k}^{(n)} \times \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}$. As we mentioned before, under Q the environment is stationary under shifts of the sequence of ladder locations and thus $\xi_{n,k}$ is stationary in rows under Q .

If $k > \log^2(n)$, then because of the reflections, $\sigma(\dots, \xi_{n,-1}, \xi_{n,0})$ and $\sigma(\xi_{n,k}, \xi_{n,k+1}, \dots)$ are independent and so $\alpha_n(k) = 0$. To handle the case when $k \leq \log^2(n)$, fix $A \in \sigma(\dots, \xi_{n,-1}, \xi_{n,0})$ and $B \in \sigma(\xi_{n,k}, \xi_{n,k+1}, \dots)$, and define the event

$$C_{n,\varepsilon} := \{M_j \leq n^{(1-\varepsilon)/s}, \text{ for } 1 \leq j \leq b_n\} = \{\xi_{n,j} = 0, \text{ for } 1 \leq j \leq b_n\}.$$

For any $j > b_n$, we have that $\xi_{n,j}$ only depends on the environment to the right of zero. Thus,

$$Q(A \cap B \cap C_{n,\varepsilon}) = Q(A)Q(B \cap C_{n,\varepsilon})$$

since $B \cap C_{n,\varepsilon} \in \sigma(\omega_0, \omega_1, \dots)$. Also, note that by (13) we have $Q(C_{n,\varepsilon}^c) \leq b_n Q(M_1 > n^{(1-\varepsilon)/s}) = o(n^{-1+2\varepsilon})$. Therefore,

$$\begin{aligned} |Q(A \cap B) - Q(A)Q(B)| &\leq |Q(A \cap B) - Q(A \cap B \cap C_{n,\varepsilon})| \\ &\quad + |Q(A \cap B \cap C_{n,\varepsilon}) - Q(A)Q(B \cap C_{n,\varepsilon})| \\ &\quad + Q(A)|Q(B \cap C_{n,\varepsilon}) - Q(B)| \\ &\leq 2Q(C_{n,\varepsilon}^c) = o(n^{-1+2\varepsilon}). \end{aligned} \quad \square$$

PROOF OF THEOREM 1.1. First, we show that the partial sums

$$\frac{1}{n^{1/s}} \sum_{k=1}^n E_{\omega}^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}$$

converge in distribution to a stable random variable of parameter s . To this end, we will apply [9], Theorem 5.1(III). We now verify the conditions of that theorem. The first condition that needs to be satisfied is

$$\lim_{n \rightarrow \infty} nQ(n^{-1/s} E_{\omega} \bar{T}_v^{(n)} \mathbf{1}_{M_1 > n^{(1-\varepsilon)/s}} > x) = K_{\infty} x^{-s}.$$

However, this is exactly the content of Lemma 3.3.

Secondly, we need a sequence m_n such that $m_n \rightarrow \infty$, $m_n = o(n)$ and $n\alpha_n(m_n) \rightarrow 0$, and such that for any $\delta > 0$,

$$(26) \quad \lim_{n \rightarrow \infty} \sum_{k=1}^{m_n} nQ(E_{\omega} \bar{T}_v^{(n)} \mathbf{1}_{M_1 > n^{(1-\varepsilon)/s}} > \delta n^{1/s}),$$

$$E_{\omega}^{v_k} \bar{T}_{v_{k+1}}^{(n)} \mathbf{1}_{M_{k+1} > n^{(1-\varepsilon)/s}} > \delta n^{1/s}) = 0.$$

However, by the independence of M_1 and M_{k+1} for any $k \geq 1$, the probability inside the sum is less than $Q(M_1 > n^{(1-\varepsilon)/s})^2$. By (13), this last expression is $\sim C_5 n^{-2+2\varepsilon}$. Thus, letting $m_n = n^{1/2-\varepsilon}$ yields (26). [Note that by Lemma 3.4, $n\alpha_n(m_n) = 0$ for all n large enough.]

Finally, we need to show that

$$(27) \quad \lim_{\delta \rightarrow 0} \limsup_{n \rightarrow \infty} nE_Q[n^{-1/s} E_{\omega} \bar{T}_v^{(n)} \mathbf{1}_{M_1 > n^{(1-\varepsilon)/s}} \mathbf{1}_{E_{\omega} \bar{T}_v^{(n)} \leq \delta}] = 0.$$

Now, by (23), there exists a constant $C_6 > 0$ such that for any $x > 0$,

$$Q(E_{\omega} \bar{T}_v^{(n)} > xn^{1/s}, M_1 > n^{(1-\varepsilon)/s}) \leq C_6 x^{-s} \frac{1}{n}.$$

Then using this, we have

$$\begin{aligned} &nE_Q[n^{-1/s} E_{\omega} \bar{T}_v^{(n)} \mathbf{1}_{M_1 > n^{(1-\varepsilon)/s}} \mathbf{1}_{E_{\omega} \bar{T}_v^{(n)} \leq \delta}] \\ &= n \int_0^{\delta} Q(E_{\omega} \bar{T}_v^{(n)} > xn^{1/s}, M_1 > n^{(1-\varepsilon)/s}) dx \\ &\leq C_6 \int_0^{\delta} x^{-s} dx = \frac{C_6 \delta^{1-s}}{1-s}, \end{aligned}$$

where the last integral is finite since $s < 1$. Equation (27) follows.

Having checked all its hypotheses, Kobus ([9], Theorem 5.1(III)) applies and yields that there exists a $b' > 0$ such that

$$(28) \quad Q\left(\frac{1}{n^{1/s}} \sum_{k=1}^n E_{\omega}^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}} \leq x\right) = L_{s,b'}(x),$$

where the characteristic function for the distribution $L_{s,b'}$ is given in (5). To get the limiting distribution of $\frac{1}{n^{1/s}} E_{\omega} T_{v_n}$, we use (19) and rewrite this as

$$(29) \quad \frac{1}{n^{1/s}} E_{\omega} T_{v_n} = \frac{1}{n^{1/s}} \sum_{k=1}^n E_{\omega}^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k > n^{(1-\varepsilon)/s}}$$

$$(30) \quad + \frac{1}{n^{1/s}} \sum_{k=1}^n E_{\omega}^{v_{k-1}} \bar{T}_{v_k}^{(n)} \mathbf{1}_{M_k \leq n^{(1-\varepsilon)/s}}$$

$$(31) \quad + \frac{1}{n^{1/s}} (E_{\omega} T_{v_n} - E_{\omega} \bar{T}_{v_n}^{(n)}).$$

Lemma 3.1 gives that (30) converges in distribution (under Q) to 0. Also, we can use Lemma 3.2 to show that (31) converges in distribution to 0 as well. Indeed, for any $\delta > 0$,

$$Q(E_{\omega} T_{v_n} - E_{\omega} \bar{T}_{v_n}^{(n)} > \delta n^{1/s}) \leq n Q(E_{\omega} T_v - E_{\omega} \bar{T}_v^{(n)} > \delta n^{1/s-1}) = \mathcal{O}(n^s e^{-\delta' b_n}).$$

Therefore, $n^{-1/s} E_{\omega} T_{v_n}$ has the same limiting distribution (under Q) as the right side of (29), which by (28) is an s -stable distribution with distribution function $L_{s,b'}$. \square

4. Localization along a subsequence. The goal of this section is to show when $s < 1$ that P -a.s. there exists a subsequence $t_m = t_m(\omega)$ of times such that the RWRE is essentially located in a section of the environment of length $\log^2(t_m)$. This will essentially be done by finding a ladder time whose crossing time is *much* larger than all the other ladder times before it. As a first step in this direction, we prove that with strictly positive probability this happens in the first n ladder locations. Recall the definition of M_k ; cf. (12).

LEMMA 4.1. *Assume $s < 1$. Then for any $C > 1$, we have*

$$\liminf_{n \rightarrow \infty} Q\left(\exists k \in [1, n/2]: M_k \geq C \sum_{j \in [1, n] \setminus \{k\}} E_{\omega}^{v_{j-1}} \bar{T}_{v_j}^{(n)}\right) > 0.$$

PROOF. Recall that $\bar{T}_x^{(n)}$ is the hitting time of x by the RWRE modified so that it never backtracks $b_n = \lfloor \log^2(n) \rfloor$ ladder locations.

To prove the lemma, first note that since $C > 1$ and $E_\omega^{v_{k-1}} \bar{T}_{v_k}^{(n)} \geq M_k$ there can only be at most one $k \leq n$ with $M_k \geq C \sum_{k \neq j \leq n} E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)}$. Therefore,

$$(32) \quad \begin{aligned} & Q\left(\exists k \in [1, n/2]: M_k \geq C \sum_{j \in [1, n] \setminus \{k\}} E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)}\right) \\ &= \sum_{k=1}^{n/2} Q\left(M_k \geq C \sum_{j \in [1, n] \setminus \{k\}} E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)}\right) \end{aligned}$$

Now, define the events

$$(33) \quad \begin{aligned} F_n &:= \{v_j - v_{j-1} \leq b_n, \forall j \in (-b_n, n]\}, \\ G_{k,n,\varepsilon} &:= \{M_j \leq n^{(1-\varepsilon)/s}, \forall j \in (k, k + b_n]\}. \end{aligned}$$

F_n and $G_{k,n,\varepsilon}$ are both *typical* events. Indeed, from Lemma 2.1, $Q(F_n^c) \leq (b_n + n)Q(v > b_n) = \mathcal{O}(ne^{-\tilde{C}_2 b_n})$, and from (13), we have $Q(G_{k,n,\varepsilon}^c) \leq b_n Q(M_1 > n^{(1-\varepsilon)/s}) = o(n^{-1+2\varepsilon})$. Now, from (3), adjusted for reflections, we have for any $j \in [1, n]$ that

$$\begin{aligned} & E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)} \\ &= (v_j - v_{j-1}) + 2 \sum_{l=v_{j-1}}^{v_j-1} W_{v_{j-1}-b_n, l} \\ &= (v_j - v_{j-1}) + 2 \sum_{v_{j-1} \leq i \leq l < v_j} \Pi_{i,l} + 2 \sum_{v_{j-1}-b_n < i < v_{j-1} \leq l < v_j} \Pi_{i, v_{j-1}-1} \Pi_{v_{j-1}, l} \\ &\leq (v_j - v_{j-1}) + 2(v_j - v_{j-1})^2 M_j + 2(v_j - v_{j-1})(v_{j-1} - v_{j-1}-b_n) M_j, \end{aligned}$$

where in the last inequality we used the facts that $\Pi_{v_{j-1}, i-1} \geq 1$ for $v_{j-1} < i < v_j$ and $\Pi_{i, v_{j-1}-1} < 1$ for all $i < v_{j-1}$. Then on the event $F_n \cap G_{k,n,\varepsilon}$, we have for $k+1 \leq j \leq k+b_n$ that

$$E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)} \leq b_n + 2b_n^2 n^{(1-\varepsilon)/s} + 2b_n^3 n^{(1-\varepsilon)/s} \leq 5b_n^3 n^{(1-\varepsilon)/s},$$

where for the first inequality we used that on the event $F_n \cap G_{k,n,\varepsilon}$ we have $v_j - v_{j-1} \leq b_n$ and $M_1 \leq n^{(1-\varepsilon)/s}$. Then using this, we get

$$\begin{aligned} & Q\left(M_k \geq C \sum_{j \in [1, n] \setminus \{k\}} E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(n)}\right) \\ &\geq Q(M_k \geq C(E_\omega \bar{T}_{v_{k-1}}^{(n)} + 5b_n^4 n^{(1-\varepsilon)/s} + E_\omega^{v_{k+b_n}} \bar{T}_{v_n}^{(n)}), F_n, G_{k,n,\varepsilon}) \\ &\geq Q(M_k \geq Cn^{1/s}, v_k - v_{k-1} \leq b_n) \\ &\quad \times Q(E_\omega \bar{T}_{v_{k-1}}^{(n)} + 5b_n^4 n^{(1-\varepsilon)/s} + E_\omega^{v_{k+b_n}} \bar{T}_{v_n}^{(n)} \leq n^{1/s}, \tilde{F}_n, G_{k,n,\varepsilon}), \end{aligned}$$

where $\tilde{F}_n := \{v_j - v_{j-1} \leq b_n, \quad \forall j \in (-b_n, n] \setminus \{k\}\} \supset F_n$. In the last inequality, we used the fact that $E_\omega^{v_{j-1}} \tilde{T}_{v_j}^{(n)}$ is independent of M_k for $j < k$ or $j > k + b_n$. Note that we can replace \tilde{F}_n by F_n in the last line above because it will only make the probability smaller. Then using the above and the fact that $E_\omega \tilde{T}_{v_{k-1}}^{(n)} + E_\omega^{v_k + b_n} \tilde{T}_{v_n}^{(n)} \leq E_\omega T_{v_n}$, we have

$$\begin{aligned} Q\left(M_k \geq C \sum_{j \in [1, n] \setminus \{k\}} E_\omega^{v_{j-1}} \tilde{T}_{v_j}^{(n)}\right) &\geq Q(M_k \geq C n^{1/s}, v_k - v_{k-1} \leq b_n) \\ &\quad \times Q(E_\omega T_{v_n} \leq n^{1/s} - 5b_n^4 n^{(1-\varepsilon)/s}, F_n, G_{k, n, \varepsilon}) \\ &\geq (Q(M_1 \geq C n^{1/s}) - Q(v > b_n)) \\ &\quad \times (Q(E_\omega T_{v_n} \leq n^{1/s} (1 - 5b_n^4 n^{-\varepsilon/s})) - Q(F_n^c) - Q(G_{k, n, \varepsilon}^c)) \\ &\sim C_5 C^{-s} L_{s, b'}(1) \frac{1}{n}, \end{aligned}$$

where the asymptotics in the last line are from (13) and Theorem 1.1. Combining the last display and (32) proves the lemma. \square

In Section 3, we showed that the proper scaling for $E_\omega T_{v_n}$ (or $E_\omega \tilde{T}_{v_n}^{(n)}$) was $n^{-1/s}$. The following lemma gives a bound on the moderate deviations under the measure P .

LEMMA 4.2. *Assume $s \leq 1$. Then for any $\delta > 0$,*

$$P(E_\omega T_{v_n} \geq n^{1/s+\delta}) = o(n^{-\delta s/2}).$$

PROOF. First, note that

$$(34) \quad P(E_\omega T_{v_n} \geq n^{1/s+\delta}) \leq P(E_\omega T_{2\bar{v}n} \geq n^{1/s+\delta}) + P(v_n \geq 2\bar{v}n),$$

where $\bar{v} := E_P v$. To handle the second term on the right side of (34) we note that v_n is the sum of n i.i.d. copies of v , and that v has exponential tails (by Lemma 2.1). Therefore, Cramér’s theorem ([1], Theorem 2.2.3) gives that $P(v_n/n \geq 2\bar{v}) = \mathcal{O}(e^{-\delta'n})$ for some $\delta' > 0$.

To handle the first term on the right side of (34), we note that for any $\gamma < s$ we have $E_P(E_\omega T_1)^\gamma < \infty$. This follows from the fact that $P(E_\omega T_1 > x) = P(1 + 2W_0 > x) \sim K 2^s x^{-s}$ by (3) and (9). Then by Chebyshev’s inequality and the fact that $\gamma < s \leq 1$, we have

$$(35) \quad P(E_\omega T_{2\bar{v}n} \geq n^{1/s+\delta}) \leq \frac{E_P(\sum_{k=1}^{2\bar{v}n} E_\omega^{k-1} T_k)^\gamma}{n^{\gamma(1/s+\delta)}} \leq \frac{2\bar{v}n E_P(E_\omega T_1)^\gamma}{n^{\gamma(1/s+\delta)}}.$$

Then choosing γ arbitrarily close to s , we can have that this last term is $o(n^{-\delta s/2})$. \square

Throughout the remainder of the paper, we will use the following subsequences of integers:

$$(36) \quad n_k := 2^{2^k}, \quad d_k := n_k - n_{k-1}$$

Note that $n_{k-1} = \sqrt{n_k}$ and so $d_k \sim n_k$ as $k \rightarrow \infty$.

COROLLARY 4.3. *For any k , define*

$$\mu_k := \max\{E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(d_k)} : n_{k-1} < j \leq n_k\}.$$

If $s < 1$, then

$$\lim_{k \rightarrow \infty} \frac{E_\omega^{v_{n_{k-1}}} \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k}{E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k} = 1, \quad P\text{-a.s.}$$

PROOF. Let $\varepsilon > 0$. Then

$$(37) \quad \begin{aligned} & P\left(\frac{E_\omega^{v_{n_{k-1}}} \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k}{E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k} \leq 1 - \varepsilon\right) \\ &= P\left(\frac{E_\omega \bar{T}_{v_{n_{k-1}}}^{(d_k)}}{E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k} \geq \varepsilon\right) \\ &\leq P(E_\omega \bar{T}_{v_{n_{k-1}}}^{(d_k)} \geq n_{k-1}^{1/s+\delta}) + P(E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k \leq \varepsilon^{-1} n_{k-1}^{1/s+\delta}). \end{aligned}$$

Lemma 4.2 gives that $P(E_\omega \bar{T}_{v_{n_{k-1}}}^{(d_k)} \geq n_{k-1}^{1/s+\delta}) \leq P(E_\omega T_{v_{n_{k-1}}} \geq n_{k-1}^{1/s+\delta}) = o(n_{k-1}^{-\delta s/2})$. To handle the second term in the right side of (37), note that if $\delta < \frac{1}{3s}$, then the subsequence n_k grows fast enough such that for all k large enough $n_k^{1/s-\delta} \geq \varepsilon^{-1} n_{k-1}^{1/s+\delta}$. Therefore, for k sufficiently large and $\delta < \frac{1}{3s}$, we have

$$P(E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k \leq \varepsilon^{-1} n_{k-1}^{1/s+\delta}) \leq P(E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k \leq n_k^{1/s-\delta}).$$

However, $E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k \leq n_k^{1/s-\delta}$ implies that $M_j < E_\omega^{v_{j-1}} \bar{T}_{v_j}^{(d_k)} \leq n_k^{1/s-\delta}$ for at least $n_k - 1$ of the $j \leq n_k$. Thus, since $P(M_1 > n_k^{1/s-\delta}) \sim C_5 n_k^{-1+\delta s}$, we have that

$$(38) \quad \begin{aligned} & P(E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k \leq \varepsilon^{-1} n_{k-1}^{1/s+\delta}) \leq n_k (1 - P(M_1 > n_k^{1/s-\delta}))^{n_k-1} \\ &= o(e^{-n_k^{\delta s/2}}). \end{aligned}$$

Therefore, for any $\varepsilon > 0$ and $\delta < \frac{1}{3s}$, we have that

$$P\left(\frac{E_\omega^{v_{n_{k-1}}} \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k}{E_\omega \bar{T}_{v_{n_k}}^{(d_k)} - \mu_k} \leq 1 - \varepsilon\right) = o(n_{k-1}^{-\delta s/2}).$$

By our choice of n_k , the sequence $n_{k-1}^{-\delta s/2}$ is summable in k . Applying the Borel–Cantelli lemma completes the proof. \square

COROLLARY 4.4. *Assume $s < 1$. Then P -a.s. there exists a random subsequence $j_m = j_m(\omega)$ such that*

$$M_{j_m} \geq m^2 E_\omega \bar{T}_{v_{j_m-1}}^{(j_m)}.$$

PROOF. Recall the definitions of n_k and d_k in (36). Then for any $C > 1$, define the event

$$D_{k,C} := \{\exists j \in (n_{k-1}, n_{k-1} + d_k/2] : M_j \geq C(E_\omega^{v_{n_{k-1}}} \bar{T}_{v_{j-1}}^{(d_k)} + E_\omega^{v_j} \bar{T}_{v_{n_k}}^{(d_k)})\}.$$

Note that due to the reflections, the event $D_{k,C}$ depends only on the environment from $v_{n_{k-1}-b_{d_k}}$ to $v_{n_k} - 1$. Then since $n_{k-1} - b_{d_k} > n_{k-2}$ for all $k \geq 4$, we have that the events $\{D_{2k,C}\}_{k=2}^\infty$ are all independent. Also, since the events do not involve the environment to the left of 0, they have the same probability under Q as under P . Then since Q is stationary under shifts of v_i , we have that for $k \geq 4$,

$$P(D_{k,C}) = Q(D_{k,C}) = Q(\exists j \in [1, d_k/2] : M_j \geq C(E_\omega \bar{T}_{v_{j-1}}^{(d_k)} + E_\omega^{v_j} \bar{T}_{v_{d_k}}^{(d_k)})).$$

Thus, for any $C > 1$, we have by Lemma 4.1 that $\liminf_{k \rightarrow \infty} P(D_{k,C}) > 0$. This combined with the fact that the events $\{D_{2k,C}\}_{k=2}^\infty$ are independent gives that for any $C > 1$ infinitely many of the events $D_{2k,C}$ occur P -a.s. Therefore, there exists a subsequence k_m of integers such that for each m , there exists $j_m \in (n_{k_m-1}, n_{k_m-1} + d_{k_m}/2]$ such that

$$M_{j_m} \geq 2m^2 (E_\omega^{v_{n_{k_m-1}}} \bar{T}_{v_{j_m-1}}^{(d_{k_m})} + E_\omega^{v_{j_m}} \bar{T}_{v_{n_{k_m}}}^{(d_{k_m})}) = 2m^2 (E_\omega^{v_{n_{k_m-1}}} \bar{T}_{v_{n_{k_m}}}^{(d_{k_m})} - \mu_{k_m}),$$

where the second equality holds due to our choice of j_m , which implies that $\mu_{k_m} = E_\omega^{v_{j_m-1}} \bar{T}_{v_{j_m}}^{(d_{k_m})}$. Then by Corollary 4.3, we have that for all m large enough

$$M_{j_m} \geq 2m^2 (E_\omega^{v_{k_m-1}} \bar{T}_{v_{n_{k_m}}}^{(d_{k_m})} - \mu_{k_m}) \geq m^2 (E_\omega \bar{T}_{v_{n_{k_m}}}^{(d_{k_m})} - \mu_{k_m}) \geq m^2 E_\omega \bar{T}_{v_{j_m-1}}^{(d_{k_m})},$$

where the last inequality is because $\mu_{k_m} = E_\omega^{v_{j_m-1}} \bar{T}_{v_{j_m}}^{(d_{k_m})}$. Now, for all k large enough, we have $n_{k-1} + d_k/2 < d_k$. Thus, we may assume (by possibly choosing a further subsequence) that $j_m < d_{k_m}$ as well, and since allowing less backtracking only decreases the crossing time we have

$$M_{j_m} \geq m^2 E_\omega \bar{T}_{v_{j_m-1}}^{(d_{k_m})} \geq m^2 E_\omega \bar{T}_{v_{j_m-1}}^{(j_m)}. \quad \square$$

The following lemma shows that the reflections that we have been using this whole time really do not affect the random walk. Recall the coupling of X_t and $\bar{X}_t^{(n)}$ introduced after (18).

LEMMA 4.5.

$$\lim_{n \rightarrow \infty} P_\omega(T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)}) = 0, \quad P\text{-a.s.}$$

PROOF. Let $\varepsilon > 0$. By Chebyshev's inequality,

$$P(P_\omega(T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)}) > \varepsilon) \leq \varepsilon^{-1} \mathbb{P}(T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)}).$$

Thus, by the Borel–Cantelli lemma, it is enough to prove that $\mathbb{P}(T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)})$ is summable. Now, the event $T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)}$ implies that there is an $i < \nu_{n-1}$ such that after reaching i for the first time, the random walk then backtracks a distance of b_n . Thus, again letting $\bar{\nu} = E_P \nu$, we have

$$\begin{aligned} \mathbb{P}(T_{\nu_{n-1}} \neq \bar{T}_{\nu_{n-1}}^{(n)}) &\leq P(\nu_{n-1} \geq 2\bar{\nu}(n-1)) + \sum_{i=0}^{2\bar{\nu}(n-1)} \mathbb{P}^i(T_{i-b_n} < \infty) \\ &= P(\nu_{n-1} \geq 2\bar{\nu}(n-1)) + 2\bar{\nu}(n-1)\mathbb{P}(T_{-b_n} < \infty). \end{aligned}$$

As noted in Lemma 4.2, $P(\nu_{n-1} \geq 2\bar{\nu}(n-1)) = \mathcal{O}(e^{-\delta'n})$, so we need only to show that $n\mathbb{P}(T_{-b_n} < \infty)$ is summable. However, [4], Lemma 3.3, gives that there exists a constant C_7 such that for any $k \geq 1$,

$$(39) \quad \mathbb{P}(T_{-k} < \infty) \leq e^{-C_7 k}.$$

Thus, $n\mathbb{P}(T_{-b_n} < \infty) \leq ne^{-C_7 b_n}$ which is summable by the definition of b_n . \square

We define the random variable $N_t := \max\{k : \exists n \leq t, X_n = \nu_k\}$ to be the maximum number of ladder locations crossed by the random walk by time t .

LEMMA 4.6.

$$\lim_{t \rightarrow \infty} \frac{\nu_{N_t} - X_t}{\log^2(t)} = 0, \quad \mathbb{P}\text{-a.s.}$$

PROOF. Let $\delta > 0$. If we can show that $\sum_{t=1}^{\infty} \mathbb{P}(|N_t - X_t| \geq \delta \log^2 t) < \infty$, then by the Borel–Cantelli lemma, we will be done. Now, the only way that N_t and X_t can differ by more than $\delta \log^2 t$ is if either one of the gaps between the first t ladder times is larger than $\delta \log^2 t$ or if for some $i < t$ the random walk backtracks $\delta \log^2 t$ steps after first reaching i . Thus,

$$(40) \quad \begin{aligned} \mathbb{P}(|N_t - X_t| \geq \delta \log^2 t) \\ \leq P(\exists j \in [1, t+1] : \nu_j - \nu_{j-1} > \delta \log^2 t) + t\mathbb{P}(T_{-\lceil \delta \log^2 t \rceil} < T_1) \end{aligned}$$

So, we need only to show that the two terms on the right side are summable. For the first term, we use Lemma 2.1 and note that

$$\begin{aligned} P(\exists j \in [1, t+1]: v_j - v_{j-1} > \delta \log^2 t) &\leq (t+1)P(v > \delta \log^2 t) \\ &\leq (t+1)C_1 e^{-C_2 \delta \log^2 t}, \end{aligned}$$

which is summable in t . By (39), the second term on the right side of (40) is also summable. \square

PROOF OF THEOREM 1.2. By Corollary 4.4, P -a.s. there exists a subsequence $j_m(\omega)$ such that $M_{j_m} \geq m^2 E_\omega \bar{T}_{v_{j_m-1}}^{(j_m)}$. Define $t_m = t_m(\omega) = \frac{1}{m} M_{j_m}$ and $u_m = u_m(\omega) = v_{j_m-1}$. Then

$$P_\omega \left(\frac{X_{t_m} - u_m}{\log^2 t_m} \notin [-\delta, \delta] \right) \leq P_\omega(N_{t_m} \neq j_m - 1) + P_\omega(|v_{N_{t_m}} - X_{t_m}| > \delta \log^2 t_m).$$

From Lemma 4.6, the second term goes to zero as $m \rightarrow \infty$. Thus, we only need to show that

$$(41) \quad \lim_{m \rightarrow \infty} P_\omega(N_{t_m} = j_m - 1) = 1.$$

To see this, first note that

$$\begin{aligned} P_\omega(N_{t_m} < j_m - 1) &= P_\omega(T_{v_{j_m-1}} > t_m) \\ &\leq P_\omega(T_{v_{j_m-1}} \neq \bar{T}_{v_{j_m-1}}^{(j_m)}) + P_\omega(\bar{T}_{v_{j_m-1}}^{(j_m)} > t_m). \end{aligned}$$

By Lemma 4.5, $P_\omega(T_{v_{j_m-1}} \neq \bar{T}_{v_{j_m-1}}^{(j_m)}) \rightarrow 0$ as $m \rightarrow \infty$, P -a.s. Also, by our definition of t_m and our choice of the subsequence j_m , we have

$$P_\omega(\bar{T}_{v_{j_m-1}}^{(j_m)} > t_m) \leq \frac{E_\omega \bar{T}_{v_{j_m-1}}^{(j_m)}}{t_m} = \frac{m E_\omega \bar{T}_{v_{j_m-1}}^{(j_m)}}{M_{j_m}} \leq \frac{1}{m} \xrightarrow{m \rightarrow \infty} 0.$$

It still remains to show $\lim_{m \rightarrow \infty} P_\omega(N_{t_m} < j_m) = 1$. To prove this, first define the stopping times $T_x^+ := \min\{n > 0: X_n = x\}$. Then

$$\begin{aligned} P_\omega(N_{t_m} < j_m) &= P_\omega(T_{v_{j_m}} > t_m) \geq P_\omega^{v_{j_m-1}} \left(T_{v_{j_m}} > \frac{1}{m} M_{j_m} \right) \\ &\geq P_\omega^{v_{j_m-1}} (T_{v_{j_m-1}}^+ < T_{v_{j_m}})^{(1/m) M_{j_m}}. \end{aligned}$$

Then using the hitting time calculations given in [14], (2.1.4), we have that

$$P_\omega^{v_{j_m-1}} (T_{v_{j_m-1}}^+ < T_{v_{j_m}}) = 1 - \frac{1 - \omega_{v_{j_m-1}}}{R_{v_{j_m-1}, v_{j_m-1}}}.$$

Therefore, since $M_{j_m} \leq R_{v_{j_m-1}, v_{j_m-1}}$, we have

$$P_\omega(N_{t_m} < j_m) \geq \left(1 - \frac{1 - \omega_{v_{j_m-1}}}{R_{v_{j_m-1}, v_{j_m-1}}} \right)^{(1/m) M_{j_m}} \geq \left(1 - \frac{1}{M_{j_m}} \right)^{(1/m) M_{j_m}} \xrightarrow{m \rightarrow \infty} 1,$$

thus proving (41) and, therefore, the theorem. \square

5. Nonlocal behavior on a random subsequence. There are two main goals of this section. The first is to prove the existence of random subsequences x_m where the hitting times T_{x_m} are approximately Gaussian random variables. This result is then used to prove the existence of random times $t_m(\omega)$ in which the scaling for the random walk is of the order t_m^s instead of $\log^2 t_m$ as in Theorem 1.2. However, before we can begin proving a quenched CLT for the hitting times T_n (at least along a random subsequence), we first need to understand the tail asymptotics of $\text{Var}_\omega T_v := E_\omega((T_v - E_\omega T_v)^2)$, the quenched variance of T_v .

5.1. *Tail asymptotics of $Q(\text{Var}_\omega T_v > x)$.* The goal of this subsection is to prove the following theorem.

THEOREM 5.1. *Let Assumptions 1 and 2 hold. Then with $K_\infty > 0$ the same as in Theorem 1.4, we have*

$$(42) \quad Q(\text{Var}_\omega T_v > x) \sim Q((E_\omega T_v)^2 > x) \sim K_\infty x^{-s/2} \quad \text{as } x \rightarrow \infty,$$

and for any $\varepsilon > 0$ and $x > 0$,

$$(43) \quad Q(\text{Var}_\omega \bar{T}_v^{(n)} > xn^{2/s}, M_1 > n^{(1-\varepsilon)/s}) \sim K_\infty x^{-s/2} \frac{1}{n} \quad \text{as } n \rightarrow \infty.$$

Consequently,

$$(44) \quad Q(\text{Var}_\omega T_v > \delta n^{1/s}, M_1 \leq n^{(1-\varepsilon)/s}) = o(n^{-1}).$$

A formula for the quenched variance of crossing times is given in [5], (2.2). Translating to our notation and simplifying, we have the formula

$$(45) \quad \text{Var}_\omega T_1 := E_\omega(T_1 - E_\omega T_1)^2 = 4(W_0 + W_0^2) + 8 \sum_{i < 0} \Pi_{i+1,0}(W_i + W_i^2).$$

Now, given the environment the crossing times $T_j - T_{j-1}$ are independent. Thus, we get the formula

$$(46) \quad \begin{aligned} \text{Var}_\omega T_v &= 4 \sum_{j=0}^{v-1} (W_j + W_j^2) + 8 \sum_{j=0}^{v-1} \sum_{i < j} \Pi_{i+1,j}(W_i + W_i^2) \\ &= 4 \sum_{j=0}^{v-1} (W_j + W_j^2) \\ &\quad + 8R_{0,v-1} \left(W_{-1} + W_{-1}^2 + \sum_{i < -1} \Pi_{i+1,-1}(W_i + W_i^2) \right) \\ &\quad + 8 \sum_{0 \leq i < j < v} \Pi_{i+1,j}(W_i + W_i^2). \end{aligned}$$

We want to analyze the tails of $\text{Var}_\omega T_\nu$ by comparison with $(E_\omega T_\nu)^2$. Using (14), we have

$$(E_\omega T_\nu)^2 = \left(\nu + 2 \sum_{j=0}^{\nu-1} W_j \right)^2 = \nu^2 + 4\nu \sum_{j=0}^{\nu-1} W_j + 4 \sum_{j=0}^{\nu-1} W_j^2 + 8 \sum_{0 \leq i < j < \nu} W_i W_j.$$

Thus, we have

$$(47) \quad (E_\omega T_\nu)^2 - \text{Var}_\omega T_\nu = \nu^2 + 4(\nu - 1) \sum_{j=0}^{\nu-1} W_j + 8 \sum_{0 \leq i < j < \nu} W_i (W_j - \Pi_{i+1,j} - \Pi_{i+1,j} W_i)$$

$$(48) \quad - 8R_{0,\nu-1} \left(W_{-1} + W_{-1}^2 + \sum_{i < -1} \Pi_{i+1,-1} (W_i + W_i^2) \right)$$

$$(49) \quad =: D^+(\omega) - 8R_{0,\nu-1} D^-(\omega).$$

Note that $D^-(\omega)$ and $D^+(\omega)$ are nonnegative random variables. The next few lemmas show that the tails of $D^+(\omega)$ and $R_{0,\nu-1} D^-(\omega)$ are much smaller than the tails of $(E_\omega T_\nu)^2$.

LEMMA 5.2. *For any $\varepsilon > 0$, we have $Q(D^+(\omega) > x) = o(x^{-s+\varepsilon})$.*

PROOF. Notice first that from (14) we have $\nu^2 + 4(\nu - 1) \sum_{j=0}^{\nu-1} W_j \leq 2\nu E_\omega T_\nu$. Also we can rewrite $W_j - \Pi_{i+1,j} - \Pi_{i+1,j} W_i = W_{i+2,j}$ when $i < j - 1$ (this term is zero when $i = j - 1$). Therefore,

$$Q(D^+(\omega) > x) \leq Q(2\nu E_\omega T_\nu > x/2) + Q\left(8 \sum_{i=0}^{\nu-3} \sum_{j=i+2}^{\nu-1} W_i W_{i+2,j} > x/2\right).$$

Lemma 2.1 and Theorem 1.4 give that $Q(2\nu E_\omega T_\nu > x) \leq Q(2\nu > \log^2(x)) + Q(E_\omega T_\nu > \frac{x}{\log^2(x)}) = o(x^{-s+\varepsilon})$ for any $\varepsilon > 0$. Thus, we need only prove that

$Q(\sum_{i=0}^{\nu-3} \sum_{j=i+2}^{\nu-1} W_i W_{i+2,j} > x) = o(x^{-s+\varepsilon})$ for any $\varepsilon > 0$. Note that for $i < \nu$, we have $W_i = W_{0,i} + \Pi_{0,i} W_{-1} \leq \Pi_{0,i} (i + 1 + W_{-1})$, thus,

$$(50) \quad Q\left(\sum_{i=0}^{\nu-3} \sum_{j=i+2}^{\nu-1} W_i W_{i+2,j} > x\right) \leq Q\left((\nu + W_{-1}) \sum_{i=0}^{\nu-3} \sum_{j=i+2}^{\nu-1} \Pi_{0,i} W_{i+2,j} > x\right) \leq Q(\nu > \log^2(x)/2) + Q(W_{-1} > \log^2(x)/2)$$

$$(51) \quad + \sum_{i=0}^{\log^2(x)-3\log^2(x)-1} \sum_{j=i+2}^{\nu-1} P\left(\Pi_{0,i} W_{i+2,j} > \frac{x}{\log^6(x)}\right),$$

where we were able to switch to P instead of Q in the last line because the event inside the probability only concerns the environment to the right of 0. Now, Lemmas 2.1 and 2.2 give that (50) is $o(x^{-s+\varepsilon})$ for any $\varepsilon > 0$, so we need only to consider (51). Under the measure P , we have that $\Pi_{0,i}$ and $W_{i+2,j}$ are independent, and by (9) we have $P(W_{i+2,j} > x) \leq P(W_j > x) \leq K_1 x^{-s}$. Thus,

$$\begin{aligned} P\left(\Pi_{0,i} W_{i+2,j} > \frac{x}{\log^6(x)}\right) &= E_P\left[P\left(W_{i+2,j} > \frac{x}{\log^6(x)\Pi_{0,i}} \mid \Pi_{0,i}\right)\right] \\ &\leq K_1 \log^{6s}(x)x^{-s} E_P[\Pi_{0,i}^s]. \end{aligned}$$

Then because $E_P \Pi_{0,i}^s = (E_P \rho^s)^{i+1} = 1$ by Assumption 1, we have

$$\begin{aligned} \sum_{i=0}^{\log^2(x)-3\log^2(x)-1} \sum_{j=i+2} P\left(\Pi_{0,i} W_{i+2,j} > \frac{x}{\log^6(x)}\right) &\leq K_1 \log^{4+6s}(x)x^{-s} \\ &= o(x^{-s+\varepsilon}). \end{aligned} \quad \square$$

LEMMA 5.3. For any $\varepsilon > 0$,

$$(52) \quad Q(D^-(\omega) > x) = o(x^{-s+\varepsilon}),$$

and thus for any $\gamma < s$,

$$(53) \quad E_Q D^-(\omega)^\gamma < \infty.$$

PROOF. It is obvious that (52) implies (53) and so we will only prove the former. For any i , we may expand $W_i + W_i^2$ as

$$\begin{aligned} W_i + W_i^2 &= \sum_{k \leq i} \Pi_{k,i} + \left(\sum_{k \leq i} \Pi_{k,i}\right)^2 = \sum_{k \leq i} \Pi_{k,i} + \sum_{k \leq i} \Pi_{k,i}^2 + 2 \sum_{k \leq i} \sum_{l < k} \Pi_{k,i} \Pi_{l,i} \\ &= \sum_{k \leq i} \Pi_{k,i} \left(1 + \Pi_{k,i} + 2 \sum_{l < k} \Pi_{l,i}\right). \end{aligned}$$

Therefore, we may rewrite

$$\begin{aligned} (54) \quad D^-(\omega) &= W_{-1} + W_{-1}^2 + \sum_{i < -1} \Pi_{i+1,-1}(W_i + W_i^2) \\ &= \sum_{i \leq -1} \sum_{k \leq i} \Pi_{k,-1} \left(1 + \Pi_{k,i} + 2 \sum_{l < k} \Pi_{l,i}\right). \end{aligned}$$

Next, for any $c > 0$ and $n \in \mathbb{N}$ define the event

$$\begin{aligned} E_{c,n} &:= \{\Pi_{j,i} \leq e^{-c(i-j+1)}, \forall -n \leq i \leq -1, \forall j \leq i - n\} \\ &= \bigcap_{-n \leq i \leq -1} \bigcap_{j \leq i - n} \{\Pi_{j,i} \leq e^{-c(i-j+1)}\}. \end{aligned}$$

Now, under the measure Q , we have that $\Pi_{k,-1} < 1$ for all $k \leq -1$, and thus on the event $E_{c,n}$ we have using the representation in (54) that

$$\begin{aligned}
 D^-(\omega) &= \sum_{i \leq -1} \sum_{k \leq i} \Pi_{k,-1} \left(1 + \Pi_{k,i} + 2 \sum_{l < k} \Pi_{l,i} \right) \\
 &\leq \sum_{-n \leq i \leq -1} \left(\sum_{k \leq i} \Pi_{k,i} (\Pi_{i+1,-1} + \Pi_{k,-1}) \right. \\
 &\quad \left. + 2 \sum_{i-n < k \leq i} \sum_{l < k} \Pi_{l,i} + 2 \sum_{l < k \leq i-n} e^{ck} \Pi_{l,i} \right) \\
 &\quad + \sum_{i < -n} \left(\sum_{k \leq i} e^{ck} + \sum_{k \leq i} e^{ck} \Pi_{k,i} + 2 \sum_{l < k \leq i} e^{ck} \Pi_{l,i} \right) \\
 (55) \quad &\leq \sum_{-n \leq i \leq -1} \left((2+n)W_i + 2 \sum_{l < k \leq i-n} e^{ck} e^{-c(i-l+1)} \right) \\
 &\quad + \sum_{i < -n} \left(\frac{e^{c(i+1)}}{e^c - 1} + e^{ci} W_i + \frac{2e^{c(i+1)}}{e^c - 1} \sum_{l < i} \Pi_{l,i} \right) \\
 &\leq (2+n) \sum_{-n \leq i \leq -1} W_i + \frac{2e^{-c(2n-1)}}{(e^c - 1)^3 (e^c + 1)} + \frac{e^{-c(n-1)}}{(e^c - 1)^2} \\
 &\quad + \sum_{i < -n} e^{ci} W_i \left(1 + \frac{2e^c}{e^c - 1} \right) \\
 &\leq (2+n) \sum_{-n \leq i \leq -1} W_i + \frac{e^c (1 + e^{2c})}{(e^c - 1)^3 (e^c + 1)} + \frac{3e^c - 1}{e^c - 1} \sum_{i < -n} e^{ci} W_i.
 \end{aligned}$$

Then using (55) with n replaced by $\lfloor \log^2 x \rfloor = b_x$ we have

$$\begin{aligned}
 Q(D^-(\omega) > x) &\leq Q(E_{c,b_x}^c) + \mathbf{1}_{\{e^c(1+e^{2c})/((e^c-1)^3(e^c+1)) > x/3\}} \\
 (56) \quad &\quad + Q\left(\sum_{-b_x \leq i \leq -1} W_i > \frac{x}{3(2+b_x)}\right) \\
 &\quad + Q\left(\sum_{i < -1} e^{ci} W_i > \frac{(e^c-1)x}{3(3e^c-1)}\right).
 \end{aligned}$$

Now, for any $0 < c < -E_P \log \rho$, Lemma 2.1 gives that $Q(\Pi_{i,j} > e^{-c(j-i+1)}) \leq \frac{A_c}{P(\mathcal{R})} e^{-\delta_c(j-i+1)}$ for some $\delta_c, A_c > 0$. Therefore,

$$\begin{aligned}
 Q(E_{c,n}^c) &\leq \sum_{-n \leq i \leq -1} \sum_{j \leq i-n} Q(\Pi_{j,i} > e^{-c(i-j+1)}) \\
 (57) \qquad &\leq \frac{nA_c e^{-\delta_c n}}{P(\mathcal{R})(e^{\delta_c} - 1)} = o(e^{-\delta_c n/2}).
 \end{aligned}$$

Thus, for any $0 < c < -E_P \log \rho$, we have that the first two terms on the right side of (56) are decreasing in x of order $o(e^{-\delta_c b_x/2}) = o(x^{-s+\varepsilon})$. To handle last two terms in the right side of (56), note first that from (9), $Q(W_i > x) \leq \frac{1}{P(\mathcal{R})} P(W_i > x) \leq \frac{K_1}{P(\mathcal{R})} x^{-s}$ for any $x > 0$ and any i . Thus,

$$\begin{aligned}
 Q\left(\sum_{-b_x \leq i \leq -1} W_i > \frac{x}{3(2+b_x)}\right) &\leq \sum_{-b_x \leq i \leq -1} Q\left(W_i > \frac{x}{3(2+b_x)b_x}\right) \\
 &= o(x^{-s+\varepsilon}),
 \end{aligned}$$

and since $\sum_{i=1}^\infty e^{-ci/2} = (e^{c/2} - 1)^{-1}$, we have

$$\begin{aligned}
 &Q\left(\sum_{i < -1} e^{ci} W_i > \frac{(e^c - 1)x}{9e^c - 3}\right) \\
 &\leq Q\left(\sum_{i=1}^\infty e^{-ci} W_{-i} > \frac{(e^c - 1)x}{9e^c - 3} (e^{c/2} - 1) \sum_{i=1}^\infty e^{-ci/2}\right) \\
 &\leq \sum_{i=1}^\infty Q\left(W_{-i} > \frac{(e^c - 1)(e^{c/2} - 1)}{9e^c - 3} x e^{ci/2}\right) \\
 &\leq \frac{K_1(9e^c - 3)^s}{P(\mathcal{R})(e^c - 1)^s (e^{c/2} - 1)^s} x^{-s} \sum_{i=1}^\infty e^{-csi/2} = \mathcal{O}(x^{-s}). \quad \square
 \end{aligned}$$

COROLLARY 5.4. For any $\varepsilon > 0$, $Q(R_{0,v-1} D^-(\omega) > x) = o(x^{-s+\varepsilon})$.

PROOF. From (11), it is easy to see that for any $\gamma < s$ there exists a $K_\gamma > 0$ such that $P(R_{0,v-1} > x) \leq P(R_0 > x) \leq K_\gamma x^{-\gamma}$. Then letting $\mathcal{F}_{-1} = \sigma(\dots, \omega_{-2}, \omega_{-1})$, we have that

$$\begin{aligned}
 Q(R_{0,v-1} D^-(\omega) > x) &= E_Q \left[Q\left(R_{0,v-1} > \frac{x}{D^-(\omega)} \mid \mathcal{F}_{-1}\right) \right] \\
 &\leq K_\gamma x^{-\gamma} E_Q(D^-(\omega))^\gamma.
 \end{aligned}$$

Since $\gamma < s$, the expectation in the last expression is finite by (53). Choosing $\gamma = s - \frac{\varepsilon}{2}$ completes the proof. \square

PROOF OF THEOREM 5.1. Recall from (49) that

$$(58) \quad (E_\omega T_v)^2 - D^+(\omega) \leq \text{Var}_\omega T_v \leq (E_\omega T_v)^2 + 8R_{0,v-1} D^-(\omega).$$

The lower bound in (58) gives that for any $\delta > 0$,

$$Q(\text{Var}_\omega T_v > x) \geq Q((E_\omega T_v)^2 > (1 + \delta)x) - Q(D^+(\omega) > \delta x).$$

Thus, from Lemma 5.2 and Theorem 1.4, we have that

$$(59) \quad \liminf_{x \rightarrow \infty} x^{s/2} Q(\text{Var}_\omega T_v > x) \geq K_\infty (1 + \delta)^{-s/2}.$$

Similarly, the upper bound in (58) and Corollary 5.4 give that for any $\delta > 0$,

$$Q(\text{Var}_\omega T_v > x) \leq Q((E_\omega T_v)^2 > (1 - \delta)x) + Q(8R_{0,v-1} D^-(\omega) > \delta x),$$

and then Corollary 5.4 and Theorem 1.4 give

$$(60) \quad \limsup_{x \rightarrow \infty} x^{s/2} Q(\text{Var}_\omega T_v > x) \leq K_\infty (1 - \delta)^{-s/2}.$$

Letting $\delta \rightarrow 0$ in (59) and (60) completes the proof of (42).

Essentially the same proof works for (43). The difference is that when evaluating the difference $(E_\omega \bar{T}_v^{(n)})^2 - \text{Var}_\omega \bar{T}_v^{(n)}$ the upper and lower bounds in (47) and (48) are smaller in absolute value. This is because every instance of W_i is replaced by $W_{v-b_n+1,i} \leq W_i$ and the sum in (48) is taken only over $v-b_n < i < -1$. Therefore, the following bounds still hold:

$$(61) \quad (E_\omega \bar{T}_v^{(n)})^2 - D^+(\omega) \leq \text{Var}_\omega \bar{T}_v^{(n)} \leq (E_\omega \bar{T}_v^{(n)})^2 + 8R_{0,v-1} D^-(\omega).$$

The rest of the proof then follows in the same manner, noting that from Lemma 3.3, we have $Q((E_\omega \bar{T}_v^{(n)})^2 > xn^{2/s}, M_1 > n^{(1-\varepsilon)/s}) \sim K_\infty x^{-s/2} \frac{1}{n}$, as $n \rightarrow \infty$. \square

5.2. *Existence of random subsequence of nonlocalized behavior.* Introduce the notation:

$$(62) \quad \begin{aligned} \mu_{i,n,\omega} &:= E_\omega^{v_i-1} \bar{T}_{v_i}^{(n)}, \\ \sigma_{i,n,\omega}^2 &:= E_\omega^{v_i-1} (\bar{T}_{v_i}^{(n)} - \mu_{i,n,\omega})^2 = \text{Var}_\omega (\bar{T}_{v_i}^{(n)} - \bar{T}_{v_{i-1}}^{(n)}). \end{aligned}$$

It is obvious (from the coupling of $\bar{X}_t^{(n)}$ and X_t) that $\mu_{i,n,\omega} \nearrow E_\omega^{v_i-1} T_{v_i}$ as $n \rightarrow \infty$. It is also true, although not as obvious, that $\sigma_{i,n,\omega}^2$ is increasing in n to $\text{Var}_\omega (T_{v_i} - T_{v_{i-1}})$. Therefore, we will use the notation $\mu_{i,\infty,\omega} := E_\omega^{v_i-1} T_{v_i}$ and $\sigma_{i,\infty,\omega}^2 := \text{Var}_\omega (T_{v_i} - T_{v_{i-1}})$. To see that $\sigma_{i,n,\omega}^2$ is increasing in n , note that the expansion for $\text{Var}_\omega \bar{T}_v^{(n)}$ is the same as the expansion for $\text{Var}_\omega T_v$ given in (46) but with each W_i replaced by $W_{v-b_n+1,i}$ and with the final sum in the second line restricted to $v-b_n < i < -1$.

The first goal of this subsection is to prove a CLT (along random subsequences) for the hitting times T_n . We begin by showing that for any $\varepsilon > 0$ only the crossing

times of ladder times with $M_k > n^{(1-\varepsilon)/s}$ are relevant in the limiting distribution, at least along a sparse enough subsequence.

LEMMA 5.5. *Assume $s < 2$. Then for any $\varepsilon, \delta > 0$, there exists an $\eta > 0$ and a sequence $c_n = o(n^{-\eta})$ such that for any $m \leq \infty$*

$$Q\left(\sum_{i=1}^n \sigma_{i,m,\omega}^2 \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} > \delta n^{2/s}\right) \leq c_n.$$

PROOF. Since $\sigma_{i,m,\omega}^2 \leq \sigma_{i,\infty,\omega}^2$, it is enough to consider only the case $m = \infty$ (that is, the walk without reflections). First, we need a bound on the probability of $\sigma_{i,\infty,\omega}^2 = \text{Var}_\omega(T_{v_i} - T_{v_{i-1}})$ being much larger than M_i^2 . Note that from (58), we have $\text{Var}_\omega T_v \leq (E_\omega T_v)^2 + 8R_{0,v-1} D^-(\omega)$. Then since $R_{0,v-1} \leq vM_1$, we have for any $\alpha, \beta > 0$ that

$$\begin{aligned} & Q(\text{Var}_\omega T_v > n^{2\beta}, M_1 \leq n^\alpha) \\ & \leq Q\left(E_\omega T_v > \frac{n^\beta}{\sqrt{2}}, M_1 \leq n^\alpha\right) + Q\left(8vD^-(\omega) > \frac{n^{2\beta-\alpha}}{2}\right). \end{aligned}$$

By (15), the first term on the right is $o(e^{-n^{(\beta-\alpha)/5}})$. To bound the second term on the right, we use Lemma 2.1 and Lemma 5.3 to get that for any $\alpha < \beta$

$$\begin{aligned} Q\left(8vD^-(\omega) > \frac{n^{2\beta-\alpha}}{2}\right) & \leq Q(v > \log^2 n) + Q\left(D^-(\omega) > \frac{n^{2\beta-\alpha}}{16\log^2 n}\right) \\ & = o(n^{-(s/2)(3\beta-\alpha)}). \end{aligned}$$

Therefore, similarly to (15), we have the bound

$$(63) \quad Q(\text{Var}_\omega T_v > n^{2\beta}, M_1 \leq n^\alpha) = o(n^{-(s/2)(3\beta-\alpha)}).$$

The rest of the proof is similar to the proof of Lemma 3.1. First, from (63),

$$\begin{aligned} & Q\left(\sum_{i=1}^n \sigma_{i,\infty,\omega}^2 \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} > \delta n^{2/s}\right) \\ & \leq Q\left(\sum_{i=1}^n \sigma_{i,\infty,\omega}^2 \mathbf{1}_{\sigma_{i,\infty,\omega}^2 \leq n^{2(1-\varepsilon/4)/s}} > \delta n^{2/s}\right) \\ & \quad + nQ(\text{Var}_\omega T_v > n^{2(1-\varepsilon/4)/s}, M_1 \leq n^{(1-\varepsilon)/s}) \\ & = Q\left(\sum_{i=1}^n \sigma_{i,\infty,\omega}^2 \mathbf{1}_{\sigma_{i,\infty,\omega}^2 \leq n^{2(1-\varepsilon/4)/s}} > \delta n^{2/s}\right) + o(n^{-\varepsilon/8}). \end{aligned}$$

Therefore, it is enough to prove that for any $\delta, \varepsilon > 0$, there exists $\eta > 0$ such that

$$Q\left(\sum_{i=1}^n \sigma_{i,\infty,\omega}^2 \mathbf{1}_{\sigma_{i,\infty,\omega}^2 \leq n^{2(1-\varepsilon/4)/s}} > \delta n^{2/s}\right) = o(n^{-\eta}).$$

We prove the above statement by choosing $C \in (1, \frac{2}{s})$, since $s > 2$, and then using Theorem 5.1 to get bounds on the size of the set $\{i \leq n : \text{Var}_\omega(T_{v_i} - T_{v_{i-1}}) \in (n^{2(1-\varepsilon C^k)/s}, n^{2(1-\varepsilon C^{k-1})/s})\}$ for all k small enough so that $\varepsilon C^k < 1$. This portion of the proof is similar to that of Lemma 3.1 and thus will be omitted. \square

COROLLARY 5.6. *Assume $s < 2$. Then for any $\delta > 0$, there exists an $\eta' > 0$ and a sequence $c'_n = o(n^{-\eta'})$ such that for any $m \leq \infty$*

$$Q\left(\left|\sum_{i=1}^n (\sigma_{i,m,\omega}^2 - \mu_{i,m,\omega}^2)\right| \geq \delta n^{2/s}\right) \leq c'_n.$$

PROOF. For any $\varepsilon > 0$,

$$\begin{aligned} & Q\left(\left|\sum_{i=1}^n (\sigma_{i,m,\omega}^2 - \mu_{i,m,\omega}^2)\right| \geq \delta n^{2/s}\right) \\ (64) \quad & \leq Q\left(\sum_{i=1}^n \sigma_{i,m,\omega}^2 \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} \geq \frac{\delta}{3} n^{2/s}\right) \\ (65) \quad & + Q\left(\sum_{i=1}^n \mu_{i,m,\omega}^2 \mathbf{1}_{M_i \leq n^{(1-\varepsilon)/s}} \geq \frac{\delta}{3} n^{2/s}\right) \\ (66) \quad & + Q\left(\sum_{i=1}^n |\sigma_{i,m,\omega}^2 - \mu_{i,m,\omega}^2| \mathbf{1}_{M_i > n^{(1-\varepsilon)/s}} \geq \frac{\delta}{3} n^{2/s}\right). \end{aligned}$$

Lemma 5.5 gives that (64) decreases polynomially in n (with a bound not depending on m). Also, essentially the same proof as in Lemmas 5.5 and 3.1 can be used to show that (65) also decreases polynomially in n (again with a bound not depending on m). Finally, (66) is bounded above by

$$\begin{aligned} & Q(\#\{i \leq n : M_i > n^{(1-\varepsilon)/s}\} > n^{2\varepsilon}) \\ & + n Q\left(|\text{Var}_\omega \bar{T}_v^{(m)} - (E_\omega \bar{T}_v^{(m)})^2| \geq \frac{\delta}{3} n^{2/s-2\varepsilon}\right), \end{aligned}$$

and since by (13), $Q(\#\{i \leq n : M_i > n^{(1-\varepsilon)/s}\} > n^{2\varepsilon}) \leq \frac{n Q(M_1 > n^{(1-\varepsilon)/s})}{n^{2\varepsilon}} \sim C_5 n^{-\varepsilon}$ we need only show that for some $\varepsilon > 0$ the second term above is decreasing faster than a power of n . However, from (61), we have $|\text{Var}_\omega \bar{T}_v^{(m)} - (E_\omega \bar{T}_v^{(m)})^2| \leq$

$D^+(\omega) + 8R_{0,v-1}D^-(\omega)$. Thus,

$$\begin{aligned} nQ\left(\left|\text{Var}_\omega \bar{T}_v^{(m)} - (E_\omega \bar{T}_v^{(m)})^2\right| \geq \frac{\delta}{3}n^{2/s-2\varepsilon}\right) \\ \leq nQ\left(D^+(\omega) + 8R_{0,v-1}D^-(\omega) > \frac{\delta}{3}n^{2/s-2\varepsilon}\right), \end{aligned}$$

and for any $\varepsilon < \frac{1}{2s}$ Lemma 5.2 and Corollary 5.4 give that the last term above decreases faster than some power of n . \square

Since $T_{v_n} = \sum_{i=1}^n (T_{v_i} - T_{v_{i-1}})$ is the sum of independent (quenched) random variables, in order to prove a CLT we cannot have any of the first n crossing times of blocks dominating all the others (note this is exactly what happens in the localization behavior we saw in Section 4). Thus, we look for a random subsequence where none of the crossing times of blocks are dominant. Now, for any $\delta \in (0, 1]$ and any positive integer $a < n/2$, define the event

$$\mathfrak{J}_{\delta,n,a} := \{\#\{i \leq \delta n : \mu_{i,n,\omega}^2 \in [n^{2/s}, 2n^{2/s}]\} = 2a, \mu_{j,n,\omega}^2 < 2n^{2/s} \ \forall j \leq \delta n\}.$$

On the event $\mathfrak{J}_{\delta,n,a}$, $2a$ of the first δn crossings times from v_{i-1} to v_i have roughly the same size expected crossing times $\mu_{i,n,\omega}$, and the rest are all smaller (we work with $\mu_{i,n,\omega}^2$ instead of $\mu_{i,n,\omega}$ so that comparisons with $\sigma_{i,n,\omega}^2$ are slightly easier). We want a lower bound on the probability of $\mathfrak{J}_{\delta,n,a}$. The difficulty in getting a lower bound is that the $\mu_{i,n,\omega}^2$ are not independent. However, we can force all the large crossing times to be independent by forcing them to be separated by at least b_n ladder locations.

Let $\mathcal{I}_{\delta,n,a}$ be the collection of all subsets I of $[1, \delta n] \cap \mathbb{Z}$ of size $2a$ with the property that any two distinct points in I are separated by at least $2b_n$. Also, define the event

$$A_{i,n} := \{\mu_{i,n,\omega}^2 \in [n^{2/s}, 2n^{2/s}]\}.$$

Then we begin with a simple lower bound:

$$\begin{aligned} (67) \quad Q(\mathfrak{J}_{\delta,n,a}) &\geq Q\left(\bigcup_{I \in \mathcal{I}_{\delta,n,a}} \left(\bigcap_{i \in I} A_{i,n} \ \bigcap_{j \in [1, \delta n] \setminus I} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right)\right) \\ &= \sum_{I \in \mathcal{I}_{\delta,n,a}} Q\left(\bigcap_{i \in I} A_{i,n} \ \bigcap_{j \in [1, \delta n] \setminus I} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right). \end{aligned}$$

Now, recall the definition of the event $G_{i,n,\varepsilon}$ from (33), and define the event

$$H_{i,n,\varepsilon} := \{M_j \leq n^{(1-\varepsilon)/s} \text{ for all } j \in [i - b_n, i]\}.$$

Also, for any $I \subset \mathbb{Z}$ let $d(j, I) := \min\{|j - i| : i \in I\}$ be the minimum distance from j to the set I . Then with minimal cost, we can assume that for any $I \in$

$\mathcal{I}_{\delta,n,a}$ and any $\varepsilon > 0$ that all $j \notin I$ such that $d(j, I) \leq b_n$ have $M_j \leq n^{(1-\varepsilon)/s}$. Indeed,

$$\begin{aligned}
 & Q\left(\bigcap_{i \in I} A_{i,n} \bigcap_{j \in [1, \delta n] \setminus I} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right) \\
 & \geq Q\left(\bigcap_{i \in I} (A_{i,n} \cap G_{i,n,\varepsilon} \cap H_{i,n,\varepsilon}) \bigcap_{j \in [1, \delta n]: d(j, I) > b_n} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right) \\
 (68) \quad & - Q\left(\bigcup_{j \notin I, d(j, I) \leq b_n} \{\mu_{j,n,\omega}^2 \geq n^{2/s}, M_j \leq n^{(1-\varepsilon)/s}\}\right) \\
 & \geq \prod_{i \in I} Q(A_{i,n} \cap H_{i,n,\varepsilon}) Q\left(\bigcap_{i \in I} G_{i,n,\varepsilon} \bigcap_{j \in [1, \delta n]: d(j, I) > b_n} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right) \\
 & \quad - 4ab_n Q(E_\omega T_v \geq n^{1/s}, M_1 \leq n^{(1-\varepsilon)/s}).
 \end{aligned}$$

From Theorem 1.4 and Lemma 3.3, we have $Q(A_{i,n}) \sim K_\infty(1 - 2^{-s/2})n^{-1}$. We wish to show the same asymptotics are true for $Q(A_{i,n} \cap H_{i,n,\varepsilon})$ as well. From (13), we have $Q(H_{i,n,\varepsilon}^c) \leq b_n Q(M_1 > n^{(1-\varepsilon)/s}) = o(n^{-1+2\varepsilon})$. Applying this, along with (13) and (15), gives that for $\varepsilon > 0$,

$$\begin{aligned}
 Q(A_{i,n}) & \leq Q(A_{i,n} \cap H_{i,n,\varepsilon}) + Q(M_1 > n^{(1-\varepsilon)/s}) Q(H_{i,n,\varepsilon}^c) \\
 & \quad + Q(E_\omega T_v > n^{1/s}, M_1 \leq n^{(1-\varepsilon)/s}) \\
 & = Q(A_{i,n} \cap H_{i,n,\varepsilon}) + o(n^{-2+3\varepsilon}) + o(e^{-n^{\varepsilon/(5s)}}).
 \end{aligned}$$

Thus, for any $\varepsilon < \frac{1}{3}$, there exists a $C_\varepsilon > 0$ such that

$$(69) \quad Q(A_{i,n} \cap H_{i,n,\varepsilon}) \geq C_\varepsilon n^{-1}.$$

To handle the next probability in (68), note that

$$\begin{aligned}
 & Q\left(\bigcap_{i \in I} G_{i,n,\varepsilon} \bigcap_{j \in [1, \delta n]: d(j, I) > b_n} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right) \\
 & \geq Q\left(\bigcap_{j \in [1, \delta n]} \{\mu_{j,n,\omega}^2 < n^{2/s}\}\right) - Q\left(\bigcup_{i \in I} G_{i,n,\varepsilon}^c\right) \\
 (70) \quad & \geq Q(E_\omega T_{v_n} < n^{1/s}) - 2a Q(G_{i,n,\varepsilon}^c) \\
 & = Q(E_\omega T_{v_n} < n^{1/s}) - ao(n^{-1+2\varepsilon}).
 \end{aligned}$$

Finally, from (15), we have $4ab_n Q(E_\omega T_v \geq n^{1/s}, M_1 \leq n^{(1-\varepsilon)/s}) = ao(e^{-n^{\varepsilon/(6s)}})$.

This along with (69) and (70) applied to (67) gives

$$Q(\mathcal{I}_{\delta,n,a}) \geq \#(\mathcal{I}_{\delta,n,a})[(C_\varepsilon n^{-1})^{2a}(Q(E_\omega T_{v_n} < n^{1/s}) - ao(n^{-1+2\varepsilon})) - ao(e^{-n^\varepsilon/(6s)})].$$

An obvious upper bound for $\#(\mathcal{I}_{\delta,n,a})$ is $\binom{\delta n}{2a} \leq \frac{(\delta n)^{2a}}{(2a)!}$. To get a lower bound on $\#(\mathcal{I}_{\delta,n,a})$, we note that any set $I \in \mathcal{I}_{\delta,n,a}$ can be chosen in the following way: first choose an integer $i_1 \in [1, \delta n]$ (δn ways to do this). Then choose an integer $i_2 \in [1, \delta n] \setminus \{j \in \mathbb{Z} : |j - i_1| \leq 2b_n\}$ (at least $\delta n - 1 - 4b_n$ ways to do this). Continue this process until $2a$ integers have been chosen. When choosing i_j , there will be at least $\delta n - (j - 1)(1 + 4b_n)$ integers available. Then since there are $(2a)!$ orders in which to choose each set of $2a$ integers, we have

$$\begin{aligned} \frac{(\delta n)^{2a}}{(2a)!} &\geq \#(\mathcal{I}_{\delta,n,a}) \geq \frac{1}{(2a)!} \prod_{j=1}^{2a} (\delta n - (j - 1)(1 + 4b_n)) \\ &\geq \frac{(\delta n)^{2a}}{(2a)!} \left(1 - \frac{(2a - 1)(1 + 4b_n)}{\delta n}\right)^{2a}. \end{aligned}$$

Therefore, applying the upper and lower bounds on $\#(\mathcal{I}_{\delta,n,a})$, we get

$$\begin{aligned} Q(\mathcal{I}_{\delta,n,a}) &\geq \frac{(\delta C_\varepsilon)^{2a}}{(2a)!} \left(1 - \frac{(2a - 1)(1 + 4b_n)}{\delta n}\right)^{2a} \\ &\quad \times (Q(E_\omega T_{v_n} < n^{1/s}) - ao(n^{-1+2\varepsilon})) \\ &\quad - \frac{(\delta n)^{2a}}{(2a)!} ao(e^{-n^\varepsilon/(6s)}). \end{aligned}$$

Recall the definitions of d_k in (36) and define

$$(71) \quad a_k := \lfloor \log \log k \rfloor \vee 1 \quad \text{and} \quad \delta_k := a_k^{-1}.$$

Now, replacing δ, n and a in the above by δ_k, d_k and a_k , respectively, we have

$$\begin{aligned} (72) \quad Q(\mathcal{I}_{\delta_k, d_k, a_k}) &\geq \frac{(\delta_k C_\varepsilon)^{2a_k}}{(2a_k)!} \left(1 - \frac{(2a_k - 1)(1 + 4b_{d_k})}{\delta_k d_k}\right)^{2a_k} \\ &\quad \times (Q(E_\omega T_{v_{d_k}} < d_k^{1/s}) - a_k o(d_k^{-1+2\varepsilon})) \\ &\quad - \frac{(\delta_k d_k)^{2a_k}}{(2a_k)!} a_k o(e^{-d_k^\varepsilon/(6s)}) \\ &\geq \frac{(\delta_k C_\varepsilon)^{2a_k}}{(2a_k)!} (1 + o(1))(L_{s,b'}(1) - o(1)) - o(1/k). \end{aligned}$$

The last inequality is a result of the definitions of δ_k, a_k , and d_k (it's enough to recall that $d_k \geq 2^{2k-1}$, $a_k \sim \log \log k$, and $\delta_k \sim \frac{1}{\log \log k}$), as well as Theorem 1.1.

Also, since $\delta_k = a_k^{-1}$, we get from Stirling's formula that $\frac{(\delta_k C_\varepsilon)^{2a_k}}{(2a_k)!} \sim \frac{(C_\varepsilon e/2)^{2a_k}}{\sqrt{2\pi a_k}}$.

Thus, since $a_k \sim \log \log k$, we have that $\frac{1}{k} = o(\frac{(\delta_k C_\varepsilon)^{2a_k}}{(2a_k)!})$. This, along with (72), gives that $Q(\mathcal{G}_{\delta_k, d_k, a_k}) > \frac{1}{k}$ for all k large enough.

We now have a good lower bound on the probability of not having any of the crossing times of the first $\delta_k d_k$ blocks dominating all the others. However, for the purpose of proving Theorem 1.3, we need a little bit more. We also need that none of the crossing times of succeeding blocks are too large either. Thus, for any $0 < \delta < c$ and $n \in \mathbb{N}$, define the events

$$U_{\delta, n, c} := \left\{ \sum_{i=\delta n+1}^{cn} \mu_{i, n, \omega} \leq 2n^{1/s} \right\}, \quad \tilde{U}_{\delta, n, c} := \left\{ \sum_{i=\delta n+b_{n+1}}^{cn} \mu_{i, n, \omega} \leq n^{1/s} \right\}.$$

LEMMA 5.7. *Assume $s < 1$. Then there exists a sequence $c_k \rightarrow \infty$, $c_k = o(\log a_k)$ such that*

$$\sum_{k=1}^{\infty} Q(\mathcal{G}_{\delta_k, d_k, a_k} \cap U_{\delta_k, d_k, c_k}) = \infty.$$

PROOF. For any $\delta < c$ and $a < n/2$, we have

$$\begin{aligned} Q(\mathcal{G}_{\delta, n, a} \cap U_{\delta, n, c}) &\geq Q(\mathcal{G}_{\delta, n, a})Q(\tilde{U}_{\delta, n, c}) - Q\left(\sum_{i=1}^{b_n} \mu_{i, n, \omega} > n^{1/s}\right) \\ (73) \qquad \qquad \qquad &\geq Q(\mathcal{G}_{\delta, n, a})Q(E_\omega T_{v_{cn}} \leq n^{1/s}) - b_n Q\left(E_\omega T_v > \frac{n^{1/s}}{b_n}\right) \\ &\geq Q(\mathcal{G}_{\delta, n, a})Q(E_\omega T_{v_{cn}} \leq n^{1/s}) - o(n^{-1/2}), \end{aligned}$$

where the last inequality is from Theorem 1.4. Now, define $c_1 = 1$ and for $k > 1$ let

$$c'_k := \max \left\{ c \in \mathbb{N} : Q(E_\omega T_{v_{cd_k}} \leq d_k^{1/s}) \geq \frac{1}{\log k} \right\} \vee 1.$$

Note that by Theorem 1.1 we have that $c'_k \rightarrow \infty$, and so we can define $c_k = c'_k \wedge \log \log(a_k)$. Then applying (73) with this choice of c_k we have

$$\begin{aligned} &\sum_{k=1}^{\infty} Q(\mathcal{G}_{\delta_k, d_k, a_k} \cap U_{\delta_k, d_k, c_k}) \\ &\geq \sum_{k=1}^{\infty} [Q(\mathcal{G}_{\delta_k, d_k, a_k})Q(E_\omega T_{v_{c_k d_k}} \leq d_k^{1/s}) - o(d_k^{-1/2})] = \infty, \end{aligned}$$

and the last sum is infinite because $d_k^{-1/2}$ is summable and for all k large enough we have

$$Q(\mathcal{S}_{\delta_k, d_k, a_k})Q(E_\omega T_{v_{c_k d_k}} \leq d_k^{1/s}) \geq \frac{1}{k \log k}. \quad \square$$

COROLLARY 5.8. *Assume $s < 1$, and let c_k be as in Lemma 5.7. Then P -a.s. there exists a random subsequence $n_{k_m} = n_{k_m}(\omega)$ of $n_k = 2^{2^k}$ such that for the sequences α_m, β_m , and γ_m defined by*

$$(74) \quad \begin{aligned} \alpha_m &:= n_{k_m-1}, \\ \beta_m &:= n_{k_m-1} + \delta_{k_m} d_{k_m}, \\ \gamma_m &:= n_{k_m-1} + c_{k_m} d_{k_m}, \end{aligned}$$

we have that for all m

$$(75) \quad \max_{i \in (\alpha_m, \beta_m]} \mu_{i, d_{k_m}, \omega}^2 \leq 2d_{k_m}^{2/s} \leq \frac{1}{a_{k_m}} \sum_{i=\alpha_m+1}^{\beta_m} \mu_{i, d_{k_m}, \omega}^2$$

and

$$\sum_{\beta_m+1}^{\gamma_m} \mu_{i, d_{k_m}, \omega} \leq 2d_{k_m}^{1/s}.$$

PROOF. Define the events

$$\begin{aligned} \mathcal{S}'_k &:= \{\#\{i \in (n_{k-1}, n_{k-1} + \delta_k d_k) : \mu_{i, d_k, \omega}^2 \in [d_k^{2/s}, 2d_k^{2/s})\} = 2a_k\} \\ &\quad \cap \{\mu_{j, d_k, \omega}^2 < 2d_k^{2/s} \ \forall j \in (n_{k-1}, n_{k-1} + \delta_k d_k)\}, \\ U'_k &:= \left\{ \sum_{n_{k-1} + \delta_k d_k + 1}^{n_{k-1} + c_k d_k} \mu_{i, d_k, \omega} \leq 2d_k^{1/s} \right\}. \end{aligned}$$

Note that due to the reflections of the random walk, the event $\mathcal{S}'_k \cap U'_k$ depends on the environment between ladder locations $n_{k-1} - b_{d_k}$ and $n_{k-1} + c_k d_k$. Thus, for k_0 large enough $\{\mathcal{S}'_{2k} \cap U'_{2k}\}_{k=k_0}^\infty$ is an independent sequence of events. Similarly, for k large enough $\mathcal{S}'_k \cap U'_k$ does not depend on the environment to left of the origin. Thus,

$$P(\mathcal{S}'_k \cap U'_k) = Q(\mathcal{S}'_k \cap U'_k) = Q(\mathcal{S}_{\delta_k, d_k, a_k} \cap U_{\delta_k, d_k, c_k})$$

for all k large enough. Lemma 5.7 then gives that $\sum_{k=1}^\infty P(\mathcal{S}'_{2k} \cap U'_{2k}) = \infty$, and the Borel–Cantelli lemma then implies that infinitely many of the events $\mathcal{S}'_{2k} \cap U'_{2k}$ occur P -a.s. Finally, note that \mathcal{S}'_{k_m} implies the event in (75). \square

Before proving a quenched CLT (along a subsequence) for the hitting times T_n , we need one more lemma that gives us some control on the quenched tails of crossing times of blocks. We can get this from an application of Kac’s moment formula. Let \bar{T}_y be the hitting time of y when we add a reflection at the starting point of the random walk. Then Kac’s moment formula ([3], (6)) and the Markov property give that $E_\omega^x(\bar{T}_y)^j \leq j!(E_\omega^x \bar{T}_y)^j$ (note that because of the reflection at x , $E_\omega^x(\bar{T}_y) \geq E_\omega^{x'}(\bar{T}_y)$ for any $x' \in (x, y)$). Thus,

$$(76) \quad \begin{aligned} E_\omega^{v_{i-1}}(\bar{T}_{v_i}^{(n)})^j &\leq E_\omega^{v_{i-1}-b_n}(\bar{T}_{v_i})^j \leq j!(E_\omega^{v_{i-1}-b_n} \bar{T}_{v_i})^j \\ &\leq j!(E_\omega^{v_{i-1}-b_n} \bar{T}_{v_{i-1}} + \mu_{i,n,\omega})^j. \end{aligned}$$

LEMMA 5.9. *For any $\varepsilon < \frac{1}{3}$, there exists an $\eta > 0$ such that*

$$Q(\exists i \leq n, j \in \mathbb{N} : M_i > n^{(1-\varepsilon)/s}, E_\omega^{v_{i-1}}(\bar{T}_{v_i}^{(n)})^j > j!2^j \mu_{i,n,\omega}^j) = o(n^{-\eta}).$$

PROOF. We use (76) to get

$$\begin{aligned} Q(\exists i \leq n, j \in \mathbb{N} : M_i > n^{(1-\varepsilon)/s}, E_\omega^{v_{i-1}}(\bar{T}_{v_i}^{(n)})^j > j!2^j \mu_{i,n,\omega}^j) \\ \leq Q(\exists i \leq n : M_i > n^{(1-\varepsilon)/s}, E_\omega^{v_{i-1}-b_n} \bar{T}_{v_{i-1}} > \mu_{i,n,\omega}) \\ \leq nQ(M_1 > n^{(1-\varepsilon)/s}, E_\omega^{v-b_n} T_0 > n^{(1-\varepsilon)/s}) \\ = nQ(M_1 > n^{(1-\varepsilon)/s})Q(E_\omega^{v-b_n} T_0 > n^{(1-\varepsilon)/s}), \end{aligned}$$

where the second inequality is due to a union bound and the fact that $\mu_{i,n,\omega} > M_i$. Now, by (13), we have $nQ(M_1 > n^{(1-\varepsilon)/s}) \sim C_5 n^\varepsilon$, and by Theorem 1.4,

$$Q(E_\omega^{v-b_n} T_0 > n^{(1-\varepsilon)/s}) \leq b_n Q\left(E_\omega T_v > \frac{n^{(1-\varepsilon)/s}}{b_n}\right) \sim K_\infty b_n^{1+s} n^{-1+\varepsilon}.$$

Therefore, $Q(\exists i \leq n, j \in \mathbb{N} : M_i > n^{(1-\varepsilon)/s}, E_\omega^{v_{i-1}}(\bar{T}_{v_i}^{(n)})^j > j!2^j \mu_{i,n,\omega}^j) = o(n^{-1+3\varepsilon})$. \square

THEOREM 5.10. *Let Assumptions 1 and 2 hold, and let $s < 1$. Then P -a.s. there exists a random subsequence $n_{k_m} = n_{k_m}(\omega)$ of $n_k = 2^{2^k}$ such that for α_m, β_m and γ_m as in (74) and any sequence $x_m \in [v_{\beta_m}, v_{\gamma_m}]$, we have*

$$(77) \quad \lim_{m \rightarrow \infty} P_\omega\left(\frac{T_{x_m} - E_\omega T_{x_m}}{\sqrt{v_{m,\omega}}} \leq y\right) = \Phi(y),$$

where

$$v_{m,\omega} := \sum_{i=\alpha_m+1}^{\beta_m} \mu_{i,d_{k_m},\omega}^2.$$

PROOF. Let $n_{k_m}(\omega)$ be the random subsequence specified in Corollary 5.8. For ease of notation, set $\tilde{a}_m = a_{k_m}$ and $\tilde{d}_m = d_{k_m}$. We have

$$\max_{i \in (\alpha_m, \beta_m]} \mu_{i, \tilde{d}_m, \omega}^2 \leq 2\tilde{d}_m^{2/s} \leq \frac{1}{\tilde{a}_m} \sum_{i=\alpha_m+1}^{\beta_m} \mu_{i, \tilde{d}_m, \omega}^2 = \frac{v_{m, \omega}}{\tilde{a}_m}$$

and

$$\sum_{i=\beta_m+1}^{\gamma_m} \mu_{i, \tilde{d}_m, \omega} \leq 2\tilde{d}_m^{1/s}.$$

Now, let $\{x_m\}_{m=1}^\infty$ be any sequence of integers (even depending on ω) such that $x_m \in [v_{\beta_m}, v_{\gamma_m}]$. Then since $(T_{x_m} - E_\omega T_{x_m}) = (T_{v_{\alpha_m}} - E_\omega T_{v_{\alpha_m}}) + (T_{x_m} - T_{v_{\alpha_m}} - E_\omega^{v_{\alpha_m}} T_{x_m})$, it is enough to prove

$$(78) \quad \frac{T_{v_{\alpha_m}} - E_\omega T_{v_{\alpha_m}}}{\sqrt{v_{m, \omega}}} \xrightarrow{\mathcal{D}_\omega} 0 \quad \text{and} \quad \frac{T_{x_m} - T_{v_{\alpha_m}} - E_\omega^{v_{\alpha_m}} T_{x_m}}{\sqrt{v_{m, \omega}}} \xrightarrow{\mathcal{D}_\omega} Z \sim N(0, 1),$$

where we use the notation $Z_n \xrightarrow{\mathcal{D}_\omega} Z$ to denote quenched convergence in distribution, that is $\lim_{n \rightarrow \infty} P_\omega(Z_n \leq z) = P_\omega(Z \leq z)$, P -a.s. For the first term in (78), note that for any $\varepsilon > 0$, we have from Chebyshev's inequality and $v_{m, \omega} \geq \tilde{d}_m^{2/s}$ that

$$P_\omega \left(\left| \frac{T_{v_{\alpha_m}} - E_\omega T_{v_{\alpha_m}}}{\sqrt{v_{m, \omega}}} \right| \geq \varepsilon \right) \leq \frac{\text{Var}_\omega T_{v_{\alpha_m}}}{\varepsilon^2 v_{m, \omega}} \leq \frac{\text{Var}_\omega T_{v_{\alpha_m}}}{\varepsilon^2 \tilde{d}_m^{2/s}}.$$

Thus, the first claim in (78) will be proved if we can show that $\text{Var}_\omega T_{v_{\alpha_m}} = o(\tilde{d}_m^{2/s})$. For this, we need the following lemma.

LEMMA 5.11. Assume $s \leq 2$. Then for any $\delta > 0$,

$$P(\text{Var}_\omega T_{v_n} \geq n^{2/s+\delta}) = o(n^{-\delta s/4}).$$

PROOF. First, we claim that

$$(79) \quad E_P(\text{Var}_\omega T_1)^\gamma < \infty \text{ for any } \gamma < \frac{s}{2}.$$

Indeed, from (45), we have that for any $\gamma < \frac{s}{2} \leq 1$,

$$\begin{aligned} E_P(\text{Var}_\omega T_1)^\gamma &\leq 4^\gamma E_P(W_0 + W_0^2)^\gamma + 8^\gamma \sum_{i < 0} E_P(\Pi_{i+1,0}^\gamma (W_i + W_i^2)^\gamma) \\ &= 4^\gamma E_P(W_0 + W_0^2)^\gamma + 8^\gamma \sum_{i=1}^\infty (E_P \rho_0^\gamma)^i E_P(W_0 + W_0^2)^\gamma, \end{aligned}$$

where we used that P is i.i.d. in the last equality. Since $E_P \rho_0^\gamma < 1$ for any $\gamma \in (0, s)$, we have that (79) follows as soon as $E_P(W_0 + W_0^2)^\gamma < \infty$. However, from (9), we get that $E_P(W_0 + W_0^2)^\gamma < \infty$ when $\gamma < \frac{s}{2}$.

As in Lemma 4.2 let $\bar{v} = E_P v$. Then,

$$P(\text{Var}_\omega T_{v_n} \geq n^{2/s+\delta}) \leq P(\text{Var}_\omega T_{2\bar{v}n} \geq n^{2/s+\delta}) + P(v_n \geq 2\bar{v}n).$$

As in Lemma 4.2, the second term is $\mathcal{O}(e^{-\delta'n})$ for some $\delta' > 0$. To handle the first term on the right side, we note that for any $\gamma < \frac{s}{2} \leq 1$,

$$\begin{aligned} P(\text{Var}_\omega T_{2\bar{v}n} \geq n^{2/s+\delta}) &\leq \frac{E_P(\sum_{k=1}^{2\bar{v}n} \text{Var}_\omega(T_k - T_{k-1}))^\gamma}{n^{\gamma(2/s+\delta)}} \\ (80) \qquad \qquad \qquad &\leq \frac{2\bar{v}n E_P(\text{Var}_\omega T_1)^\gamma}{n^{\gamma(2/s+\delta)}}. \end{aligned}$$

Then since $E_P(\text{Var}_\omega T_1)^\gamma < \infty$ for any $\gamma < \frac{s}{2}$, we can choose γ arbitrarily close to $\frac{s}{2}$ so that the last term on the right of (80) is $o(n^{-\delta s/4})$. \square

As a result of Lemma 5.11 and the Borel–Cantelli lemma, we have that $\text{Var}_\omega T_{v_{n_k}} = o(n_k^{2/s+\delta})$ for any $\delta > 0$. Therefore, for any $\delta \in (0, \frac{2}{s})$, we have $\text{Var}_\omega T_{v_{\alpha_m}} = o(\alpha_m^{2/s+\delta}) = o(n_{k_m-1}^{2/s+\delta}) = o(\tilde{d}_m^{2/s})$ (in the last equality we use that $d_k \sim n_k$ to grow much faster than exponentially in k).

For the next step in the proof, we show that reflections can be added without changing the limiting distribution. Specifically, we show that it is enough to prove the following lemma, whose proof we postpone.

LEMMA 5.12. *With notation as in Theorem 5.10, we have*

$$(81) \qquad \lim_{m \rightarrow \infty} P_\omega^{v_{\alpha_m}} \left(\frac{\bar{T}_{x_m}^{(\tilde{d}_m)} - E_\omega^{v_{\alpha_m}} \bar{T}_{x_m}^{(\tilde{d}_m)}}{\sqrt{v_{m,\omega}}} \leq y \right) = \Phi(y).$$

Assuming Lemma 5.12, we complete the proof of Theorem 5.10. It is enough to show that

$$(82) \qquad \lim_{m \rightarrow \infty} P_\omega^{v_{\alpha_m}} (\bar{T}_{x_{k_m}}^{(\tilde{d}_m)} \neq T_{x_m}) = 0 \quad \text{and} \quad \lim_{m \rightarrow \infty} E_\omega^{v_{\alpha_m}} (T_{x_m} - \bar{T}_{x_{k_m}}^{(\tilde{d}_m)}) = 0.$$

Recall that the coupling introduced after (18) gives that $T_{x_m} - \bar{T}_{x_m}^{(\tilde{d}_m)} \geq 0$. Thus,

$$P_\omega^{v_{\alpha_m}} (\bar{T}_{x_m}^{(\tilde{d}_m)} \neq T_{x_m}) = P_\omega^{v_{\alpha_m}} (T_{x_m} - \bar{T}_{x_m}^{(\tilde{d}_m)} \geq 1) \leq E_\omega^{v_{\alpha_m}} (T_{x_m} - \bar{T}_{x_m}^{(\tilde{d}_m)}).$$

Then since $x_m \leq v_{\gamma_m}$ and $\gamma_m = n_{k_m-1} + c_{k_m} \tilde{d}_m \leq n_{k_m+1}$ for all m large enough, (82) will follow from

$$(83) \qquad \lim_{k \rightarrow \infty} E_\omega^{v_{n_{k-1}}} (T_{v_{n_{k+1}}} - \bar{T}_{v_{n_{k+1}}}^{(d_k)}) = 0, \quad P\text{-a.s.}$$

To prove (83), we argue as follows. From Lemma 3.2 we have that for any $\varepsilon > 0$

$$\begin{aligned} Q(E_\omega^{v_{n_{k-1}}} (T_{v_{n_{k+1}}} - \bar{T}_{v_{n_{k+1}}}^{(d_k)}) > \varepsilon) &\leq n_{k+1} Q\left(E_\omega T_v - E_\omega \bar{T}_v^{(d_k)} > \frac{\varepsilon}{n_{k+1}}\right) \\ &= n_{k+1} \mathcal{O}(n_{k+1}^s e^{-\delta' b d_k}). \end{aligned}$$

Since $n_k \sim d_k$, the last term on the right is summable. Therefore, by the Borel–Cantelli lemma,

$$(84) \quad \lim_{k \rightarrow \infty} E_\omega^{v_{n_{k+1}}-1} (T_{v_{n_{k+1}}} - \bar{T}_{v_{n_{k+1}}}^{(d_k)}) = 0, \quad Q\text{-a.s.}$$

This is almost the same as (83), but with Q instead of P . To use this to prove (83), note that for $i > b_n$ using (19), we can write

$$E_\omega^{v_i-1} T_{v_i} - E_\omega^{v_i-1} \bar{T}_{v_i}^{(n)} = A_{i,n}(\omega) + B_{i,n}(\omega)W_{-1},$$

where $A_{i,n}(\omega)$ and $B_{i,n}(\omega)$ are nonnegative random variables depending only on the environment to the right of 0. Thus, $E_\omega^{v_{n_{k+1}}-1} (T_{v_{n_{k+1}}} - \bar{T}_{v_{n_{k+1}}}^{(d_k)}) = A_{d_k}(\omega) + B_{d_k}(\omega)W_{-1}$ where $A_{d_k}(\omega)$ and $B_{d_k}(\omega)$ are nonnegative and only depend on the environment to the right of zero (so A_{d_k} and B_{d_k} have the same distribution under P as under Q). Therefore, (83) follows from (84), which completes the proof of the theorem. \square

PROOF OF LEMMA 5.12. Clearly, it suffices to show the following claims:

$$(85) \quad \frac{\bar{T}_{x_m}^{(\tilde{d}_m)} - \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} - E_\omega^{v_{\beta_m}} \bar{T}_{x_m}^{(\tilde{d}_m)}}{\sqrt{v_{m,\omega}}} \xrightarrow{\mathcal{D}_\omega} 0$$

and

$$(86) \quad \frac{\bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} - \bar{T}_{v_{\alpha_m}}^{(\tilde{d}_m)} - E_\omega^{v_{\alpha_m}} \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)}}{\sqrt{v_{m,\omega}}} \xrightarrow{\mathcal{D}_\omega} Z \sim N(0, 1).$$

To prove (85), we note that

$$P_\omega \left(\left| \frac{\bar{T}_{x_m}^{(\tilde{d}_m)} - \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} - E_\omega^{v_{\beta_m}} \bar{T}_{x_m}^{(\tilde{d}_m)}}{\sqrt{v_{m,\omega}}} \right| \geq \varepsilon \right) \leq \frac{\text{Var}_\omega(\bar{T}_{x_m}^{(\tilde{d}_m)} - \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)})}{\varepsilon^2 v_{m,\omega}} \leq \frac{\sum_{i=\beta_m+1}^{\gamma_m} \sigma_{i,\tilde{d}_m,\omega}^2}{\varepsilon^2 \tilde{d}_m \tilde{d}_m^{2/s}},$$

where the last inequality is because $x_m \leq v_{\gamma_m}$ and $v_{m,\omega} \geq \tilde{d}_m \tilde{d}_m^{2/s}$. However, by Corollary 5.6 and the Borel–Cantelli lemma,

$$\sum_{i=\beta_m+1}^{\gamma_m} \sigma_{i,\tilde{d}_m,\omega}^2 = \sum_{i=\beta_m+1}^{\gamma_m} \mu_{i,\tilde{d}_m,\omega}^2 + o((c_{k_m} \tilde{d}_m)^{2/s}).$$

The application of Corollary 5.6 uses the fact that for k large enough the reflections ensure that the events in question do not involve the environment to the left of zero, and thus have the same probability under P or Q . (This type of argument will be used a few more times in the remainder of the proof without mention.) By our choice of the subsequence n_{k_m} , we have

$$\sum_{i=\beta_m+1}^{\gamma_m} \mu_{i,\tilde{d}_m,\omega}^2 \leq \left(\sum_{i=\beta_m+1}^{\gamma_m} \mu_{i,\tilde{d}_m,\omega} \right)^2 \leq 4\tilde{d}_m^{2/s}.$$

Therefore,

$$\begin{aligned} & \lim_{m \rightarrow \infty} P_\omega \left(\left| \frac{\bar{T}_{x_m}^{(\tilde{d}_m)} - \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} - E_\omega^{v_{\beta_m}} \bar{T}_{x_m}^{(\tilde{d}_m)}}{\sqrt{v_{m,\omega}}} \right| \geq \varepsilon \right) \\ & \leq \lim_{m \rightarrow \infty} \frac{4\tilde{d}_m^{2/s} + o((c_{k_m} \tilde{d}_m)^{2/s})}{\varepsilon^2 \tilde{d}_m \tilde{d}_m^{2/s}} = 0, \quad P\text{-a.s.} \end{aligned}$$

where the last limit equals zero because $c_k = o(\log a_k)$.

It only remains to prove (86). Rewriting, we express

$$\bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} - \bar{T}_{v_{\alpha_m}}^{(\tilde{d}_m)} - E_\omega^{v_{\alpha_m}} \bar{T}_{v_{\beta_m}}^{(\tilde{d}_m)} = \sum_{i=\alpha_m+1}^{\beta_m} ((\bar{T}_{v_i}^{(\tilde{d}_m)} - \bar{T}_{v_{i-1}}^{(\tilde{d}_m)}) - \mu_{i,\tilde{d}_m,\omega})$$

as the sum of independent, zero-mean random variables (quenched), and thus we need only show the Lindberg–Feller condition. That is, we need to show

$$(87) \quad \lim_{m \rightarrow \infty} \frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} \sigma_{i,\tilde{d}_m,\omega}^2 = 1, \quad P\text{-a.s.},$$

and for all $\varepsilon > 0$

$$(88) \quad \begin{aligned} & \lim_{m \rightarrow \infty} \frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} E_\omega^{v_{i-1}} [(\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega})^2 \mathbf{1}_{|\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega}| > \varepsilon \sqrt{v_{m,\omega}}}] \\ & = 0, \quad P\text{-a.s.} \end{aligned}$$

To prove (87), note that

$$\frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} \sigma_{i,\tilde{d}_m,\omega}^2 = 1 + \frac{\sum_{i=\alpha_m+1}^{\beta_m} (\sigma_{i,\tilde{d}_m,\omega}^2 - \mu_{i,\tilde{d}_m,\omega}^2)}{v_{m,\omega}}.$$

However, again by Corollary 5.6 and the Borel–Cantelli lemma, we have $\sum_{i=\alpha_m+1}^{\beta_m} (\sigma_{i,\tilde{d}_m,\omega}^2 - \mu_{i,\tilde{d}_m,\omega}^2) = o((\delta_{k_m} \tilde{d}_m)^{2/s})$. Recalling that $v_{m,\omega} \geq \tilde{d}_m \tilde{d}_m^{2/s}$ we have that (87) is proved.

To prove (88), we break the sum up into two parts depending on whether M_i is “small” or “large.” Specifically, for $\varepsilon' \in (0, \frac{1}{3})$, we decompose the sum as

$$(89) \quad \frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} E_\omega^{v_{i-1}} [(\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega})^2 \mathbf{1}_{|\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega}| > \varepsilon \sqrt{v_{m,\omega}}}] \mathbf{1}_{M_i \leq \tilde{d}_m^{(1-\varepsilon')/s}}$$

$$(90) \quad \begin{aligned} & + \frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} E_\omega^{v_{i-1}} [(\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega})^2 \mathbf{1}_{|\bar{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega}| > \varepsilon \sqrt{v_{m,\omega}}}] \\ & \quad \times \mathbf{1}_{M_i > \tilde{d}_m^{(1-\varepsilon')/s}}. \end{aligned}$$

We get an upper bound for (89) by first omitting the indicator function inside the expectation, and then expanding the sum to be up to $n_{k_m} \geq \beta_m$. Thus, (89) is bounded above by

$$\frac{1}{v_{m,\omega}} \sum_{i=\alpha_m+1}^{\beta_m} \sigma_{i,\tilde{d}_m,\omega}^2 \mathbf{1}_{M_i \leq \tilde{d}_m^{(1-\varepsilon')/s}} \leq \frac{1}{v_{m,\omega}} \sum_{i=n_{k_m-1}+1}^{n_{k_m}} \sigma_{i,\tilde{d}_m,\omega}^2 \mathbf{1}_{M_i \leq \tilde{d}_m^{(1-\varepsilon')/s}}.$$

However, since d_k grows exponentially fast, the Borel–Cantelli lemma and Lemma 5.5 give that

$$(91) \quad \sum_{i=n_{k-1}+1}^{n_k} \sigma_{i,d_k,\omega}^2 \mathbf{1}_{M_i \leq d_k^{(1-\varepsilon')/s}} = o(d_k^{2/s}).$$

Therefore, since our choice of the subsequence n_{k_m} gives that $v_{m,\omega} \geq \tilde{d}_m^{2/s}$, we have that (89) tends to zero as $m \rightarrow \infty$.

To get an upper bound for (90), first note that our choice of the subsequence n_{k_m} gives that $\varepsilon \sqrt{v_{m,\omega}} \geq \varepsilon \sqrt{\tilde{a}_m} \mu_{i,\tilde{d}_m,\omega}$ for any $i \in (\alpha_m, \beta_m]$. Thus, for m large enough, we can replace the indicators inside the expectations in (90) by the indicators of the events $\{\tilde{T}_{v_i}^{(\tilde{d}_m)} > (1 + \varepsilon \sqrt{\tilde{a}_m}) \mu_{i,\tilde{d}_m,\omega}\}$. Thus, for m large enough and $i \in (\alpha_m, \beta_m]$, we have

$$\begin{aligned} & E_{\omega}^{v_i-1} [(\tilde{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega})^2 \mathbf{1}_{|\tilde{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega}| > \varepsilon \sqrt{v_{m,\omega}}}] \\ & \leq E_{\omega}^{v_i-1} [(\tilde{T}_{v_i}^{(\tilde{d}_m)} - \mu_{i,\tilde{d}_m,\omega})^2 \mathbf{1}_{\tilde{T}_{v_i}^{(\tilde{d}_m)} > (1 + \varepsilon \sqrt{\tilde{a}_m}) \mu_{i,\tilde{d}_m,\omega}}] \\ (92) \quad & = \varepsilon^2 \tilde{a}_m \mu_{i,\tilde{d}_m,\omega}^2 P_{\omega}^{v_i-1} (\tilde{T}_{v_i}^{(\tilde{d}_m)} > (1 + \varepsilon \sqrt{\tilde{a}_m}) \mu_{i,\tilde{d}_m,\omega}) \\ & \quad + \int_{1 + \varepsilon \sqrt{\tilde{a}_m}}^{\infty} P_{\omega}^{v_i-1} (\tilde{T}_{v_i}^{(\tilde{d}_m)} > x \mu_{i,\tilde{d}_m,\omega}) 2(x - 1) \mu_{i,\tilde{d}_m,\omega}^2 dx. \end{aligned}$$

We want to use Lemma 5.9 get an upper bounds on the probabilities in the last line above. Lemma 5.9 and the Borel–Cantelli lemma give that for k large enough, $E_{\omega}^{v_i-1} (\tilde{T}_{v_i}^{(d_k)})^j \leq 2^j j! \mu_{i,d_k,\omega}^j$, for all $n_{k-1} < i \leq n_k$ such that $M_i > d_k^{(1-\varepsilon')/s}$. Multiplying by $(4\mu_{i,d_k,\omega})^{-j}$ and summing over j gives that $E_{\omega}^{v_i-1} e^{\tilde{T}_{v_i}^{(d_k)}/(4\mu_{i,d_k,\omega})} \leq 2$. Therefore, Chebyshev’s inequality gives that

$$\begin{aligned} P_{\omega}^{v_i-1} (\tilde{T}_{v_i}^{(d_k)} > x \mu_{i,d_k,\omega}) & \leq e^{-x/4} E_{\omega}^{v_i-1} e^{\tilde{T}_{v_i}^{(d_k)}/(4\mu_{i,d_k,\omega})} \\ & \leq 2e^{-x/4}. \end{aligned}$$

Thus, for all m large enough and for all i with $\alpha_m < i \leq \beta_m \leq n_{k_m}$ and $M_i >$

$\tilde{d}_m^{(1-\varepsilon')/s}$, we have from (92) that

$$\begin{aligned} & E_\omega^{v_i-1} \left[(\bar{T}_{v_i}^{\tilde{d}_m} - \mu_{i, \tilde{d}_m, \omega})^2 \mathbf{1}_{|\bar{T}_{v_i}^{\tilde{d}_m} - \mu_{i, \tilde{d}_m, \omega}| > \varepsilon \sqrt{v_{m, \omega}}} \right] \\ & \leq \varepsilon^2 \tilde{a}_m \mu_{i, \tilde{d}_m, \omega}^2 2e^{-(1+\varepsilon\sqrt{\tilde{a}_m})/4} + \int_{1+\varepsilon\sqrt{\tilde{a}_m}}^\infty 2e^{-x/4} 2(x-1) \mu_{i, \tilde{d}_m, \omega}^2 dx \\ & = (2\varepsilon^2 \tilde{a}_m + 16(4 + \varepsilon\sqrt{\tilde{a}_m})) e^{-(1+\varepsilon\sqrt{\tilde{a}_m})/4} \mu_{i, \tilde{d}_m, \omega}^2. \end{aligned}$$

Recalling the definition of $v_{m, \omega} = \sum_{i=\alpha_m+1}^{\beta_m} \mu_{i, \tilde{d}_m, \omega}^2$, we have that as $m \rightarrow \infty$, (90) is bounded above by

$$\begin{aligned} & \lim_{m \rightarrow \infty} \frac{1}{v_{m, \omega}} \sum_{i=\alpha_m+1}^{\beta_m} (2\varepsilon^2 \tilde{a}_m + 16(4 + \varepsilon\sqrt{\tilde{a}_m})) e^{-(1+\varepsilon\sqrt{\tilde{a}_m})/4} \mu_{i, \tilde{d}_m, \omega}^2 \mathbf{1}_{M_i > \tilde{d}_m^{(1-\varepsilon')/s}} \\ & \leq \lim_{m \rightarrow \infty} (2\varepsilon^2 \tilde{a}_m + 16(4 + \varepsilon\sqrt{\tilde{a}_m})) e^{-(1+\varepsilon\sqrt{\tilde{a}_m})/4} = 0. \end{aligned}$$

This completes the proof of (88), and thus of Lemma 5.12. \square

PROOF OF THEOREM 1.3. Note first that from Lemma 4.2 and the Borel–Cantelli lemma, we have that for any $\varepsilon > 0$, $E_\omega T_{v_{n_k}} = o(n_k^{(1+\varepsilon)/s})$, P -a.s. This is equivalent to

$$(93) \quad \limsup_{k \rightarrow \infty} \frac{\log E_\omega T_{v_{n_k}}}{\log n_k} \leq \frac{1}{s}, \quad P\text{-a.s.}$$

We can also get bounds on the probability of $E_\omega T_{v_n}$ being small. Since $E_\omega^{v_i-1} T_{v_i} \geq M_i$, we have

$P(E_\omega T_{v_n} \leq n^{(1-\varepsilon)/s}) \leq P(M_i \leq n^{(1-\varepsilon)/s}, \forall i \leq n) \leq (1 - P(M_1 > n^{(1-\varepsilon)/s}))^n$, and since $P(M_1 > n^{(1-\varepsilon)/s}) \sim C_5 n^{-1+\varepsilon}$; see (13), we have $P(E_\omega T_{v_n} \leq n^{(1-\varepsilon)/s}) \leq e^{-n^{\varepsilon/2}}$. Thus, by the Borel–Cantelli lemma, for any $\varepsilon > 0$, we have that $E_\omega T_{v_{n_k}} \geq n_k^{(1-\varepsilon)/s}$ for all k large enough, P -a.s., or equivalently

$$(94) \quad \liminf_{k \rightarrow \infty} \frac{\log E_\omega T_{v_{n_k}}}{\log n_k} \geq \frac{1}{s}, \quad P\text{-a.s.}$$

Let n_{k_m} be the subsequence specified in Theorem 5.10 and define $t_m := E_\omega T_{n_{k_m}}$. Then by (93) and (94), $\lim_{m \rightarrow \infty} \frac{\log t_m}{\log n_{k_m}} = 1/s$.

For any t , define $X_t^* := \max\{X_n : n \leq t\}$. Then for any $x \in (0, \infty)$, we have

$$\begin{aligned} P_\omega \left(\frac{X_{t_m}^*}{n_{k_m}} < x \right) &= P(X_{t_m}^* < x n_{k_m}) = P_\omega(T_{x n_{k_m}} > t_m) \\ &= P_\omega \left(\frac{T_{x n_{k_m}} - E_\omega T_{x n_{k_m}}}{\sqrt{v_{m, \omega}}} > \frac{E_\omega T_{n_{k_m}} - E_\omega T_{x n_{k_m}}}{\sqrt{v_{m, \omega}}} \right). \end{aligned}$$

Now, with notation as in Theorem 5.10, we have that for all m large enough $v_{\beta_m} < xn_{k_m} < v_{\gamma_m}$ (note that this also uses the fact that $v_n/n \rightarrow E_P v$, P -a.s.). Thus, $\frac{T_{xn_{k_m}} - E_\omega T_{xn_{k_m}}}{\sqrt{v_{m,\omega}}} \xrightarrow{\mathcal{D}_\omega} Z \sim N(0, 1)$. Then we will have proved that $\lim_{m \rightarrow \infty} P_\omega\left(\frac{X_{i_m}^*}{n_{k_m}} < x\right) = \frac{1}{2}$ for any $x \in (0, \infty)$, if we can show

$$(95) \quad \lim_{m \rightarrow \infty} \frac{E_\omega T_{n_{k_m}} - E_\omega T_{xn_{k_m}}}{\sqrt{v_{m,\omega}}} = 0, \quad P\text{-a.s.}$$

For m large enough, we have $n_{k_m}, xn_{k_m} \in (v_{\beta_m}, v_{\gamma_m})$. Thus, for m large enough,

$$\begin{aligned} \left| \frac{E_\omega T_{xn_{k_m}} - E_\omega T_{n_{k_m}}}{\sqrt{v_{m,\omega}}} \right| &\leq \frac{E_\omega^{v_{\beta_m}} T_{v_{\gamma_m}}}{\sqrt{v_{m,\omega}}} \\ &= \frac{1}{\sqrt{v_{m,\omega}}} \left(E_\omega^{v_{\beta_m}} (T_{v_{\gamma_m}} - \bar{T}_{v_{\gamma_m}}^{(\tilde{d}_m)}) + \sum_{i=\beta_m+1}^{\gamma_m} \mu_{i, \tilde{d}_m, \omega} \right). \end{aligned}$$

Since $\alpha_m \leq \beta_m \leq \gamma_m \leq n_{k_m+1}$ for all m large enough, we can apply (83) to get

$$\lim_{m \rightarrow \infty} E_\omega^{v_{\beta_m}} (T_{v_{\gamma_m}} - \bar{T}_{v_{\gamma_m}}^{(\tilde{d}_m)}) \leq \lim_{m \rightarrow \infty} E_\omega^{v_{\alpha_m}} (T_{v_{n_{k_m+1}}} - \bar{T}_{v_{n_{k_m+1}}}^{(\tilde{d}_m)}) = 0.$$

Also, from our choice of n_{k_m} we have that $\sum_{i=\beta_m+1}^{\gamma_m} \mu_{i, \tilde{d}_m, \omega} \leq 2\tilde{d}_m^{1/s}$ and $v_{m,\omega} \geq \tilde{a}_m \tilde{d}_m^{2/s}$. Thus (95) is proved. Therefore

$$\lim_{m \rightarrow \infty} P_\omega\left(\frac{X_{i_m}^*}{n_{k_m}} \leq x\right) = \frac{1}{2} \quad \forall x \in (0, \infty),$$

and obviously $\lim_{m \rightarrow \infty} P_\omega\left(\frac{X_{i_m}^*}{n_{k_m}} < 0\right) = 0$ since X_n is transient to the right \mathbb{P} -a.s. due to Assumption 1. Finally, note that

$$\frac{X_t^* - X_t}{\log^2 t} = \frac{X_t^* - v_{N_t}}{\log^2 t} + \frac{v_{N_t} - X_t}{\log^2 t} \leq \frac{\max_{i \leq t} (v_i - v_{i-1})}{\log^2 t} + \frac{v_{N_t} - X_t}{\log^2 t}.$$

However, Lemma 4.6 and an easy application of Lemma 2.1 and the Borel–Cantelli lemma gives that

$$\lim_{t \rightarrow \infty} \frac{X_t^* - X_t}{\log^2 t} = 0, \quad \mathbb{P}\text{-a.s.}$$

This completes the proof of the theorem. \square

6. Asymptotics of the tail of $E_\omega T_v$. Recall that $E_\omega T_v = v + 2 \sum_{j=0}^{v-1} W_j = v + 2 \sum_{i \leq j, 0 \leq j < v} \Pi_{i,j}$, and for any $A > 1$ define

$$\sigma = \sigma_A = \inf\{n \geq 1 : \Pi_{0,n-1} \geq A\}.$$

Note that $\sigma - 1$ is a stopping time for the sequence $\Pi_{0,k}$. For any $A > 1$, $\{\sigma > \nu\} = \{M_1 < A\}$. Thus, we have by (15) that for any $A > 1$,

$$(96) \quad Q(E_\omega T_\nu > x, \sigma > \nu) = Q(E_\omega T_\nu > x, M_1 < A) = o(x^{-s}).$$

Thus, we may focus on the tail estimates $Q(E_\omega T_\nu > x, \sigma < \nu)$ in which case we can use the following expansion of $E_\omega T_\nu$:

$$(97) \quad \begin{aligned} E_\omega T_\nu &= \nu + 2 \sum_{i < 0 \leq j < \sigma-1} \Pi_{i,j} + 2 \sum_{0 \leq i \leq j < \sigma-1} \Pi_{i,j} \\ &\quad + 2 \sum_{\sigma \leq i \leq j < \nu} \Pi_{i,j} + 2 \sum_{i \leq \sigma-1 \leq j < \nu} \Pi_{i,j} \\ &= \nu + 2W_{-1}R_{0,\sigma-2} + 2 \sum_{j=0}^{\sigma-2} W_{0,j} \\ &\quad + 2 \sum_{i=\sigma}^{\nu-1} R_{i,\nu-1} + 2W_{\sigma-1}(1 + R_{\sigma,\nu-1}). \end{aligned}$$

We will show that the dominant term in (97) is the last term: $2W_{\sigma-1}(1 + R_{\sigma,\nu-1})$. A few easy consequences of Lemmas 2.1 and 2.2 are that the tails of the first three terms in the expansion (97) are negligible. The following statements are true for any $\delta > 0$ and any $A > 1$:

$$(98) \quad Q(\nu > \delta x) = P(\nu > \delta x) = o(x^{-s}),$$

$$(99) \quad \begin{aligned} Q(2W_{-1}R_{0,\sigma-2} > \delta x, \sigma < \nu) &\leq Q(W_{-1} > \sqrt{\delta x}) \\ &\quad + P(2R_{0,\sigma-2} > \sqrt{\delta x}, \sigma < \nu) \\ &\leq Q(W_{-1} > \sqrt{\delta x}) + P(2\nu A > \sqrt{\delta x}) \\ &= o(x^{-s}), \end{aligned}$$

$$(100) \quad \begin{aligned} Q\left(2 \sum_{j=0}^{\sigma-2} W_{0,j} > \delta x, \sigma < \nu\right) &\leq P\left(2 \sum_{j=1}^{\sigma-1} jA > \delta x, \sigma < \nu\right) \\ &\leq P(\nu^2 A > \delta x) = o(x^{-s}). \end{aligned}$$

In the first inequality in (100), we used the fact that $\Pi_{i,j} \leq \Pi_{0,j}$ for any $0 < i < \nu$ since $\Pi_{0,i-1} \geq 1$.

The fourth term in (97) is not negligible, but we can make it arbitrarily small by taking A large enough.

LEMMA 6.1. *For all $\delta > 0$, there exists an $A_0 = A_0(\delta) < \infty$ such that*

$$P\left(2 \sum_{\sigma A \leq i < \nu} R_{i,\nu-1} > \delta x\right) < \delta x^{-s} \quad \forall A \geq A_0(\delta).$$

PROOF. This proof is essentially a copy of the proof of Lemma 3 in [8].

$$\begin{aligned} P\left(2 \sum_{\sigma_A \leq i < \nu} R_{i, \nu-1} > \delta x\right) &\leq P\left(\sum_{\sigma_A \leq i < \nu} R_i > \frac{\delta}{2}x\right) \\ &= P\left(\sum_{i=1}^{\infty} \mathbf{1}_{\sigma_A \leq i < \nu} R_i > \frac{\delta}{2}x \frac{6}{\pi^2} \sum_{i=1}^{\infty} i^{-2}\right) \\ &\leq \sum_{i=1}^{\infty} P\left(\mathbf{1}_{\sigma_A \leq i < \nu} R_i > x \frac{3\delta}{\pi^2} i^{-2}\right). \end{aligned}$$

However, since the event $\{\sigma_A \leq i < \nu\}$ depends only on ρ_j for $j < i$, and R_i depends only on ρ_j for $j \geq i$, we have that

$$P\left(2 \sum_{\sigma_A \leq i < \nu} R_{i, \nu-1} > \delta x\right) \leq \sum_{i=1}^{\infty} P(\sigma_A \leq i < \nu) P\left(R_i > x \frac{3\delta}{\pi^2} i^{-2}\right).$$

Now, from (11), we have that there exists a $K_1 > 0$ such that $P(R_0 > x) \leq K_1 x^{-s}$ for all $x > 0$. We then conclude that

$$\begin{aligned} (101) \quad P\left(\sum_{\sigma_A \leq i < \nu} R_{i, \nu-1} > \delta x\right) &\leq K_1 \left(\frac{3\delta}{\pi^2}\right)^{-s} x^{-s} \sum_{i=1}^{\infty} P(\sigma_A \leq i < \nu) i^{2s} \\ &= K_1 \left(\frac{3\delta}{\pi^2}\right)^{-s} x^{-s} E_P \left[\sum_{i=1}^{\infty} \mathbf{1}_{\sigma_A \leq i < \nu} i^{2s} \right] \\ &\leq K_1 \left(\frac{3\delta}{\pi^2}\right)^{-s} x^{-s} E_P [v^{2s+1} \mathbf{1}_{\sigma_A < \nu}]. \end{aligned}$$

Since $E_P v^{2s+1} < \infty$ and $\lim_{A \rightarrow \infty} P(\sigma_A < \nu) = 0$, we have that the right side of (101) can be made less than δx^{-s} by choosing A large enough. \square

We need one more lemma before analyzing the dominant term in (97).

LEMMA 6.2. $E_Q[W_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] < \infty$ for any $A > 1$.

PROOF. First, note that on the event $\{\sigma_A < \nu\}$, we have that $\Pi_{i, \sigma_A-1} \leq \Pi_{0, \sigma_A-1}$ for any $i \in [0, \sigma_A)$. Thus,

$$W_{\sigma_A-1} = W_{0, \sigma_A-1} + \Pi_{0, \sigma_A-1} W_{-1} \leq (\sigma_A + W_{-1}) \Pi_{0, \sigma_A-1}.$$

Also, note that $\Pi_{0, \sigma_A-1} \leq A \rho_{\sigma_A-1}$ by the definition of σ_A . Therefore,

$$E_Q[W_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] \leq E_Q[(\sigma_A + W_{-1})^s A^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}].$$

Therefore, it is enough to prove that both $E_Q[W_{-1}^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}]$ and $E_Q[\sigma_A^s \times \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}]$ are finite (note that this is trivial if we assume that ρ has bounded support). Since W_{-1} is independent of $\rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}$ we have that

$$E_Q[W_{-1}^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] = E_Q[W_{-1}^s] E_P[\rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}],$$

where we may take the second expectation over P instead of Q because the random variable only depends on the environment to the right of zero. By Lemma 2.2, we have that $E_Q[W_{-1}^s] < \infty$. Also, $E_P[\rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] \leq E_P[\sigma_A^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}]$, and so the lemma will be proved once we prove the latter is finite. However,

$$E_P[\sigma_A^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] = \sum_{k=1}^{\infty} E_P[k^s \rho_{k-1}^s \mathbf{1}_{\sigma_A=k < \nu}] \leq \sum_{k=1}^{\infty} k^s E_P[\rho_{k-1}^s \mathbf{1}_{k \leq \nu}],$$

and since the event $\{k \leq \nu\}$ depends only on $(\rho_0, \rho_1, \dots, \rho_{k-2})$ we have that $E_P[\rho_{k-1}^s \mathbf{1}_{k \leq \nu}] = E_P \rho^s P(\nu \geq k)$ since P is a product measure. Then since $E_P \rho^s = 1$, we have that

$$E_P[\sigma_A^s \rho_{\sigma_A-1}^s \mathbf{1}_{\sigma_A < \nu}] \leq \sum_{k=1}^{\infty} k^s P(\nu \geq k).$$

This last sum is finite by Lemma 2.1. \square

Finally, we turn to the asymptotics of the tail of $2W_{\sigma-1}(1 + R_{\sigma, \nu-1})$, which is the dominant term in (97).

LEMMA 6.3. *For any $A > 1$, there exists a constant $K_A \in (0, \infty)$ such that*

$$\lim_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) = K_A.$$

PROOF. The strategy of the proof is as follows. First, note that on the event $\{\sigma < \nu\}$ we have $W_{\sigma-1}(1 + R_{\sigma}) = W_{\sigma-1}(1 + R_{\sigma, \nu-1}) + W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu}$. We will begin by analyzing the asymptotics of the tails of $W_{\sigma-1}(1 + R_{\sigma})$ and $W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu}$. Next, we will show that $W_{\sigma-1}(1 + R_{\sigma, \nu-1})$ and $W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu}$ are essentially independent in the sense that they cannot both be large. This will allow us to use the asymptotics of the tails of $W_{\sigma-1}(1 + R_{\sigma})$ and $W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu}$ to compute the asymptotics of the tails of $W_{\sigma-1}(1 + R_{\sigma, \nu-1})$.

To analyze the asymptotics of the tail of $W_{\sigma-1}(1 + R_{\sigma})$, we first recall from (11) that there exists a $K > 0$ such that $P(R_{\sigma} > x) \sim Kx^{-s}$. Let $\mathcal{F}_{\sigma-1} = \sigma(\dots, \omega_{\sigma-2}, \omega_{\sigma-1})$ be the σ -algebra generated by the environment to the left of σ . Then on the event $\{\sigma < \infty\}$, R_{σ} has the same distribution as R_0 and is independent of $\mathcal{F}_{\sigma-1}$. Thus,

$$\begin{aligned} & \lim_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma}) > x, \sigma < \nu) \\ (102) \quad &= \lim_{x \rightarrow \infty} E_Q \left[x^s Q \left(1 + R_{\sigma} > \frac{x}{W_{\sigma-1}}, \sigma < \nu \mid \mathcal{F}_{\sigma-1} \right) \right] \\ &= K E_Q[W_{\sigma-1}^s \mathbf{1}_{\sigma < \nu}]. \end{aligned}$$

A similar calculation yields

$$\begin{aligned}
 & \lim_{x \rightarrow \infty} x^s Q(W_{\sigma-1} \Pi_{\sigma, \nu-1} R_\nu > x, \sigma < \nu) \\
 (103) \quad &= \lim_{x \rightarrow \infty} E_Q \left[x^s Q \left(R_\nu > \frac{x}{W_{\sigma-1} \Pi_{\sigma, \nu-1}}, \sigma < \nu \mid \mathcal{F}_{\nu-1} \right) \right] \\
 &= E_Q [W_{\sigma-1}^s \Pi_{\sigma, \nu-1}^s \mathbf{1}_{\sigma < \nu}] K.
 \end{aligned}$$

Next, we wish to show that

$$(104) \quad \lim_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > \varepsilon x, W_{\sigma-1} \Pi_{\sigma, \nu-1} R_\nu > \varepsilon x, \sigma < \nu) = 0.$$

Since $\Pi_{\sigma, \nu-1} < \frac{1}{A}$ on the event $\{\sigma < \nu\}$, we have for any $\varepsilon > 0$ that

$$\begin{aligned}
 & x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > \varepsilon x, W_{\sigma-1} \Pi_{\sigma, \nu-1} R_\nu > \varepsilon x, \sigma < \nu) \\
 & \leq x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > \varepsilon x, W_{\sigma-1} R_\nu > A \varepsilon x, \sigma < \nu) \\
 (105) \quad &= x^s E_Q \left[Q \left(1 + R_{\sigma, \nu-1} > \frac{\varepsilon x}{W_{\sigma-1}} \mid \mathcal{F}_{\sigma-1} \right) \right. \\
 & \quad \times Q \left(R_\nu > A \frac{\varepsilon x}{W_{\sigma-1}} \mid \mathcal{F}_{\sigma-1} \right) \mathbf{1}_{\sigma < \nu} \left. \right] \\
 & \leq E_Q \left[x^s Q \left(1 + R_\sigma > \frac{\varepsilon x}{W_{\sigma-1}} \mid \mathcal{F}_{\sigma-1} \right) \right. \\
 & \quad \times Q \left(R_\nu > A \frac{\varepsilon x}{W_{\sigma-1}} \mid \mathcal{F}_{\sigma-1} \right) \mathbf{1}_{\sigma < \nu} \left. \right],
 \end{aligned}$$

where the equality on the third line is because $R_{\sigma, \nu-1}$ and R_ν are independent when $\sigma < \nu$ (note that $\{\sigma < \nu\} \in \mathcal{F}_{\sigma-1}$), and the last inequality is because $R_{\sigma, \nu-1} \leq R_\sigma$. Now, conditioned on $\mathcal{F}_{\sigma-1}$, R_σ and R_ν have the same distribution as R_0 . Then since by (11) for any $\gamma \leq s$, there exists a $K_\gamma > 0$ such that $P(1 + R_0 > x) \leq K_\gamma x^{-\gamma}$, we have that the integrand in (105) is bounded above by $K_\gamma^2 \varepsilon^{-2\gamma} W_{\sigma-1}^{2\gamma} \mathbf{1}_{\sigma < \nu} x^{s-2\gamma}$, Q -a.s. Choosing $\gamma = \frac{s}{2}$ gives that the integrand in (105) is Q -a.s. bounded above by $K_{\frac{s}{2}}^2 \varepsilon^{-s} W_{\sigma-1}^s \mathbf{1}_{\sigma < \nu}$ which by Lemma 6.2 has finite mean. However, if we choose $\gamma = s$, then we get that the integrand of (105) tends to zero Q -a.s. as $x \rightarrow \infty$. Thus, by the dominated convergence theorem, we have that (104) holds.

Now, since $R_\sigma = R_{\sigma, \nu-1} + \Pi_{\sigma, \nu-1} R_\nu$, we have that for any $\varepsilon > 0$,

$$\begin{aligned}
 & Q(W_{\sigma-1}(1 + R_\sigma) > (1 + \varepsilon)x, \sigma < \nu) \\
 & \leq Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > \varepsilon x, W_{\sigma-1} \Pi_{\sigma, \nu-1} R_\nu > \varepsilon x, \sigma < \nu) \\
 & \quad + Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) + Q(W_{\sigma-1} \Pi_{\sigma, \nu-1} R_\nu > x, \sigma < \nu).
 \end{aligned}$$

Applying (102), (103) and (104), we get that for any $\varepsilon > 0$,

$$(106) \quad \begin{aligned} & \liminf_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) \\ & \geq KE_Q[W_{\sigma-1}^s \mathbf{1}_{\sigma < \nu}](1 + \varepsilon)^{-s} - KE_Q[W_{\sigma-1}^s \Pi_{\sigma, \nu-1}^s \mathbf{1}_{\sigma < \nu}]. \end{aligned}$$

Similarly, for a bound in the other direction, we have

$$\begin{aligned} & Q(W_{\sigma-1}(1 + R_{\sigma}) > x, \sigma < \nu) \\ & \geq Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \text{ or } W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu} > x, \sigma < \nu) \\ & = Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) + Q(W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu} > x, \sigma < \nu) \\ & \quad - Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, W_{\sigma-1} \Pi_{\sigma, \nu-1} R_{\nu} > x, \sigma < \nu). \end{aligned}$$

Thus, again applying (102), (103) and (104), we get

$$(107) \quad \begin{aligned} & \limsup_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) \\ & \leq KE_Q[W_{\sigma-1}^s \mathbf{1}_{\sigma < \nu}] - KE_Q[W_{\sigma-1}^s \Pi_{\sigma, \nu-1}^s \mathbf{1}_{\sigma < \nu}]. \end{aligned}$$

Finally, applying (106) and (107) and letting $\varepsilon \rightarrow 0$, we get that

$$\begin{aligned} & \lim_{x \rightarrow \infty} x^s Q(W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu) \\ & = KE_Q[W_{\sigma-1}^s (1 - \Pi_{\sigma, \nu-1}^s) \mathbf{1}_{\sigma < \nu}] =: K_A, \end{aligned}$$

and $K_A \in (0, \infty)$ by Lemma 6.2 and the fact that $1 - \Pi_{\sigma, \nu-1} \in (1 - \frac{1}{A}, 1)$. \square

Finally, we are ready to analyze the tail of $E_{\omega} T_{\nu}$ under the measure Q .

PROOF OF THEOREM 1.4. Let $\delta > 0$, and choose $A \geq A_0(\delta)$ as in Lemma 6.1. Then using (97), we have

$$\begin{aligned} Q(E_{\omega} T_{\nu} > x) &= Q(E_{\omega} T_{\nu} > x, \sigma > \nu) + Q(E_{\omega} T_{\nu} > x, \sigma < \nu) \\ &\leq Q(E_{\omega} T_{\nu} > x, \sigma > \nu) + Q(\nu > \delta x) \\ &\quad + Q(2W_{-1} R_{0, \sigma-2} > \delta x, \sigma < \nu) \\ &\quad + Q\left(2 \sum_{j=0}^{\sigma-2} W_{0, j} > \delta x, \sigma < \nu\right) + Q\left(2 \sum_{\sigma \leq i < \nu} R_{i, \nu-1} > \delta x\right) \\ &\quad + Q(2W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > (1 - 4\delta)x, \sigma < \nu). \end{aligned}$$

Thus, combining equations (96), (98), (99) and (100) and Lemmas 6.1 and 6.3, we get that

$$(108) \quad \limsup_{x \rightarrow \infty} x^s Q(E_{\omega} T_{\nu} > x) \leq \delta + 2^s K_A (1 - 4\delta)^{-s}.$$

The lower bound is easier, since $Q(E_\omega T_\nu > x) \geq Q(2W_{\sigma-1}(1 + R_{\sigma, \nu-1}) > x, \sigma < \nu)$. Thus,

$$(109) \quad \liminf_{x \rightarrow \infty} x^s Q(E_\omega T_\nu > x) \geq 2^s K_A.$$

From (108) and (109), we get that $\overline{K} := \limsup_{A \rightarrow \infty} 2^s K_A < \infty$. Therefore, letting $\underline{K} := \liminf_{A \rightarrow \infty} 2^s K_A$, we have from (108) and (109) that

$$\overline{K} \leq \liminf_{x \rightarrow \infty} x^s Q(E_\omega T_\nu > x) \leq \limsup_{x \rightarrow \infty} x^s Q(E_\omega T_\nu > x) \leq \delta + \underline{K}(1 - 4\delta)^{-s}.$$

Then letting $\delta \rightarrow 0$ completes the proof of the theorem with $K_\infty = \underline{K} = \overline{K}$. \square

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