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MEETING TIME OF INDEPENDENT RANDOM WALKS IN RANDOM ENVIRONMENT

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Abstract. We consider, in the continuous time version, γ independent random walks on \mathbb{Z}_+ in random environment in Sinai's regime. Let T_{γ} be the first meeting time of one pair of the γ random walks starting at different positions. We first show that the tail of the quenched distribution of T_{γ} , after a suitable rescaling, converges in probability, to some functional of the Brownian motion. Then we compute the law of this functional. Eventually, we obtain results about the moments of this meeting time. Being \mathbf{E}_{ω} the quenched expectation, we show that, for almost all environments ω , $\mathbf{E}_{\omega}[T_{\gamma}^{c}]$ is finite for $c < \gamma(\gamma - 1)/2$ and infinite for $c > \gamma(\gamma - 1)/2$.

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1. Introduction and results

1.1. Preliminary facts

Let $\omega = (\omega_x^+, \omega_x^-)_{x \in \mathbb{Z}_+}$ be a sequence of positive vectors. Then, consider γ independent continuous-time Markov chains $\xi_i = (\xi_i(t))_{t \geq 0}, \ 1 \leq i \leq \gamma$, on \mathbb{Z}_+ , which jump from y to y+1 with rate ω_y^+ , and to y-1 with rate ω_y^- if $y \geq 1$ and jump from 0 to 1 with rate ω_0^+ . Now, we suppose that ω is a fixed realization of an i.i.d. sequence of positive random variables. We refer to ω as the environment, and to the ξ_i -s, $1 \leq i \leq \gamma$, as the random walks in the random environment ω . We will denote by \mathbb{P}, \mathbb{E} the probability and expectation with respect to ω , and by P_ω , E_ω the (so-called "quenched") probability and expectation for random walks in the fixed environment ω . Observe that the ξ_i -s are independent under the quenched law P_ω but not in general under the annealed law $\mathbb{P} \times P_\omega$. As usual, we will use the notation P_ω^x with $\mathbf{x} = (x_1, \dots, x_\gamma)$ for the quenched law of $(\xi_1, \dots, \xi_\gamma)$ starting from $\xi_i(0) = x_i$, with $x_i < x_{i+1}$ for $1 \leq i < \gamma$. Nevertheless, for the sake of brevity, we will omit the superscript \mathbf{x} whenever the initial positions are $\xi_i(0) = i$, for $1 \leq i \leq \gamma$ (for the sake of simplicity, we decided to choose these initial positions, but our results do not depend on the initial positions of the ξ_i -s). We also formally define the meeting time of one pair of the random walks as

$$T_{\gamma} = \inf\{s > 0; \xi_i(s) = \xi_j(s), \text{ for some } 1 \le i \ne j \le \gamma\}.$$

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In this paper we study only the case of Sinai's regime, which means that the following condition is satisfied:

Condition S. We have

$$\mathbb{E} \ln \frac{\omega_0^+}{\omega_0^-} = 0, \quad \sigma^2 := \mathbb{E} \ln^2 \frac{\omega_0^+}{\omega_0^-} \in (0, +\infty).$$

The one-dimensional RWRE has been a model of constant interest in the last three decades. The first important results are due to Solomon [19], Sinai [18] and Kesten et al. [14]. Solomon showed that the first part of condition S is equivalent to recurrence for a random walk ξ on \mathbb{Z} . Sinai proved that abnormal diffusion takes place in the recurrent case: $\xi(t)$ is of order $\ln^2 t$; Kesten et al. computed the limit law of the random walk in the transient case. Since then, we know that this model presents many interesting properties. For example, the large deviation properties were studied by Greven and den Hollander [11]. The moderate deviation properties were studied by Comets and Popov in [3, 4]. Moreover, the model presents many variations of interest, for example, in [5], Comets et al. use Lyapunov functions method to study random strings and in [12] Hu and Shi use stochastic calculus to obtain results on moderate deviations for the one-dimensional diffusion in a Brownian potential, which is the continuous-space analogue of the RWRE.

More recently, in the sub-ballistic transient case, Fribergh et al. [8] worked on moderate deviations and Enriquez et al. [7] refined, in particular, the results of [14]. In Sinai's regime, Dembo et al. [6] and Gantert et al. [10] studied the properties of the local times. For surveys on the subject, the reader is referred to the lecture notes of a course by Zeitouni [20], to the book of Hughes [13] and to the stochastic calculus approach by Shi [17]. In addition, we mention that the results we obtain in Theorem 1.4 contrast with those of the analogous deterministic model which is the simple random walk on \mathbb{Z}_+ . In the case of 2 independent random walks, it is well known that $E[\tau^c] < \infty$ for c < 1/2 and $E[\tau^c] = \infty$ for c > 1/2. In the case of 3 independent random walks, we have that $E[\tau^c] < \infty$ for c < 3/2 and $E[\tau^c] = \infty$ for c > 3/2 (see for example [1]).

For technical reasons we also need the following uniform ellipticity condition:

Condition B. There exists a positive number $\kappa > 1$ such that

$$\kappa^{-1} \le \omega_0^+ \le \kappa, \qquad \kappa^{-1} \le \omega_0^- \le \kappa \quad \mathbb{P}\text{-a.s.}$$

Given the realization of the random environment ω , define the potential function for $x \in \mathbb{R}_+$, by

$$V(x) = \begin{cases} \sum_{i=0}^{[x]-1} \ln \frac{\omega_i^-}{\omega_i^+}, & x > 0, \\ 0, & x = 0 \end{cases}$$

where [x] is the integer which is the closest to x.

We will focus on estimating the tail of the quenched distribution of T_{γ} for typical configurations ω of the environment. This enables us to approximate V by Brownian motion, but it is most convenient to use the well-known Komlós-Major-Tusnády [15] strong approximation theorem. Indeed, it allows us to relate the features of long time behavior for the walk to Brownian functionals directly built on the model, simplifying much the proof of limit properties: in a possibly enlarged probability space there exist a version of our environment ω and a Brownian motion $(W(x), x \in \mathbb{R}_+)$ with diffusion constant σ (i.e., $Var(W(x)) = \sigma^2 x$), such that for some $\hat{K} > 0$

$$\mathbb{P}\left[\limsup_{x \to +\infty} \frac{|V(x) - W(x)|}{\ln x} \le \hat{K}\right] = 1. \tag{1.1}$$

The next definition is central in our construction. Although it applies to more general numerical functions defined on \mathbb{Z}_+ or \mathbb{R}_+ , we will use it only in the case of the Brownian motion W.

Definition 1.1. Let t > 1 and $m^* \in \mathbb{R}_+$. Denote $r = r(t, m^*)$ the smallest $x > m^*$ such that $W(x) \ge W(m^*) + \ln t$ (observe that as W is a Brownian motion r exists \mathbb{P} -a.s). We will say that m^* is the first t-stable

point to the right of the origin if $W(m^*) = \min_{x \in [0,r]} W(x)$ and m^* is the smallest point on \mathbb{R}_+ verifying this condition. Then, we define the other t-stable points as follows. For $m \in \mathbb{R}_+$, denote by l = l(t, m) the largest x < m such that $W(x) \ge W(m) + \ln t$ (if such an x exists). We say that m is a t-stable point (different from the first t-stable point m^*) if $m \ne m^*$, l(m,t) and r(m,t) exist and m satisfies $W(m) = \min_{x \in [l,r]} W(x)$.

For each t > 1, it is actually possible to enumerate the t-stable points in such a way that the sequence of t-stable points is \mathbb{P} -a.s. strictly increasing to infinity. Besides, observe that, according to Definition 1.1, the t-stable points are not generally integers. Thus, throughout this paper, the statement "the random walk hits a t-stable point" means that it hits the site $x \in \mathbb{Z}_+$ which is closest to the t-stable point. As a rule, real points $x \in \mathbb{R}_+$ will be replaced, if the context requires, with the closest integer, that we may still denote by the same symbol x, if no confusion can occur.

1.2. Results

Let $\hat{m}_1(t), \hat{m}_2(t), \dots, \hat{m}_{\gamma}(t)$ be the first γ t-stable points to the right of the origin and for $1 \leq i \leq \gamma - 1$

$$\hat{h}_i(t) = \underset{x \in (\hat{m}_i(t), \hat{m}_{i+1}(t))}{\arg \max} W(x).$$

We define the process $\zeta_{\gamma} = (\zeta_{\gamma}(t))_{t>1}$ by

$$\zeta_{\gamma}(t) := \sum_{i=1}^{\gamma-1} (\gamma - i) \frac{W(\hat{h}_i(t)) - W(\hat{m}_i(t))}{\ln t}.$$
 (1.2)

We obtain the following results.

Theorem 1.2. With the process ζ_{γ} defined above it holds that

$$\lim_{t \to \infty} \left| \frac{\ln \mathsf{P}_{\omega}[T_{\gamma} > t]}{\ln t} + \zeta_{\gamma}(t) \right| = 0 \quad in \; \mathbb{P}\text{-probability}.$$

Then we set out results about the distribution of $\zeta_{\gamma}(t)$.

Theorem 1.3. The distribution of $\zeta_{\gamma}(t)$ does not depend on t. Moreover, for each t > 1, the random variable $\zeta_{\gamma}(t) - \frac{\gamma(\gamma - 1)}{2}$ has density f_{γ} given by

$$f_{\gamma}(x) = \sum_{i=1}^{\gamma-1} \frac{(-1)^{\gamma-1-i} i^{\gamma-2}}{i!(\gamma-1-i)!} e^{-x/i}$$
(1.3)

for $x \geq 0$.

We finish this section by formulating the result about the quenched moments of the meeting time.

Theorem 1.4. Let c be a positive number. Then, we have \mathbb{P} -a.s.,

$$\mathbb{E}_{\omega}[T_{\gamma}^{c}] < \infty \quad \text{if } c < \frac{\gamma(\gamma - 1)}{2}$$

and

$$\mathbb{E}_{\omega}[T_{\gamma}^{c}] = \infty \quad \text{if } c > \frac{\gamma(\gamma - 1)}{2}.$$

At this point let us explain why we have chosen to work with reflected random walks on \mathbb{Z}_+ and not random walks on \mathbb{Z} . Despite the main ideas of our proofs should work for random walks on \mathbb{Z} , it is much more difficult to write them because the number of strategies to reach time t without meeting is much greater in the case of \mathbb{Z} than in the case of \mathbb{Z}_+ . For example, in Figure 1 we represented two random walks on \mathbb{Z}_+ by black and grey particles. Imagine that at some time of order smaller than t_1 , the particles are in two different t_1 -stable

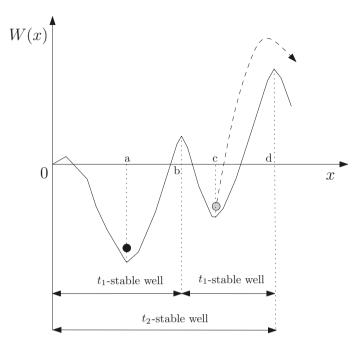


FIGURE 1. Example of a strategy for two random walks on \mathbb{Z}_+ .

wells but in the same t_2 -stable well with $t_1 < t_2 < t$, that is, $W(b) - W(a) \ge \ln t_1$, $\ln t_2 > W(b) - W(c) \ge \ln t_1$, $W(d) - W(c) \ge \ln t_1$ and $W(d) - W(a) \ge \ln t_2$. For the two particles not to meet until time t_2 , we need to expel the grey one to the right into an other t_2 -stable well. More generally, in the reflected case, we will see in the Proof of Theorem 1.4 that at each time scale of smaller order than t, we have to expel the grey particle to the right into a sufficiently deep well until the two particles are in the first two t-stable wells (indeed, these two t-stable wells are sufficiently deep to hold the particles until time t with high probability). Therefore, in some sense the strategy does not depend on the realization ω of our random environment since we always force the rightmost particle to go to the right. On Z, this not the case anymore. The possible strategies for the random walks not to meet until time t will strongly depend on the realization ω of our random environment. In Figure 2, we have two particles on \mathbb{Z} which are in two different t_1 -stable wells but in the same t_2 -stable well. In this case, we could expel the black particle to the left into an other t_2 -stable well instead of expelling the grey one to the right since the barrier of potential H_1 the black particle has to jump over is smaller than the barrier H_2 the grey particle has to jump over. Note, however, that we may want to force the grey particle out to the right because this would lead to a situation when on next steps the "cost" decreases. Thus, we have to know the exact topography of the potential V (that is the sequence of stable wells at each time scale of smaller order than t) to determine the strategy which has the cheapest cost. This leads to a great number of strategies to reach time t, which leaves the problem on \mathbb{Z} much more complicated. At the moment we have no conjecture on the precise behavior of the moments of the meeting time on Z, nevertheless we are inclined to believe that the behavior should be different from \mathbb{Z}_+ .

In the Proof of Theorem 1.2, another possible approach would be using the Karlin–McGregor formula (see for example [2]) for determining $P_{\omega}[T_{\gamma} > t]$. However, to apply this formula, one would have to obtain estimates on transition probabilities which are much finer then those obtained in [3]; with the existing estimates the leading term in the determinant in the Karlin–McGregor formula equals zero, which yields no nontrivial results.

The rest of the paper is organized as follows. In Section 2, we will characterize "good environments" and prove Lemma 2.4 that will be essential in order to prove Theorem 1.2. Section 3 is dedicated to the proofs of

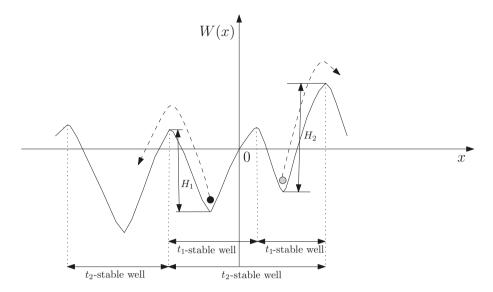


Figure 2. Example of two possible strategies for two random walks on \mathbb{Z} . $H_1 < H_2$.

Theorems 1.2 and 1.4 in the case $\gamma = 2$. As we intend to use an inductive argument to prove Theorems 1.2 and 1.4 in the general case, the preliminary treatment of the case $\gamma = 2$ is necessary. Section 4 is dedicated to the proofs of Theorems 1.2 and 1.4 in the general case. Finally, Section 5 is dedicated to the Proof of Theorem 1.3.

We mention here that all the constants that will be defined do not depend on the realization ω of the random environment but they may depend on the law of the environment and γ . Furthermore, we will denote by K'_1 , K'_2 , ..., the "local" constants, that is, those that are used only in a small neighbourhood of the place where they appear for the first time. In this case we restart the numeration at the beginning of each section.

2. Definition of t-good trajectories

First of all, we need to give more definitions. Observe that between two successive t-stable points m and m' there is always a point

$$h = \operatorname*{arg\,max}_{x \in (m,m')} W(x)$$

which is a maximum of W. Thus it is possible to separate two successive t-stable points of W. Denoting by \mathcal{S}_t the set of all the t-stable points we define

$$\mathcal{H}_t := \{h \ge 0 \text{ such that there exist } m, m' \in \mathcal{S}_t : h = \underset{x \in (m, m')}{\arg \max} W(x)\} \cup \{0\}.$$

Definition 2.1. If m is a t-stable point then we define the t-stable well of m as $\mathcal{D}_t(m) := [\max \mathcal{H}_t \cap [0, m), \min \mathcal{H}_t \cap (m, \infty))$.

Furthermore, we have the property that \mathbb{P} -a.s., for all t > 1, $\bigcup_{m \in \mathcal{S}_t} \mathcal{D}_t(m) = \mathbb{R}_+$, where the union is disjoint. We end this section by giving the definition of the elevation of a finite interval [a, b].

Definition 2.2. The *elevation* of I = [a, b] is defined as the Brownian functional

$$\mathfrak{E}(I) = \max_{x,y \in I} \max_{z \in [x,y]} \{W(z) - W(x) - W(y) + \min_{v \in I} W(v)\},\$$

where [x, y] denotes the interval with endpoints x, y regardless of x < y or x > y.

It can be seen that in the definition of $\mathfrak{E}(I)$ one may assume that y is the global minimum of W on I, x is one of the local minima, and z is one of the local maxima of W in I.

In order to prove Theorems 1.2 and 1.4, it will be convenient to exclude some particular trajectories of the Brownian motion W; that is why we will give in this section a formal definition of what we call a "t-good trajectory" of the Brownian motion W. We will see that our definition of a t-good trajectory is not very restrictive since for almost every trajectory, there exists t_0 (which depends on the trajectory of W) such that for all $t > t_0$ the trajectory is t-good. For the sake of brevity, we will not indicate in our notations the dependence on t and ω when no confusion can occur.

The following construction is rather technical, its usefulness will appear more clearly in Sections 3 and 4. In words, to describe the "optimal" strategy for the random walks not to meet until time t, we will construct a special sequence of time scales $(t^{a_n})_n$ for typical trajectories of W. To begin, we will fix a positive small threshold $\alpha \ll 1$ and will consider the first γ t^{α} -stable wells. Then, we will identify the smallest height between two successive t^{α} -stable wells. This height will be denoted by $a_1 \ln t$ and by construction we have $a_1 \geq \alpha$. Considering now the first γ t^{a_1} -stable wells, we will denote by $a_2 \ln t$ the smallest height between two successive t^{a_1} -stable wells (we have $a_2 \geq a_1$). Iterating the procedure will produce a non decreasing sequence $(t^{a_n})_{n\geq 1}$ of time scales and we will stop the procedure at the first time scale such that $a_n \geq 1$ (because then the random walks can avoid the collision for free). Observe that, for each $n \geq 1$, the configuration where the γ random walks are in the first γ t^{a_n} -stable wells avoids the meeting of two of them with high probability until a time of order t^{a_n} .

Formally, let us first take a positive decreasing function $\alpha(t)$ such that $\lim_{t\to\infty}\alpha(t)=0$ (for the moment it is not necessary to explicit α , but in Lemma 2.4 we will take $\alpha(t)=\ln^{-5/6}t$). Fix t>1 and consider the first γ $t^{\alpha(t)}$ -stable points to the right of the origin. Let us denote by $m_1(1),\ldots,m_{\gamma-1}(1)$ and $m'_{\gamma}(1)$ these γ $t^{\alpha(t)}$ -stable points. Then let us define for $1 \le j \le \gamma - 2$,

$$h_j(1) = \arg \max_{x \in (m_j(1), m_{j+1}(1))} W(x)$$

and

$$h_{\gamma-1}(1) = \arg \max_{x \in (m_{\gamma-1}(1), m'_{\gamma}(1))} W(x).$$

Then, for $1 \leq j \leq \gamma - 1$, define

$$r_j(1) = \frac{W(h_j(1)) - W(m_j(1))}{\ln t}$$

and for $1 \leq j \leq \gamma - 2$,

$$l_j(1) = \frac{W(h_j(1)) - W(m_{j+1}(1))}{\ln t}$$

and

$$l_{\gamma-1}(1) = \frac{W(h_{\gamma-1}(1)) - W(m'_{\gamma}(1))}{\ln t}.$$

Finally let,

$$a_1 = \min_{j < \gamma} \{r_j(1)\} \wedge \min_{j < \gamma} \{l_j(1)\}.$$

We will use the following rule for $n \geq 1$.

Case 1:
$$a_n = l_{\gamma-1}(n)$$

If the point $m'_{\gamma}(n)$ is t^{a_n} -stable then do the following. Rename $m'_{\gamma}(n)$ as $m_{\gamma}(n)$. Then consider the first t^{a_n} -stable point immediately after $m_{\gamma}(n)$. Let us call it $m_{\gamma+1}(n)$. Then define

$$h_{\gamma}(n) = \arg \max_{x \in (m_{\gamma}(n), m_{\gamma+1}(n))} W(x),$$

$$r_{\gamma}(n) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma}(n))}{\ln t}$$

and

$$l_{\gamma}(n) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma+1}(n))}{\ln t}.$$

Then, for $1 \le j \le \gamma - 1$, define

$$m_j(n+1) = m_j(n)$$

and

$$m_{\gamma}'(n+1) = m_{\gamma+1}(n).$$

For $1 < j < \gamma - 2$, define

$$h_j(n+1) = \arg\max_{x \in (m_j(n+1), m_{j+1}(n+1))} W(x)$$

and

$$h_{\gamma-1}(n+1) = \arg\max_{x \in (m_{\gamma-1}(n+1), m'_{\gamma}(n+1))} W(x).$$

For $1 \le j \le \gamma - 1$, we define

$$r_j(n+1) = \frac{W(h_j(n+1)) - W(m_j(n+1))}{\ln t},$$

for $1 \le j \le \gamma - 2$,

$$l_j(n+1) = \frac{W(h_j(n+1)) - W(m_{j+1}(n+1))}{\ln t}$$

and

$$l_{\gamma-1}(n+1) = \frac{W(h_{\gamma-1}(n+1)) - W(m'_{\gamma}(n+1))}{\ln t}.$$

Note that for $1 \le j \le \gamma - 2$,

$$r_j(n+1) = r_j(n)$$
$$l_j(n+1) = l_j(n)$$

and

$$r_{\gamma-1}(n+1) = r_{\gamma-1}(n) - l_{\gamma-1}(n) + r_{\gamma}(n)$$

 $l_{\gamma-1}(n+1) = l_{\gamma}(n).$

Finally define

$$a_{n+1} = \min_{j < \gamma} \{r_j(n+1)\} \wedge \min_{j < \gamma} \{l_j(n+1)\}.$$

If the point $m'_{\gamma}(n)$ is not t^{a_n} -stable, nevertheless it belongs to a t^{a_n} -stable well. Then there exists a t^{a_n} -stable point x_r such that $x_r > m'_{\gamma}(n)$ and such that $m'_{\gamma}(n) \in \mathcal{D}_{t^{a_n}}(x_r)$. Then rename x_r as $m_{\gamma}(n)$. In this case we define

$$h^*(n) = \arg \max_{x \in (m'_{\gamma}(n), m_{\gamma}(n))} W(x)$$

and

$$r_{\gamma}(n) = \frac{W(h^*(n)) - W(m'_{\gamma}(n))}{\ln t}.$$

Next, for $1 \le j \le \gamma - 1$, we define

$$m_j(n+1) = m_j(n)$$

and

$$m_{\gamma}'(n+1) = m_{\gamma}(n).$$

Define for $1 \le j \le \gamma - 2$,

$$h_j(n+1) = \arg \max_{x \in (m_j(n+1), m_{j+1}(n+1))} W(x)$$
(2.1)

and

$$h_{\gamma-1}(n+1) = \arg \max_{x \in (m_{\gamma-1}(n+1), m'_{\gamma}(n+1))} W(x).$$
(2.2)

For $1 \le j \le \gamma - 1$,

$$r_j(n+1) = \frac{W(h_j(n+1)) - W(m_j(n+1))}{\ln t}$$
(2.3)

and for $1 \le j \le \gamma - 2$,

$$l_j(n+1) = \frac{W(h_j(n+1)) - W(m_{j+1}(n+1))}{\ln t}$$
(2.4)

and

$$l_{\gamma-1}(n+1) = \frac{W(h_{\gamma-1}(n+1)) - W(m'_{\gamma}(n+1))}{\ln t}.$$
 (2.5)

Note that in this case for $j \leq \gamma - 2$,

$$r_j(n+1) = r_j(n)$$
$$l_j(n+1) = l_j(n)$$

and

$$r_{\gamma-1}(n+1) = r_{\gamma-1}(n).$$

Finally define

$$a_{n+1} = \min_{j < \gamma} \{r_j(n+1)\} \wedge \min_{j < \gamma} \{l_j(n+1)\}.$$

Case 2: $a_n \neq l_{\gamma-1}(n)$

If the point $m'_{\gamma}(n)$ is t^{a_n} -stable then do the following. Rename $m'_{\gamma}(n)$ as $m_{\gamma}(n)$. Then consider the first t^{a_n} -stable point immediately after $m_{\gamma}(n)$. Let us call it $m_{\gamma+1}(n)$. Then define

$$h_{\gamma}(n) = \arg \max_{x \in (m_{\gamma}(n), m_{\gamma+1}(n))} W(x),$$

$$r_{\gamma}(n) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma}(n))}{\ln t}$$

and

$$l_{\gamma}(n) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma+1}(n))}{\ln t}.$$

If the point $m'_{\gamma}(n)$ is not t^{a_n} -stable, nevertheless it belongs to a t^{a_n} -stable well. Hence there exists a t^{a_n} -stable point x_r such that $x_r > m'_{\gamma}(n)$ and such that $m'_{\gamma}(n) \in \mathcal{D}_{t^{a_n}}(x_r)$ then consider the first t^{a_n} -stable point immediately after x_r . Call it $m_{\gamma+1}(n)$. Let us also rename x_r as $m_{\gamma}(n)$. Then let us define

$$h_{\gamma}(n) = \arg \max_{x \in (m_{\gamma}(n), m_{\gamma+1}(n))} W(x)$$

and

$$r_{\gamma}(n) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma}(n))}{\ln t}.$$

Next, we have to distinguish two cases.

If $a_n = r_i(n)$ for some $1 \le i \le \gamma - 1$, we define the following.

For j < i,

$$m_j(n+1) = m_j(n).$$

For $i \leq j \leq \gamma - 1$

$$m_j(n+1) = m_{j+1}(n)$$

and

$$m_{\gamma}'(n+1) = m_{\gamma+1}(n).$$

Then for $1 \le j \le \gamma - 1$, define $h_j(n+1)$, $r_j(n+1)$ and $l_j(n+1)$ as in (2.1)-(2.5). Observe that in this case for j < i - 1,

$$l_j(n+1) = l_j(n),$$

 $l_{i-1}(n+1) = l_{i-1}(n) - r_i(n) + l_i(n),$

for $i \leq j \leq \gamma - 3$,

$$l_i(n+1) = l_{i+1}(n)$$

and

$$l_{\gamma-2}(n+1) = \frac{W(h_{\gamma-1}(n)) - W(m_{\gamma}(n))}{\ln t},$$

$$l_{\gamma-1}(n+1) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma+1}(n))}{\ln t}.$$

Besides, for j < i,

$$r_j(n+1) = r_j(n)$$

and for $i \leq j \leq \gamma - 1$

$$r_j(n+1) = r_{j+1}(n).$$

If $a_n = l_i(n)$ for some $1 \le i \le \gamma - 2$, we define the following quantities. For $j \le i$,

$$m_i(n+1) = m_i(n),$$

for $i < j \le \gamma - 1$,

$$m_i(n+1) = m_{i+1}(n)$$

and

$$m_{\gamma}'(n+1) = m_{\gamma+1}(n).$$

Then for $1 \le j \le \gamma - 1$, define $h_j(n+1)$, $r_j(n+1)$ and $l_j(n+1)$ as in (2.1)-(2.5). Observe that in this case, for j < i,

$$r_j(n+1) = r_j(n),$$

 $r_i(n+1) = r_i(n) - l_i(n) + r_{i+1}(n),$

and for $i < j \le \gamma - 1$,

$$r_i(n+1) = r_{i+1}(n).$$

For i < i,

$$l_i(n+1) = l_i(n),$$

for $i < j < \gamma - 3$,

$$l_i(n+1) = l_{i+1}(n)$$

and

$$l_{\gamma-2}(n+1) = \frac{W(h_{\gamma-1}(n)) - W(m_{\gamma}(n))}{\ln t},$$

$$l_{\gamma-1}(n+1) = \frac{W(h_{\gamma}(n)) - W(m_{\gamma+1}(n))}{\ln t}.$$

Finally define

$$a_{n+1} = \min_{j < \gamma} \left\{ r_j(n+1) \right\} \wedge \min_{j < \gamma} \left\{ l_j(n+1) \right\}.$$

Thus by construction, we obtain a non decreasing sequence $(a_n)_{n>1}$, \mathbb{P} -a.s. In the sequel, we will even see that the sequence reaches 1 very quickly. Hence, heuristically, the sequence $(a_n)_{n\geq 1}$ will allow us to approximate time t by the sequence $(t^{a_n})_{n\geq 1}$. For each order of time t^{a_n} with $a_n\leq 1$, we will define a strategy for the random walks not to meet during this order of time. This argument will be developed in Sections 3 for 2 random walks and 4 for the general case. Furthermore, observe that by construction in all the cases we have that

$$\max_{1 \le j \le \gamma - 1} \{ \mathfrak{E}[m_j(n), h_j(n)] \lor \mathfrak{E}[h_j(n), m_{j+1}(n)] \} \lor \{ \mathfrak{E}[m_{\gamma - 1}(n), h_{\gamma - 1}(n)] \lor \mathfrak{E}[h_{\gamma - 1}(n), m'_{\gamma}(n)] \} \le a_{n-1} \ln t,$$
(2.6)

for $n \geq 1$, with the convention $a_0 = \alpha$. This last property will turn out to be important in the proofs of Theorems 1.2 and 1.4. Now, let us introduce $N = N(\omega, t) = \min\{n \ge 1; a_n \ge 1\}$. As, by construction, the points $m_i(N)$, $1 \le j \le \gamma$ are the first t-stable points to the right of the origin, we can give a new definition of our process ζ_{γ} of Theorem 1.2 in function of the $r_i(N)$ -s with $1 \leq i \leq \gamma - 1$. Observe that we have,

$$\zeta_{\gamma}(t) = \sum_{i=1}^{\gamma-1} (\gamma - i) r_i(N).$$
 (2.7)

Now take $0 < \varepsilon(t) < \alpha(t)$ and for $1 \le k \le \gamma$, let $J_k^1(n)$ the number of $t^{a_n-\varepsilon}$ -stable wells in the interval $[m_k(n), h_k(n)]$ and $J_k^2(n)$ the number of $t^{a_n-\varepsilon}$ -stable wells in $[h_k(n), m_{k+1}(n)]$ (if $h_{\gamma}(n)$ is not defined we pose $J_{\gamma}^{1}(n)=0$ and $J_{\gamma}^{2}(n)=0$). We define the set of t-good trajectories of W in the following way. For the sake of brevity, we will use the notations $\ln_2 t := \ln \ln t$ and $\ln_3 t := \ln \ln \ln t$.

Definition 2.3. Fix $t > e^e$. Let A and B be two constants (they do not depend on ω). A realization ω of the Brownian motion W is called t-good, if

- (i) $N = N(\omega, t) \le A \ln_2 t;$
- (ii) $m'_{\gamma}(N) \leq B \ln_3 t \ln^2 t$;
- (iii) $m'_{\alpha}(1) < \ln^{1/2} t$;
- (iv) $\sum_{k=1}^{l_1} J_k^1(i) + J_k^2(i) \le \ln_2 t \text{ for } 1 \le i \le N;$ (v) $\max\{|W(x)|; x \in [0, B \ln_3 t \ln^2 t]\} \le \ln_3 t \ln t.$

The following lemma plays a crucial role.

Lemma 2.4. Take $\alpha(t) = \ln^{-5/6} t$ and $\varepsilon = \varepsilon(t) = \ln^{-11/12} t$ and let Λ_t be the set of ω which are t-good. We can choose A and B sufficiently large such that $\mathbb{P}[\lim_{t\to\infty} \Lambda_t] = 1$.

Proof. We start by showing that we can choose a constant A sufficiently large such that there exists a set Ω' of probability 1 such that for all $\omega \in \Omega'$ there exists $t_0(\omega)$ such that for all $t > t_0(\omega)$ the item (i) of Definition 2.3 is true. First take $\delta > 0$ and define the sequence of times $t_n = e^{(1+\delta)^n}$ for $n \geq 0$. Then we define the family of events

$$G_n = \{\omega; N(\omega, t_n) > A_1 \ln_2 t_n\}$$

for A_1 a positive constant. We now argue by contradiction that

$$\mathbb{P}\left[\limsup_{n\to\infty} G_n\right] = 0$$

for A_1 sufficiently large. Suppose that $\mathbb{P}[\limsup_{n\to\infty} G_n] > 0$. Then there exists $\omega \in \limsup_{n\to\infty} G_n$ such that there exists a subsequence of times $(t_{n_l})_{l\geq 1}$ such that

$$N(\omega, t_{n_l}) > A_1 \ln_2 t_{n_l} \tag{2.8}$$

for $l \ge 1$. By construction of our t-good environments we know that the sequence $(a_i)_{i\ge 1}$ is non decreasing. As W is a Brownian motion, we can even say that this sequence is increasing \mathbb{P} -a.s. Nevertheless it is not clear how fast it grows. After a deeper look at the construction and using Lemma 6.1 of [3], we can check that the sequence $(u_i)_{i\ge 1} = (a_{i\gamma})_{i\ge 1}$ verifies the following properties. For each t > 1 there exists a family of positive continuous i.i.d. random variables $(\varphi_i(t))_{i\ge 1}$ such that

$$u_{i+1}(t) \ge (1 + \varphi_i(t))u_i(t) \tag{2.9}$$

for $i \geq 1$. Using item (ii) of Lemma 6.1 of [3], we can even show that the random variables $(\varphi_i(t))_{i\geq 1}$ have exponential moments. Moreover, observe that the law of $(\varphi_i(t))_{i\geq 1}$ does not depend on t. Iterating (2.9) and taking the logarithm of both sides, we obtain

$$\ln u_{i+1}(t) \ge \sum_{j=1}^{i} \ln(1 + \varphi_j(t)) + \ln u_1(t)$$

for $i \ge 1$. Then defining $N' := \max\{i \ge 1; u_i < 1\}$ and using the facts that $u_{N'} < 1$ and $u_1 \ge \alpha(t)$ we obtain

$$\frac{5}{6}\ln_2 t \ge \sum_{j=1}^{N'-1}\ln(1+\varphi_j(t)). \tag{2.10}$$

Now using (2.10), (2.8) and the fact that $\gamma(N'+1) \geq N$, we obtain that for $\omega \in \limsup G_n$

$$\frac{5}{6} \ln_2 t_{n_l} \ge \sum_{j=1}^{\lfloor (A_1/2\gamma) \ln_2 t_{n_l} \rfloor} \ln(1 + \varphi_j(t_{n_l}))$$
 (2.11)

for all sufficiently large l.

On the other hand, by Cramér's theorem and the Borel-Cantelli lemma, we can deduce that

$$\frac{1}{(A_1/2\gamma)\ln_2 t_{n_l}} \sum_{j=1}^{\lfloor (A_1/2\gamma)\ln_2 t_{n_l}\rfloor} \ln(1+\varphi_j(t_{n_l})) \longrightarrow \mathbb{E}[\ln(1+\varphi_1(t_1))]$$

 \mathbb{P} -a.s., as $l \to \infty$ with $\mathbb{E}[\ln(1+\varphi_1(t_1))]$ a positive finite constant. Combining this last result with (2.11) we obtain for \mathbb{P} -a.a. $\omega \in \limsup_{n \to \infty} G_n$,

$$\frac{5}{6} \ln_2 t_{n_l} \ge \frac{A_1}{4\gamma} \mathbb{E} \left[\ln(1 + \varphi_1(t_1)) \right] \ln_2 t_{n_l}$$

for all l large enough. Taking $A_1 > \frac{10\gamma}{3\mathbb{E}[\ln(1+\varphi_1(t_1))]}$ this last relation is clearly impossible, and we showed that

$$\mathbb{P}[\limsup_{n\to\infty} G_n] = 0.$$

Now let us define the sequence of intervals $I_n = [t_n, t_{n+1}]$ for $n \geq 0$. It remains to show that we can find a new constant A sufficiently large such that \mathbb{P} -a.s., $N(\omega,t) \leq A \ln \ln t$ for t sufficiently large. This is easy if we observe that for $t \in I_n$, $N(\omega, t) \leq N(\omega, t_n) + N(\omega, t_{n+1})$ (this follows from the simple fact that that for $t \in I_n$, $[\alpha(t) \ln t, \ln t] \subset [\alpha(t_n) \ln t_n, \ln t_n] \cup [\alpha(t_{n+1}) \ln t_{n+1}, \ln t_{n+1}]$. This implies together with $\mathbb{P}[\limsup_{n \to \infty} G_n] = 0$ that \mathbb{P} -a.s., for every $t \in I_n$ $N(\omega, t) \leq 2A_1 \ln \ln t_{n+1}$ for all n sufficiently large. With this last observation it is sufficient to take $A \geq 3A_1$ to have that (i) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large.

We continue the proof of Lemma 2.4 by showing that item (ii) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large. Again consider the sequence of intervals $I_n = [t_n, t_{n+1}]$ for $n \ge 0$ and define the events

$$D_n = \{\text{there exist } \gamma \ t_{n+1}\text{-stable wells in the interval } [0, \ln^2 t_n] \}$$

for $n \geq 0$. By the scaling property of the Brownian motion the probability of D_n is independent of n. Furthermore observe that

$$D_0 \supset \bigcap_{j=1}^{2\gamma-1} \left\{ W\left(\frac{j}{2\gamma}\right) - W\left(\frac{j-1}{2\gamma}\right) < -3, W\left(\frac{j+1}{2\gamma}\right) - W\left(\frac{j}{2\gamma}\right) > 3 \right\},\,$$

for $\delta < 2$. As this last event has positive probability we obtain that there exists a positive constant C such that $\mathbb{P}[D_n] \geq C$ for $n \geq 0$ and $\delta < 2$. Then consider the events

 $E_n = \{\text{there exist } \gamma \ t_{n+1}\text{-stable wells in the interval } [0, M \ln n \ln^2 t_n] \}$

for $M > (-\ln(1-C))^{-1}$ and $n \ge 1$. By the Markov property and the fact that $\mathbb{P}[D_n] \ge C$ we obtain that

$$\mathbb{P}[E_n] \ge 1 - n^{-M \ln \frac{1}{1 - C}}$$

for $n \geq 1$. Using the Borel-Cantelli lemma, we obtain that \mathbb{P} -a.s. there exists $n_0 = n_0(\omega)$ such that for every $n > n_0$ there exist γt_{n+1} -stable wells in the interval $[0, M \ln n \ln^2 t_n]$. Now observe that for $t \in I_n$ we have the two following inequalities

$$(1+\delta)^n \le \ln t \le (1+\delta)^{n+1} \tag{2.12}$$

and

$$L_{\omega}(t_{n+1}, M \ln n \ln^2 t_n) \le L_{\omega}(t, M \ln n \ln^2 t_n) \le L_{\omega}(t_n, M \ln n \ln^2 t_n)$$
(2.13)

where $L_{\omega}(x,y)$ is the number of x-stable wells in the interval [0,y]. Inequality (2.13) follows from the fact that $S_{t_{n+1}} \subset S_t \subset S_{t_n}$. Let us take $t \in I_n$ for $n > n_0$. Then we know that there exist γt_{n+1} -stable wells in the interval $[0, M \ln n \ln^2 t_n]$. By inequality (2.13) we know that there exist γ t-stable wells in the interval $[0, M \ln n \ln^2 t_n]$. And by inequality (2.12) we know that there exists γ t-stable wells in the interval

$$[0, M[\ln_3 t - \ln_2(1+\delta)] \ln^2 t].$$

Finally, taking $\delta = e - 1$ we showed that (ii) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large. The proof of (iii) is completely equivalent to the proof of (ii).

We will now show that item (iv) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large. In fact, it is sufficient to show the following statement. Fix t>e and consider the first γ t-stable points to the right of the origin. Let us call the last one \hat{m}_{γ} . We will show that \mathbb{P} -a.s., for t sufficiently large $L_{\omega}(t^{1-\varepsilon}, \hat{m}_{\gamma}) \leq \ln_2 t$. Note that as $\varepsilon(t) \to 0$ as $t \to \infty$ we have for t sufficiently large that $L_{\omega}(t^{1-\varepsilon}, \hat{m}_{\gamma}) \leq L_{\omega}(\frac{t}{2}, \hat{m}_{\gamma})$. Then, consider the sequence of intervals I_n with $\delta < 2$. By the proof of (ii), we know that there exists a positive constant M such that \mathbb{P} -a.s. there exists $n_0 = n_0(\omega)$ such that for all $n > n_0$ there exists γ t_n -stable wells in the interval $[0, M \ln n \ln^2 t_n]$.

Now, let ψ_i be the width of the *i*-th $\frac{t_n}{2}$ -stable well divided by $(1+\delta)^{2n}$. By Lemma 2.1 of [4] we know that the variables $(\psi_i)_{i\geq 2}$ are i.i.d. with exponential tail. It follows from Cramér's theorem that there exists a positive finite constant M_1 such that

$$\mathbb{P}\left[\sum_{i=1}^{n} \psi_i < M \ln n\right] \le \exp\{-M_1 n\},\,$$

for all n > 1. Therefore, by the Borel–Cantelli lemma we have that \mathbb{P} -a.s. there exists $n_1 = n_1(\omega)$ such that for every $n > n_1$ we have $L_{\omega}(\frac{t_n}{2}, \hat{m}_{\gamma}) \leq n$. Finally, for $t \in I_n$, by (2.13) we have that $L_{\omega}(\frac{t}{2}, \hat{m}_{\gamma}) \leq n$ and by (2.12) $L_{\omega}(\frac{t}{2}, \hat{m}_{\gamma}) \leq \frac{\ln_2 t}{\ln(1+\delta)}$. Taking $\delta = e - 1$ we showed that (iv) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large.

We end the proof of Lemma 2.4 by showing that item (v) of Definition 2.3 holds for \mathbb{P} -a.a. ω for all t sufficiently large. By the law of the Iterated Logarithm for Brownian motion, we have that, \mathbb{P} -a.s., there exists $t_1(\omega)$ such that $|W(t)| \leq 2\sqrt{\sigma^2 t \ln_2 t}$, $\forall t \geq t_1(\omega)$, which at its turn implies that there exists $t_2(\omega) > t_1(\omega)$ such that $\max_{x \in [0,t]} |W(x)| \leq 2\sqrt{\sigma^2 t \ln_2 t}$, $\forall t \geq t_2(\omega)$. And finally we deduce for t sufficiently large,

$$\max_{x \in [0, B \ln_3 t \ln^2 t]} |W(x)| \le \ln_3 t \ln t$$

which ends the proof of Lemma 2.4.

3. Proofs of Theorems 1.2 and 1.4 in the case $\gamma = 2$

3.1. Some auxiliary results

We will systematically use the following notation,

$$\tau_A(\xi_i) = \inf\{t > 0; \xi_i(t) \in A\}$$

with $1 \leq i \leq \gamma$ and $A \subset \mathbb{Z}_+$. We now recall some results that we will need later.

For any integers a < x < b, the probability for a random walk ξ starting at x to reach b before a is given by:

$$P_{\omega}^{x}[\tau_{b}(\xi) < \tau_{a}(\xi)] = \frac{\sum_{y=a+1}^{x} e^{V(y)-V(a)}}{\sum_{y=a+1}^{b} e^{V(y)-V(a)}},$$
(3.1)

see e.g. Lemma 1 in [18].

We will also need the following upper bound on the probability of confinement which is derived from Proposition 4.1 of [8]. Let I = [a, b] with $0 \le a < b < \infty$ be an interval of \mathbb{Z}_+ . Then define

$$H_{+}(I) = \max_{x \in I} \left(\max_{y \in [x,b]} V(y) - \min_{y \in [a,x)} V(y) \right),$$

$$H_{-}(I) = \max_{x \in I} \left(\max_{y \in [a,x]} V(y) - \min_{y \in (x,b]} V(y) \right)$$

and let $H(I) = H_{+}(I) \wedge H_{-}(I)$. For a < x < b we have

$$P_{\omega}^{x}[\tau_{\{a,b\}}(\xi) > t] \le \exp\left\{-\frac{t}{C_{1}(b-a)^{4}e^{H(I)}}\right\}$$
(3.2)

with C_1 a positive constant.

The upper bound (3.2) can be too rough, for our purpose, if we consider an interval [a, b] = [m, m'] where m and m' are two neighboring t-stable points (i.e. $(m, m') \cap \mathcal{S}_t = \emptyset$). That is why we also need the following upper bound (see Lem. 3.1 of [3]). Let t > 1, and $I^+ := [h, m']$ and $I^- := [m, h]$ with $h = \arg\max_{x \in (m, m')} W(x)$. Let

$$\Delta_1 = m' - m,$$

$$\gamma_1 = \max_{x \in [m, m']} V(x) - \min_{x \in [m, m']} V(x).$$

For all $x \in [m, m']$, it holds on Γ_t that

$$P_{\omega}^{x}[\tau_{\{m',m\}}(\xi) > t/k] \leq \exp\left\{-t^{\frac{1}{2}(1-\mathfrak{E}(I^{+})\ln^{-1}t)} \left(C_{2}(\Delta_{1}\ln^{2K_{0}}t)^{-1}\right) - C_{3}e^{\gamma_{1}/2} \exp\left\{-\frac{\lambda(I^{+})e^{\mathfrak{E}(I^{+})}t^{\frac{1}{2}(1-\mathfrak{E}(I^{+})\ln^{-1}t)}}{2k}\right\}\right) + \exp\left\{-t^{\frac{1}{2}(1-\mathfrak{E}(I^{-})\ln^{-1}t)} \left(C_{2}(\Delta_{1}\ln^{2K_{0}}t)^{-1}\right) - C_{3}e^{\gamma_{1}/2} \exp\left\{-\frac{\lambda(I^{-})e^{\mathfrak{E}(I^{-})}t^{\frac{1}{2}(1-\mathfrak{E}(I^{-})\ln^{-1}t)}}{2k}\right\}\right), \tag{3.3}$$

where C_2 and C_3 are positive constants, $\lambda(I^+)$ (respectively $\lambda(I^-)$) is the spectral gap of the reflected random walk on the interval I^+ (respectively I^-) and k is such that $\ln k = o(\ln t)$ as $t \to \infty$.

We eventually need to estimate the cost of escaping a well to the right. Let again I = [a, b], $m = \arg\min_{x \in [a,b]} W(x)$ and suppose that $W(b) = \max_{x \in (m,b]} W(x)$. We will use the following estimate (see e.g. Lem. 3.4 in [3]), for any s > 0,

$$P_{\omega}^{m}[\tau_{b}(\xi) < s] \le C_{4} s e^{-V(b) + V(m)}$$
 (3.4)

with C_4 a positive constant.

3.2. Upper bound for $P_{\omega}[T_2 > t]$

Let us remind at this point a useful consequence of the KMT strong approximation theorem: if x is not too far from the origin, then V(x) and W(x) are rather close for the vast majority of environments. Hence, it is convenient to introduce the following set of "good" environments, and to restrict our forthcoming computations to this set. Fix an arbitrary $M_0 > 0$; for any t > e, let

$$\Gamma_t = \left\{ \omega : |V(x) - W(x)| \le K_0 \ln \ln t \; , \; x \in \left[0, \ln^{M_0} t \right] \right\},$$
(3.5)

where we can choose $K_0 \in (0, \infty)$ in such a way that for \mathbb{P} -almost all ω , it holds that $\omega \in \Gamma_t$ for all t large enough (cf. e.g. (13) of [3]).

To bound the probability distribution tail $P_{\omega}[T_2 > t]$ from above the main idea is to define a sequence of increasing stopping times $(\sigma_i)_{i\geq 0}$ such that on each interval of the form $[\sigma_i, \sigma_{i+1}]$ we can find a simple strategy for the random walks not to meet. We formalize this argument as follows.

First, for technical reasons let us find an upper bound for $P_{\omega}[T_2 > t^{1+\delta}]$ for $\delta > 0$ arbitrary instead of $P_{\omega}[T_2 > t]$. Fix t > 1 and define for $1 \le i \le N$, the following stopping times

$$\sigma_i = \inf\{s > 0; \xi_2(s) = h_1(i)\}$$

and the following events

$$B_i = \{T_2 \notin [\sigma_{i-1}, \sigma_i]\}$$

for $1 \le i \le N-1$ (with the convention $\sigma_0 = 0$) and

$$B_N = \left\{ T_2 \notin \left[\sigma_{N-1}, t^{1+\delta} \right] \right\} \cap \left\{ \sigma_{N-1} \le \frac{t^{1+\delta}}{2} \right\}.$$

To find the upper bound for $P_{\omega}[T_2 > t^{1+\delta}]$, the following decomposition is the key of our analysis

$$P_{\omega}\left[T_{2} > t^{1+\delta}\right] = P_{\omega}\left[T_{2} > t^{1+\delta}, \sigma_{N-1} \le \frac{t^{1+\delta}}{2}\right] + P_{\omega}\left[T_{2} > t^{1+\delta}, \sigma_{N-1} > \frac{t^{1+\delta}}{2}\right]. \tag{3.6}$$

As by (2.6), $\mathfrak{E}[0, h_1(N-1)] < \ln t$, by a similar argument as that we will use in Section 3.2.1 for the term $P_{\omega}[B_1, \sigma_1 \geq t^{b_1}]$, we can show that

$$P_{\omega}\left[T_{2} > t^{1+\delta}, \sigma_{N-1} > \frac{t^{1+\delta}}{2}\right] \le o\left(\exp(-\ln^{8} t)\right)$$
(3.7)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

For the upper bound of the term $P_{\omega}[T_2 > t, \sigma_{N-1} \le t^{1+\delta}/2]$ we start by noting that

$$P_{\omega} \left[T_{2} > t^{1+\delta}, \sigma_{N-1} \leq \frac{t^{1+\delta}}{2} \right] \leq P_{\omega}[B_{1} \cap \ldots \cap B_{N}]
= P_{\omega}[B_{1}] P_{\omega}[B_{2} \mid B_{1}] \ldots P_{\omega}[B_{N} \mid B_{1} \cap B_{2} \cap \ldots \cap B_{N-1}].$$
(3.8)

In the next subsections we will find upper bounds for the terms of the product of the right-hand side of (3.8).

3.2.1. Upper bound for $P_{\omega}[B_1]$

Fix t > 1 and let $\varepsilon' = \varepsilon'(t) = (28 + 8K_0) \ln_2 t \ln^{-1} t$ where K_0 is from (3.5). We bound the event B_1 from above in the following way,

$$P_{\omega}\left[B_{1}\right] \leq P_{\omega}\left[B_{1}, \sigma_{1} \geq t^{\alpha + \varepsilon'}\right] + P_{\omega}\left[\sigma_{1} < t^{\alpha + \varepsilon'}\right]. \tag{3.9}$$

From now on, for the sake of brevity we will denote $b_1 = \alpha + \varepsilon'$. In the next two paragraphs we treat both terms of the right-hand side of (3.9).

Upper bound for $P_{\omega}[\sigma_1 < t^{b_1}]$

Using the Markov property, we obtain

$$P_{\omega} \left[\sigma_1 < t^{b_1} \right] \leq P_{\omega}^{m_1(1)} \left[\tau_{h_1(1)}(\xi_2) < t^{b_1} \right].$$

Therefore, applying (3.4) to the last term, we obtain for t sufficiently large and $\omega \in \Gamma_t \cap \Lambda_t$

$$\mathsf{P}_{\omega}\left[\sigma_{1} < t^{b_{1}}\right] \leq C_{4} t^{\varepsilon'} t^{-(r_{1}(1) - \alpha)}. \tag{3.10}$$

Upper bound for $P_{\omega}[B_1, \sigma_1 \geq t^{b_1}]$

We will show that this term is negligible in comparison with (3.10) as $t \to \infty$ on the set of environments which belong to $\Gamma_t \cap \Lambda_t$. To show this, we will couple the two random walks with the random walks restricted to the interval $[0, h_1(1)]$. So, we need to consider two other processes, which are reflected versions of our random walks in random environment: let $\hat{\xi}_1$ and $\hat{\xi}_2$ be the reflected RWRE on the interval $I_1 = [0, h_1(1)]$. The processes $\hat{\xi}_1$ and $\hat{\xi}_2$ have the same jump rates as ξ_1 and ξ_2 on $[0, h_1(1)]$, but jump from h_1 to $h_1 - 1$ at rate $\omega_{h_1(1)}^-$. By construction we obtain

$$P_{\omega}\left[B_{1},\sigma_{1} \geq t^{b_{1}}\right] \leq P_{\omega}\left[\hat{\tau} > t^{b_{1}}\right],\tag{3.11}$$

where $\hat{\tau}$ is the first meeting time of the two random walks $\hat{\xi}_1$ and $\hat{\xi}_2$.

The idea is now to use the spectral properties of the reflected random walks to find an upper bound for the right-hand side of (3.11). We start by showing that, for $\omega \in \Gamma_t \cap \Lambda_t$, with probability at least $\ln^{-K} t$, with K a positive constant, the two random walks restricted to the interval I_1 will meet in a time of order $t^{\alpha + \frac{\varepsilon'}{2}}$. We will denote $b'_1 = \alpha + \frac{\varepsilon'}{2}$. First observe that

$$\mathbf{P}_{\omega} \left[\hat{\tau} \leq t^{b_1'} \right] \geq \mathbf{P}_{\omega} \left[\hat{\xi}_1(t^{b_1'}) \in \left[m_1(1), h_1(1) \right], \hat{\xi}_2(t^{b_1'}) \in \left[0, m_1(1) \right] \right].$$

As the two random walks in fixed environment are independent we have

$$\mathbf{P}_{\omega}\left[\hat{\tau} \leq t^{b_1'}\right] \geq \mathbf{P}_{\omega}\left[\hat{\xi}_1(t^{b_1'}) \in [m_1(1), h_1(1)]\right] \mathbf{P}_{\omega}\left[\hat{\xi}_2(t^{b_1'}) \in [0, m_1(1)]\right].$$

Then we can write

$$\mathbf{P}_{\omega}\left[\hat{\xi}_{1}(t^{b'_{1}}) \in [m_{1}(1), h_{1}(1)]\right] = \sum_{i=m_{1}(1)}^{h_{1}(1)} \mathbf{P}_{\omega}\left[\hat{\xi}_{1}(t^{b'_{1}}) = i\right]$$

and apply Corollary 2.1.5 in [16] to obtain

$$P_{\omega}\left[\hat{\xi}_{1}(t^{b'_{1}}) \in [m_{1}(1), h_{1}(1)]\right] \geq \sum_{i=m_{1}(1)}^{h_{1}(1)} \mu^{I_{1}}(i) - \exp(-\lambda(I_{1})t^{b'_{1}}) \sum_{i=m_{1}(1)}^{h_{1}(1)} \left(\frac{\mu^{I_{1}}(i)}{\mu^{I_{1}}(1)}\right)^{\frac{1}{2}}$$
(3.12)

where μ^{I_1} and $\lambda(I_1)$ are respectively the invariant measure and the spectral gap of the reflected random walks $\hat{\xi}_1$ and $\hat{\xi}_2$. At this point, let us define

$$U_1 := \sum_{i=m_1(1)}^{h_1(1)} \mu^{I_1}(i) \text{ and } U_2 := \sum_{i=m_1(1)}^{h_1(1)} \left(\frac{\mu^{I_1}(i)}{\mu^{I_1}(1)}\right)^{\frac{1}{2}}.$$

We can write a similar estimate for $P_{\omega}\left[\hat{\xi}_{2}(t^{b'_{1}}) \in [0, m_{1}(1)]\right]$ that is

$$\mathbb{P}_{\omega}\left[\hat{\xi}_{2}(t^{b'_{1}}) \in [0, m_{1}(1)]\right] \geq \sum_{i=0}^{m_{1}(1)} \mu^{I_{1}}(i) - \exp(-\lambda(I_{1})t^{b'_{1}}) \sum_{i=0}^{m_{1}(1)} \left(\frac{\mu^{I_{1}}(i)}{\mu^{I_{1}}(2)}\right)^{\frac{1}{2}} \\
:= V_{1} - V_{2} \exp(-\lambda(I_{1})t^{b'_{1}}). \tag{3.13}$$

Combining (3.12) and (3.13) we obtain that

$$\mathsf{P}_{\omega} \left[\hat{\tau} \le t^{b_1'} \right] \ge U_1 V_1 - \left(U_1 V_2 + U_2 V_1 \right) \exp \left(-\lambda (I_1) t^{b_1'} \right). \tag{3.14}$$

Now, we treat the term U_1V_1 . Note that if $\omega \in \Gamma_t \cap \Lambda_t$ we obtain, using (14) of [3], that

$$U_1 = \sum_{i=0}^{m_1(1)} \mu^{I_1}(i) \ge \mu^{I_1}(m_1) \ge \frac{K_1' e^{-V(m_1(1))}}{\sum_{y \in I_1} e^{-V(y)}} \ge \frac{K_1'}{(h_1(1)+1)} \ln^{-2K_0} t$$

where K_0 is the positive constant from (3.5) and K'_1 is a positive constant. Applying the same treatment to V_1 , we obtain that

$$U_1 V_1 \ge \frac{K_2'}{(h_1(1)+1)^2} \ln^{-4K_0} t$$

where K_2' is a positive constant. On Λ_t , we have that $(h_1(1)+1) \leq A \ln_3 t \ln^2 t$, so we obtain

$$U_1 V_1 \ge K_3' \ln_3^{-2} t \ln^{-4(1+K_0)} t \tag{3.15}$$

with K_3' a positive constant.

Concerning the second term of (3.14), we will see that if $\omega \in \Gamma_t \cap \Lambda_t$, it is negligible in comparison with the first one. We use the upper bound

$$U_1V_2 + U_2V_1 \le (h_1(1) + 1) \left[\frac{1}{(\mu^{I_1}(2))^{\frac{1}{2}}} + \frac{1}{(\mu^{I_1}(1))^{\frac{1}{2}}} \right].$$

Using (14) of [3], it is then elementary to show that if $\omega \in \Gamma_t \cap \Lambda_t$,

$$U_1 V_2 + U_2 V_1 \le K_4' t^{\ln_3 t} \ln^5 t \tag{3.16}$$

with K'_4 a positive constant.

On the other hand, by Proposition 3.1 of [3] and the fact that by (2.6), $\mathfrak{E}(I_1) < \alpha \ln t$, we obtain

$$\exp\left(-\lambda(I_1)t^{b_1'}\right) = o\left(\exp\left(-\ln^{3/2}t\right)\right).$$

This shows together with (3.16) that the second term of the right-hand side of (3.14) is negligible in comparison with the first one. Finally, from (3.15) and (3.16), we deduce that

$$P_{\omega}\left[\hat{\tau} \le t^{b_1'}\right] \ge \frac{K_3'}{2} \ln_3^{-2} t \ln^{-4(1+K_0)} t$$

for t large enough and $\omega \in \Gamma_t \cap \Lambda_t$.

Now, dividing the time interval $[0, t^{\alpha+\varepsilon'}]$ into (roughly speaking) $t^{\frac{\varepsilon'}{2}}$ intervals of size $t^{\alpha+\frac{\varepsilon'}{2}}$ and applying the Markov property we get that

$$\mathsf{P}_{\omega} \Big[\hat{\tau} > t^{b_1} \Big] \leq \Big(1 - K_4' \ln_3^{-2} t \ln^{-4(1+K_0)} t \Big)^{\lfloor t^{\frac{\varepsilon'}{2}} \rfloor} \leq \exp\left(-K_4' \ln^9 t \right)$$

for t large enough and $\omega \in \Gamma_t \cap \Lambda_t$. As a consequence, the term $P_{\omega}[B_1, \sigma_1 \geq t^{b_1}]$ is negligible in comparison with $P_{\omega}[\sigma_1 < t^{b_1}]$.

To sum up this subsection, we obtained

$$P_{\omega}[B_1] \le 2C_4 t^{-(r_1(1) - \alpha - \varepsilon')} = t^{-(r_1(1) + o(1))}$$
(3.17)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.2.2. Upper bounds for $P_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}]$, where $2 \leq i \leq N-1$

If $h_1(i) = h_1(i-1)$ then we obviously have $P_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}] = 1$. If $h_1(i) > h_1(i-1)$ we do the following. Let us write $b_i = a_{i-1} + \varepsilon'$. We will use the following decomposition

$$P_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}] \leq P_{\omega}[B_i \cap \{\sigma_i - \sigma_{i-1} \geq t^{b_i}\} \mid B_1 \cap \ldots \cap B_{i-1}] + P_{\omega}[\sigma_i - \sigma_{i-1} < t^{b_i} \mid B_1 \cap \ldots \cap B_{i-1}].$$

As the event $\{\sigma_i - \sigma_{i-1} < t^{b_i}\}$ is independent of $B_1 \cap \ldots \cap B_{i-1}$ we obtain

$$P_{\omega}[B_i \mid B_1 \cap \dots \cap B_{i-1}] \leq P_{\omega}[B_i \cap \{\sigma_i - \sigma_{i-1} \geq t^{b_i}\} \mid B_1 \cap \dots \cap B_{i-1}] + P_{\omega}[\sigma_i - \sigma_{i-1} < t^{b_i}].$$
(3.18)

Upper bounds for $P_{\omega}[\sigma_i - \sigma_{i-1} < t^{b_i}]$

Using the Markov property and (3.4), we obtain for $\omega \in \Gamma_t \cap \Lambda_t$

$$P_{\omega} \left[\sigma_i - \sigma_{i-1} < t^{b_i} \right] \le C_4 t^{\varepsilon'} t^{-(r_2(i-1) - a_{i-1})}. \tag{3.19}$$

Upper bounds for $P_{\omega}[B_i \cap \{\sigma_i - \sigma_{i-1} \geq t^{b_i}\} \mid B_1 \cap \ldots \cap B_{i-1}]$

As the event $B_1 \cap ... \cap B_{i-1}$ belongs to $\mathcal{F}_{\sigma_{i-1}}$ the σ -field generated by ξ_1 and ξ_2 up to the stopping time σ_{i-1} , applying the Markov property, we obtain

$$\begin{split} \mathbf{P}_{\omega}[B_{i} \cap \{\sigma_{i} - \sigma_{i-1} \geq t^{b_{i}}\} \mid B_{1} \cap \ldots \cap B_{i-1}] &= \sum_{x=0}^{h_{1}(i-1)} \mathbf{P}_{\omega} \left[B_{i} \cap \{\sigma_{i} - \sigma_{i-1} \geq t^{b_{i}}\} \mid \xi_{1}(\sigma_{i-1}) = x\right] \\ &\times \mathbf{P}_{\omega} \left[\xi_{1}(\sigma_{i-1}) = x \mid B_{1} \cap \ldots \cap B_{i-1}\right] \\ &\leq \max_{x < h_{1}(i-1)} \mathbf{P}_{\omega} \left[B_{i} \cap \{\sigma_{i} - \sigma_{i-1} \geq t^{b_{i}}\} \mid \xi_{1}(\sigma_{i-1}) = x\right]. \end{split}$$

The next step is to find an upper bound for

$$\mathbf{P}_{\omega}\left[B_{i}\cap\left\{\sigma_{i}-\sigma_{i-1}\geq t^{b_{i}}\right\}\mid\xi_{1}(\sigma_{i-1})=x\right]$$

uniformly in x for $x \in [0, h_1(i-1))$. As we did in Section 3.2.1, we can couple the two random walks with the random walks restricted to the interval $I_i = [0, h_1(i)]$. We immediately conclude that

$$\mathsf{P}_{\omega}[B_i \cap \left\{\sigma_i - \sigma_{i-1} \ge t^{b_i}\right\} \mid B_1 \cap \ldots \cap B_{i-1}] \le \exp\left(-K_4' \ln^9 t\right).$$

which is negligible in comparison with the right-hand side of (3.19).

Thus, in the case $h_1(i) > h_1(i-1)$, we obtain

$$P_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}] \le 2C_4 t^{-(r_2(i-1) - a_{i-1} - \varepsilon')} = t^{-(r_2(i-1) - a_{i-1} + o(1))}$$
(3.20)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.2.3. Upper bound for $P_{\omega}[B_N \mid B_1 \cap \ldots \cap B_{N-1}]$

If $h_1(N) = h_1(N-1)$ then we obviously have $P_{\omega}[B_N \mid B_1 \cap ... \cap B_{N-1}] = 1$.

If $h_1(N) > h_1(N-1)$ we use the following decomposition.

$$P_{\omega}\left[B_{N} \mid B_{1} \cap \ldots \cap B_{N-1}\right] \leq P_{\omega}\left[B_{N} \cap \left\{\sigma_{N} - \sigma_{N-1} \geq t^{a_{N-1} + \varepsilon'}\right\} \mid B_{1} \cap \ldots \cap B_{N-1}\right] + P_{\omega}\left[\sigma_{N} - \sigma_{N-1} < t^{a_{N-1} + \varepsilon'}\right]. \tag{3.21}$$

The upper estimate of the last term of (3.21) is easily computed.

$$P_{\omega} \left[\sigma_{N} - \sigma_{N-1} < t^{a_{N-1} + \varepsilon'} \right] \le 2C_{4} t^{-(r_{2}(N-1) - a_{N-1} - \varepsilon')}$$
(3.22)

for t large enough and $\omega \in \Gamma_t \cap \Lambda_t$.

It is not much more difficult to show that the first term of (3.21) is negligible in comparison with the last one. To this end, just notice that

$$\mathbb{P}_{\omega} \Big[B_N \cap \Big\{ \sigma_N - \sigma_{N-1} \ge t^{a_{N-1} + \varepsilon'} \Big\} \mid B_1 \cap \ldots \cap B_{N-1} \Big] \\
\le \mathbb{P}_{\omega} [T_2 \notin [\sigma_{N-1}, \sigma_N \wedge t^{1+\delta}], \sigma_N \wedge t^{1+\delta} - \sigma_{N-1} \ge t^{a_{N-1} + \varepsilon'} \mid B_1 \cap \ldots \cap B_{N-1}]$$

holds for sufficiently large t since $a_{N-1} + \varepsilon' < 1 + \delta/2$ for sufficiently large t. This last term can be bounded from above by the same method we used for the term $P_{\omega}[B_1 \cap \sigma_1 \geq t^{b_1}]$ of Section 3.2.1. A similar estimate can be computed which shows that it is negligible in comparison with (3.22).

Finally in the case $h_1(N) > h_1(N-1)$ we obtained

$$P_{\omega}[B_N \mid B_1 \cap \ldots \cap B_{N-1}] \le 2C_4 t^{-(r_2(N-1) - a_{N-1} - \varepsilon')} = t^{-(r_2(N-1) - a_{N-1} + o(1))}$$
(3.23)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.2.4. Conclusion

Using the results of the precedent sections, we now show that

$$\prod_{i=1}^{N} \mathsf{P}_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}] \le t^{-(r_1(N) + o(1))}$$
(3.24)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

We will proceed by induction. For every $i \leq 1$, consider the product

$$\Pi_i = \prod_{j=1}^i P_{\omega}[B_j \mid B_1 \cap \ldots \cap B_{j-1}].$$

Then we make the following induction hypothesis

$$\Pi_i \le t^{-(r_1(i)+o(1))}$$

for some $i \geq 1$. We will show that it implies

$$\Pi_{i+1} \le t^{-(r_1(i+1)+o(1))}.$$

Indeed we have,

$$\Pi_{i+1} = \Pi_i P_{\omega}[B_{i+1} \mid B_1 \cap \ldots \cap B_i].$$

If $h_1(i+1) = h_1(i)$ we have $P_{\omega}[B_{i+1} \mid B_1 \cap \ldots \cap B_i] = 1$ and $\Pi_{i+1} = \Pi_i$. But in this case note that by construction we have $r_1(i+1) = r_1(i)$. So, we obtain

$$\Pi_{i+1} \le t^{-(r_1(i+1)+o(1))}.$$

If $h_1(i+1) > h_1(i)$ we have by (3.20) and (3.23)

$$P_{\omega}[B_{i+1} \mid B_1 \cap \ldots \cap B_i] \le t^{-(r_2(i) - a_i + o(1))}$$

which implies with the induction hypothesis

$$\prod_{i+1} < t^{-(r_1(i)-a_i+r_2(i)+o(1))}$$

But in this case we have by construction either $r_1(i+1) = r_1(i) - a_i + r_2(i)$ or $r_1(i+1) = r_2(i)$ and $r_1(i) = a_i$. This leads to

$$\Pi_{i+1} \le t^{-(r_1(i+1)+o(1))}.$$

As the induction hypothesis is verified for i = 1 by (3.17), we showed that

$$\prod_{i=1}^{N} \mathsf{P}_{\omega}[B_i \mid B_1 \cap \ldots \cap B_{i-1}] \le t^{-(r_1(N) + o(1))}$$
(3.25)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

We are now able to deduce the first part of Theorem 1.4. Indeed, from Lemma 2.4, (3.7) and (3.25), we have \mathbb{P} -a.s.,

$$P_{\omega}[T_2 > t^{1+\delta}] \le t^{-(r_1(N) - o(1))}$$

as $t \to \infty$. We obtain \mathbb{P} -a.s.,

$$P_{\omega}[T_2 > t] \le t^{-\left(\frac{r_1(N)}{1+\delta} + o(1)\right)}$$
 (3.26)

as $t \to \infty$. We can immediately deduce the first part of Theorem 1.4. Indeed, as $r_1(N) \ge 1$, we obtain \mathbb{P} -a.s.,

$$\mathrm{P}_{\omega}[T_2 > t] \leq t^{-\left(\frac{1}{1+\delta} + o(1)\right)}$$

as $t \to \infty$. To conclude, fix c < 1, as δ is arbitrary we can choose δ sufficiently small such that $\delta < \frac{1-c}{c}$. This shows that $\mathsf{E}_{\omega}[T_2^c] < \infty$, \mathbb{P} -a.s.

Now, in order to prove Theorem 1.2, we shall improve (3.26). That is why we shall consider convergence in \mathbb{P} -probability instead of \mathbb{P} -a.s. to obtain

$$ln P_{\omega}[T_2 > t] \le -\zeta_2(t) \ln t + o(\ln t) \tag{3.27}$$

as $t \to \infty$. At this point, we can explain why we chose to bound $P_{\omega}[T_2 > t^{1+\delta}]$ instead of $P_{\omega}[T_2 > t]$. We know by definition of the elevation that $\mathfrak{E}[0,h_1(N)] < \ln t$ but when considering \mathbb{P} -a.s. convergence, we do not control the elevation and it can be arbitrarily close to $\ln t$. In order to show (3.7) and to bound the term $P_{\omega}[B_N \mid B_1 \dots B_{N-1}]$, we had to consider an order of time greater than t, that is $t^{1+\delta}$. Considering convergence in \mathbb{P} -probability things become easier since from now on, we can control the elevation of the interval $[0,h_1(N)]$ as follows. Fix t>0 and $0<\rho<1$. Let \mathcal{E}_t be the set of environments ω such that for every $\omega\in\mathcal{E}_t$ we have $\mathfrak{E}[0,h_1(N)]<(1-\rho)\ln t$. By Lemma 4.1 of [3], for any $\epsilon>0$, we can choose $\rho>0$ small enough in such a way that $\mathbb{P}[\mathcal{E}_t]>(1-\epsilon)\ln t$. Furthermore, $\mathbb{P}[\mathcal{E}_t]$ does not depend on t. Since our goal is now to prove convergence in \mathbb{P} -probability, we can restrict ourselves to the set $\mathcal{E}_t\cap\Lambda_t\cap\Gamma_t$ and bound directly $P_{\omega}[T_2>t]$ instead of $P_{\omega}[T_2>t^{1+\delta}]$. Finally, to obtain (3.27) we can repeat exactly the same computations we used to get (3.26).

3.3. Lower bound for $P_{\omega}[T_2 > t]$

In order to bound $P_{\omega}[T_2 > t]$ from below, the main idea is to find the best strategy for the two random walks not to meet until time t. To this end we define the following events. First, let us write $\tau_i^2(\xi_2)$ for the time of second return to state i for the random walk ξ_2 . Let

$$D_0 = \{ \tau_{\{0,2\}}(\xi_1) > \tau_{m_2'(1)}(\xi_2) \} \cap \{ \xi_2(u) \le \xi_2(s), 0 \le u \le s \le \tau_{m_2'(1)}(\xi_2) \}.$$

Then for $1 \leq i < N$,

$$D_i = \{\tau_{h_1(i)}(\xi_1) - \tau_{m_2'(i)}(\xi_2) > t^{a_i}\} \cap \{\tau_{m_2'(i+1)}(\xi_2) < \tau_{h_1(i)}^2(\xi_2)\} \cap \{\tau_{m_2'(i+1)}(\xi_2) - \tau_{m_2'(i)}(\xi_2) \le t^{a_i}\}$$

and finally

$$D_N = \{ \tau_{h_1(N)}(\xi_1) - \tau_{m_2'(N)}(\xi_2) > t \} \cap \{ \tau_{h_1(N)}^2(\xi_2) - \tau_{m_2'(N)}(\xi_2) > t \}.$$

On the event D_0 , we want the random walk ξ_1 to stay at position 1 until ξ_2 reaches the point $m'_2(1)$. On the events D_i , for $1 \le i < N$, we want ξ_1 to reach for the first time the point $h_1(i)$ in a time greater than t^{a_i} and we want ξ_2 to reach the point $m'_2(i+1)$ in a time less than t^{a_i} before it reaches $h_1(i)$ for the second time. Finally, on D_N , we want the random walks to stay in their respective t-stable wells until time t.

Thus we obtain

$$P_{\omega}[T_2 > t] \ge P_{\omega}[D_0 \cap \ldots \cap D_N]$$

$$\ge P_{\omega}[D_0]P_{\omega}[D_1 \mid D_0] \ldots P_{\omega}[D_N \mid D_0 \cap D_1 \cap \ldots \cap D_{N-1}]. \tag{3.28}$$

We will now find a lower bound for each term of the right-hand side of (3.28).

3.3.1. Lower bound for $P_{\omega}[D_0]$

First of all, note that by condition B, the random walks remain in mean a time smaller than $\kappa/2$ in every state. Thus, we can obtain a lower estimate of the event D_0 writing

$$D_0 \supset \{\tau_{\{0,2\}}(\xi_1) > \tau_{m'_2(1)}(\xi_2)\} \cap \{\xi_2(u) \leq \xi_2(s), 0 \leq u \leq s \leq \tau_{m'_2(1)}(\xi_2)\} \cap \left\{\tau_{m'_2(1)}(\xi_2) \leq \frac{m'_2(1)\kappa}{2}\right\}$$

which implies by independence of the two random walks

$$\mathbf{P}_{\omega}[D_0] \geq \mathbf{P}_{\omega} \left[\xi_2(u) \leq \xi_2(s), 0 \leq u \leq s \leq \tau_{m_2'(1)}(\xi_2), \tau_{m_2'(1)}(\xi_2) \leq \frac{m_2'(1)\kappa}{2} \right] \mathbf{P}_{\omega} \left[\tau_{\{0,2\}}(\xi_1) > \frac{m_2'(1)\kappa}{2} \right] \cdot \mathbf{P}_{\omega}[D_0] \leq \mathbf{P}_{\omega} \left[\tau_{\{0,2\}}(\xi_1) \leq \tau_{m_2'(1)}(\xi_2) + \tau_{m_2'(1)}(\xi_2) \leq \frac{m_2'(1)\kappa}{2} \right] \cdot \mathbf{P}_{\omega}[D_0] \leq \mathbf{P}_{\omega} \left[\tau_{\{0,2\}}(\xi_1) \leq \tau_{m_2'(1)}(\xi_2) + \tau_{m_2'(1)}(\xi_2) \leq \frac{m_2'(1)\kappa}{2} \right] \cdot \mathbf{P}_{\omega}[D_0] \leq \mathbf{P}_{\omega}[D_0]$$

Then, using condition B and the fact that jump Markov processes have independent and exponentially distributed jump time increments, we can write

$$\mathbf{P}_{\omega}[D_0] \geq \mathbf{P}_{\omega}\left[\tau_{\{0,2\}}(\xi_1) > \frac{m_2'(1)\kappa}{2}\right] \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)] \left(P_X\left[X \leq \frac{\kappa}{2}\right]\right)^{m_2'(1)} \mathbf{P}_{\omega}[\xi_2(s) \leq \xi_2(t), 0 \leq s \leq t \leq \tau_{m_2'(1)}(\xi_2)$$

where X is a random variable with exponential law of parameter $2/\kappa$ and P_X is the law of X. From the last expression we obtain

$$P_{\omega}[D_0] \ge \left(\frac{e^{-\kappa^2}(e-1)}{2e\kappa^2}\right)^{\ln^{1/2}t} \tag{3.29}$$

for $\omega \in \Gamma_t \cap \Lambda_t$. Note that, as $\kappa > 1$, we have that $\left(\frac{e^{-\kappa^2}(e^{-1})}{2e\kappa^2}\right) < 1$.

3.3.2. Lower bound for $P_{\omega}[D_1 \mid D_0]$

First, for the sake of brevity let us write $\beta_i = \tau_{m'_2(i)}(\xi_2)$ for $i \ge 1$. Using again the Markov property and the independence of the two random walks, we obtain

$$P_{\omega}[D_1 \mid D_0] = P_{\omega}^{m_2'(1)}[\tau_{m_2'(2)}(\xi_2) < \tau_{h_1(1)}(\xi_2), \tau_{m_2'(2)}(\xi_2) \le t^{a_1}]P_{\omega}^1[\tau_{h_1(1)}(\xi_1) > t^{a_1}]. \tag{3.30}$$

We first show that the second term of the right-hand side of (3.30) is greater than 1/2 for t sufficiently large and $\omega \in \Gamma_t \cap \Lambda_t$. Observe that by the Markov property and (3.4) we have

$$\mathsf{P}^{1}_{\omega}[\tau_{h_{1}(1)}(\xi_{1}) > t^{a_{1}}] \ge \mathsf{P}^{m_{1}(1)}_{\omega}[\tau_{h_{1}(1)}(\xi_{1}) > t^{a_{1}}] \ge \left(1 - C_{4}t^{a_{1}}e^{-V(h_{1}(1)) + V(m_{1}(1))}\right)^{+} \tag{3.31}$$

where $(\cdot)^+$ is the positive part. One can see that for $\omega \in \Gamma_t \cap \Lambda_t$ and t sufficiently large the right-hand side of the last inequality will be larger than 1/2. By construction, we have $W(h_1(1)) - W(m_1(1)) \ge a_1 \ln t$ so we obtain

$$C_4 t^{a_1} e^{-V(h_1(1)) + V(m_1(1))} \le C_4 \ln^{-2K_0} t \le \frac{1}{2}$$
 (3.32)

for t sufficiently large.

Now we will treat the first term of the right-hand side of (3.30). We have to consider two cases: $m'_2(1)$ is t^{a_1} -stable and $m'_2(1)$ is not t^{a_1} -stable.

Case 1: $m'_2(1)$ is t^{a_1} -stable

First using the Markov property let us write

$$\mathbf{P}_{\omega}^{m'_{2}(1)}\left[\tau_{m'_{2}(2)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{m'_{2}(2)}(\xi_{2}) \leq t^{a_{1}}\right] \geq \mathbf{P}_{\omega}^{m'_{2}(1)}\left[\tau_{h_{2}(1)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{h_{2}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{2}\right] \\
\times \mathbf{P}_{\omega}^{h_{2}(1)}\left[\tau_{m'_{2}(2)}(\xi_{2}) < \tau_{m'_{2}(1)}(\xi_{2}), \tau_{m'_{2}(2)}(\xi_{2}) \leq \frac{t^{a_{1}}}{2}\right] \\
:= R_{1} \times U_{1}. \tag{3.33}$$

Let us bound from below the first term of the right-hand side of (3.33).

Lower bound for R_1

In this case $m_2'(1) \equiv m_2(1)$. Fix t > e and let $\varepsilon = \varepsilon(t)$ from Lemma 2.4. Then define

$$\eta_2(1) = \max\{x < m_2(1); W(x) - W(m_2(1)) = (a_1 - \varepsilon) \ln t\}.$$

Furthermore, suppose that we have $J_2^1(1) = J_2^1(1)(\omega,t)$ $t^{a_1-\varepsilon}$ -stable wells in the interval $[m_2(1),h_2(1)]$. If $J_2^1(1) \geq 1$, then we define for $1 \leq i \leq J_2^1(1)$ the point $w_i(1)$ as the minimum of the ith $t^{a_1-\varepsilon}$ -stable well and

$$v_i(1) = \underset{w_i(1) < x < w_{i+1}(1)}{\arg \max} W(x).$$

We also define $v_0(1) := \arg \max_{m_2(1) < x < w_1(1)} W(x)$. See Figure 3.

If $J_2^1(1) = 0$, we can directly find a lower bound for R_1 . We write

$$R_1 \ge \mathsf{P}_{\omega}^{m_2(1)}[\tau_{h_2(1)}(\xi_2) < \tau_{\eta_2(1)}(\xi_2)] - \mathsf{P}_{\omega}^{m_2(1)}\left[\tau_{\{\eta_2(1),h_2(1)\}}(\xi_2) > \frac{t^{a_1}}{2}\right] \cdot$$

Then using (3.1) for the first term and (3.2) for the second term (noting that $H([\eta_2(1), h_2(1)]) \leq (a_1 - \varepsilon) \ln t + 2K_0 \ln_2 t$), we obtain for $\omega \in \Gamma_t \cap \Lambda_t$ and $t \to \infty$

$$R_1 \ge \frac{e^{-(W(h_2(1)) - W(\eta_2(1)))} \ln^{-2K_0} t}{h_2(1) - \eta_2(1)} - o(\exp(-t^{\frac{\varepsilon}{2}})).$$
(3.34)

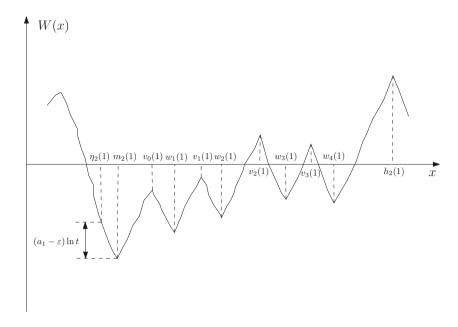


FIGURE 3. On the definition of $\eta_2(1)$, $v_i(1)$ and $w_i(1)$. Case $J_2^1(1) = 4$.

If $J_2^1(1) \geq 1$ we have a bit more work to do. By the Markov property we obtain

$$R_1 \ge \prod_{i=0}^{J_2^1(1)} \mathsf{P}_{\omega}^{w_i(1)} \left[\tau_{w_{i+1}(1)}(\xi_2) < \tau_{v_{i-1}(1)}(\xi_2), \tau_{w_{i+1}(1)}(\xi_2) \le \frac{t^{a_1}}{2(J_2^1(1)+1)} \right]$$
(3.35)

with the conventions $v_{-1}(1) = \eta_2(1)$, $w_0(1) = m_2(1)$ and $w_{J_2^1(1)+1}(1) = h_2(1)$.

Let us first bound from below the first term of the product. The other terms of the product will be treated in a similar way. First note that by the Markov property we have

$$P_{\omega}^{m_{2}(1)} \left[\tau_{w_{1}(1)}(\xi_{2}) < \tau_{\eta_{2}(1)}(\xi_{2}), \tau_{w_{1}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{2(J_{2}^{1}(1)+1)} \right] \\
\ge P_{\omega}^{m_{2}(1)} \left[\tau_{v_{0}(1)}(\xi_{2}) < \tau_{\eta_{2}(1)}(\xi_{2}), \tau_{v_{0}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{4(J_{2}^{1}(1)+1)} \right] \\
\times P_{\omega}^{v_{0}(1)} \left[\tau_{w_{1}(1)}(\xi_{2}) < \tau_{m_{2}(1)}(\xi_{2}), \tau_{w_{1}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{4(J_{2}^{1}(1)+1)} \right] \\
:= R_{2} \times R_{3}. \tag{3.36}$$

We have

$$R_2 \ge \mathsf{P}_{\omega}^{m_2(1)}[\tau_{v_0(1)}(\xi_2) < \tau_{\eta_2(1)}(\xi_2)] - \mathsf{P}_{\omega}\left[\tau_{\{v_0(1),\eta_2(1)\}}(\xi_2) > \frac{t^{a_1}}{4(J_2^1(1)+1)}\right] \cdot$$

Then, using (3.1) for the first term and (3.2) for the second term, we obtain for $\omega \in \Gamma_t$

$$R_2 \ge \frac{e^{-(W(v_0(1)) - W(\eta_2(1)))} \ln^{-2K_0} t}{v_0(1) - \eta_2(1)} - o(\exp(-t^{\frac{\varepsilon}{2}}))$$
(3.37)

as $t \to \infty$.

For R_3 we write

$$R_3 \geq \mathtt{P}_{\omega}^{v_0(1)}[\tau_{w_1(1)}(\xi_2) < \tau_{m_2(1)}(\xi_2)] - \mathtt{P}_{\omega}^{v_0(1)}\left[\tau_{\{w_1(1), m_2(1)\}}(\xi_2) > \frac{t^{a_1}}{4(J_2^1(1) + 1)}\right] \cdot$$

Analogously, using (3.1) for the first term and (3.3) for the second term, we obtain for $\omega \in \Gamma_t \cap \Lambda_t$

$$R_3 \ge \frac{\ln^{-2K_0} t}{w_1(1) - m_2(1)} - o(\exp(-t^{\frac{\varepsilon}{3}}))$$
(3.38)

as $t \to \infty$. To obtain $o(\exp(-t^{\frac{\varepsilon}{3}}))$ we used Proposition 3.1 of [3] and the facts that on $\Gamma_t \cap \Lambda_t$, $\Delta_1 \leq B \ln_3 t \ln^2 t$, $\gamma_1 \leq 2 \ln_3 t \ln t$, $J_2^1(1) \leq \ln_2 t$ and $a_1 - \frac{\mathfrak{E}(I^+) \vee \mathfrak{E}(I^-)}{\ln t} \geq \varepsilon$. We will now bound the terms of the right-hand side product of (3.35) from below, for $1 \leq i \leq J_2^1(1)$. First

note that by the Markov property we have

$$\begin{split} \mathbf{P}_{\omega}^{w_{i}(1)} &\left[\tau_{w_{i+1}(1)}(\xi_{2}) < \tau_{v_{i-1}(1)}(\xi_{2}), \tau_{w_{i+1}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{2(J_{2}^{1}(1)+1)} \right] \\ &\geq \mathbf{P}_{\omega}^{w_{i}(1)} \left[\tau_{v_{i}(1)}(\xi_{2}) < \tau_{v_{i-1}(1)}(\xi_{2}), \tau_{v_{i}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{4(J_{2}^{1}(1)+1)} \right] \\ &\times \mathbf{P}_{\omega}^{v_{i}(1)} \left[\tau_{w_{i+1}(1)}(\xi_{2}) < \tau_{w_{i}(1)}(\xi_{2}), \tau_{w_{i+1}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{4(J_{2}^{1}(1)+1)} \right] \\ &:= R_{4}^{i} \times R_{5}^{i}. \end{split}$$

Now, to bound R_4^i , we are going to distinguish two cases: $W(v_{i-1}(1)) < W(v_i(1))$ and $W(v_{i-1}(1)) > W(v_i(1))$. In the first case, we can directly use a similar computation as that we used for R_2 . In the second case note that $W(v_{i-1}(1)) - W(v_i(1)) < \varepsilon \ln t$ since $(a_1 - \varepsilon) \ln t \le W(v_i(1)) - W(w_i(1)) < a_1 \ln t$. With this observation, we can use a similar computation as that we used for R_2 . So, suming up, we obtain in both cases

$$R_4^i \ge \frac{e^{-(W(v_i(1)) - W(v_{i-1}(1)))} t^{-\varepsilon} \ln^{-2K_0} t}{v_i(1) - v_{i-1}(1)} - o(\exp(-t^{\frac{\varepsilon}{2}})$$
(3.39)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

Then, for R_5^i using (3.1) and (3.3), we obtain for $\omega \in \Gamma_t \cap \Lambda_t$

$$R_5^i \ge \frac{\ln^{-2K_0} t}{w_{i+1}(1) - w_i(1)} - o(\exp(-t^{\frac{\epsilon}{3}})).$$
 (3.40)

as $t \to \infty$.

Finally from (3.34) and (3.37)-(3.40) we obtain for $\omega \in \Gamma_t \cap \Lambda_t$ and sufficiently large t

$$R_1 \ge \left[\frac{t^{-\varepsilon} \ln^{-2K_0} t}{(2h_2(1))^2} \right]^{J_2^1(1)+1} t^{-(r_2(1)-a_1)}. \tag{3.41}$$

Now let us go back to equation (3.33).

Lower bound for U_1

First observe that in this case, $m'_2(2) = m_3(1)$. Now, we divide the interval $[h_2(1), m_3(1)]$ into $t^{a_1-\varepsilon}$ stable wells. We denote by $J_2^2(1) = J_2^2(1)(\omega, t)$ the number of $t^{a_1-\varepsilon}$ -stable wells in the interval $[h_2(1), m_3(1)]$.

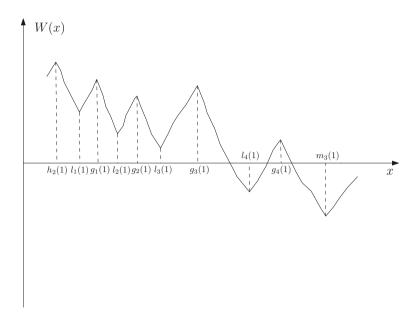


FIGURE 4. On the definition of $l_i(1)$ and $g_i(1)$. Case $J_2^2(1) = 4$.

If $J_2^2(1) \ge 1$, then we define for $1 \le i \le J_2^2(1)$ the point $l_i(1)$ as the minimum of the ith $t^{a_1-\varepsilon}$ -stable well and $g_i(1) := \arg\max_{l_i(1) < x < l_{i+1}(1)} W(x)$. We also define $g_{J_2^2(1)}^1 = \arg\max_{l_{J_2^2(1)}(1) < x < m_3(1)} W(x)$. See Figure 4. If $J_2^2(1) = 0$, using (3.1) and (3.3), we can directly compute a lower bound for U_1 for $\omega \in \Gamma_t \cap \Lambda_t$. As $t \to \infty$, we obtain

$$U_1 \ge \frac{\ln^{-2K_0} t}{m_3(1) - w_{J_0^1}(1)} - o(\exp(-t^{\frac{\epsilon}{3}}). \tag{3.42}$$

If $J_2^2(1) \geq 1$ we use the following decomposition

$$U_{1} \geq \mathsf{P}_{\omega}^{h_{2}(1)} \left[\tau_{l_{1}(1)}(\xi_{2}) < \tau_{w_{J_{2}(1)}(1)}(\xi_{2}), \tau_{l_{1}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{2(J_{2}^{2}(1)+1)} \right]$$

$$\times \prod_{i=1}^{J_{2}^{2}(1)} \mathsf{P}_{\omega}^{l_{i}(1)} \left[\tau_{l_{i+1}(1)}(\xi_{2}) < \tau_{g_{i-1}(1)}(\xi_{2}), \tau_{l_{i+1}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{2(J_{2}^{2}(1)+1)} \right]. \tag{3.43}$$

Let us define

$$U_2^i := \mathbf{P}_{\omega}^{l_i(1)} \left[\tau_{l_{i+1}(1)}(\xi_2) < \tau_{g_{i-1}(1)}(\xi_2), \tau_{l_{i+1}(1)}(\xi_2) \le \frac{t^{a_1}}{2(J_2^2(1)+1)} \right]$$

for $0 \le i \le J_2^2(1)$. Then, let us note that by the Markov property,

$$U_{2}^{i} \geq P_{\omega}^{l_{i}(1)} \left[\tau_{g_{i}(1)}(\xi_{2}) < \tau_{g_{i-1}(1)}(\xi_{2}), \tau_{g_{i}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{4(J_{2}^{2}(1)+1)} \right]$$

$$\times P_{\omega}^{g_{i}(1)} \left[\tau_{l_{i+1}(1)}(\xi_{2}) < \tau_{l_{i}(1)}(\xi_{2}), \tau_{l_{i+1}(1)}(\xi_{2}) \leq \frac{t^{a_{1}}}{4(J_{2}^{2}(1)+1)} \right]$$

$$:= U_{3}^{i} \times U_{4}^{i}.$$

$$(3.44)$$

Now, to bound the term U_3^i we have to distinguish two cases.

Case 1: $W(g_i(1)) < W(g_{i-1}(1))$

In this case, by a similar computation as that we used for R_4^i in the case $W(v_{i-1}(1) < W(v_i(1))$, we obtain for $\omega \in \Gamma_t \cap \Lambda_t$,

$$U_3^i \ge \frac{\ln^{-2K_0} t}{g_i(1) - g_{i-1}(1)} - o(\exp\left(-t^{\frac{\epsilon}{2}}\right)$$
(3.45)

as $t \to \infty$.

Case 2: $W(g_i(1)) > W(g_{i-1}(1))$

By a similar computation as that we used for R_4^i in the case $W(v_{i-1}(1)) > W(v_i(1))$, we obtain for $\omega \in \Gamma_t \cap \Lambda_t$,

$$U_3^i \ge \frac{t^{-\varepsilon} \ln^{-2K_0} t}{g_i(1) - g_{i-1}(1)} - o(\exp(-t^{\frac{\varepsilon}{2}}))$$
(3.46)

 $t \to \infty$.

Then, for U_4^i , using (3.1) and (3.3), we obtain for $\omega \in \Gamma_t \cap \Lambda_t$,

$$U_4^i \ge \frac{\ln^{-2K_0} t}{l_{i+1}(1) - l_i(1)} - o(\exp(-t^{\frac{\epsilon}{3}}))$$
(3.47)

as $t \to \infty$. Finally as the first term of the right-hand side of (3.43) is completely similar to the term U_3^i , from (3.42), (3.45), (3.46) and (3.47) we obtain for $\omega \in \Gamma_t \cap \Lambda_t$ and sufficiently large t,

$$U_1 \ge \left[\frac{t^{-\varepsilon} \ln^{-2K_0} t}{(2m_3(1))^2} \right]^{J_2^2(1)+1}.$$
(3.48)

Case 2: $m'_2(1)$ is not t^{a_1} -stable

In this case we have to distinguish two subcases.

• $a_1 = l_1(1)$

To find a lower bound for $P_{\omega}^{m_2'(1)}[\tau_{m_2'(2)}(\xi_2) < \tau_{h_1(1)}(\xi_2), \tau_{m_2'(2)}(\xi_2) \le t^{a_1}]$ in this case, we can use a similar computation as for the term U_1 above. We obtain that for $\omega \in \Gamma_t \cap \Lambda_t$ and sufficiently large t,

$$P_{\omega}^{m_{2}'(1)}[\tau_{m_{2}'(2)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{m_{2}'(2)}(\xi_{2}) \le t^{a_{1}}] \ge \left[\frac{t^{-\varepsilon} \ln^{-2K_{0}} t}{(2m_{2}(1))^{2}}\right]^{J_{1}^{2}(1)}.$$
(3.49)

• $a_1 = r_1(1)$

In this case, we use the decomposition (3.33). The unique difference is that, using the Markov property we decompose the term R_1 as follows

$$\mathbb{P}_{\omega}^{m_{2}'(1)} \left[\tau_{h_{2}(1)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{h_{2}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{2} \right] \ge \mathbb{P}_{\omega}^{m_{2}'(1)} \left[\tau_{m_{2}(1)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{m_{2}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{4} \right] \\
\times \mathbb{P}_{\omega}^{m_{2}(1)} \left[\tau_{h_{2}(1)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{h_{2}(1)}(\xi_{2}) \le \frac{t^{a_{1}}}{4} \right] . \tag{3.50}$$

To find a lower bound for the first term of the right-hand side of (3.50) we can use a similar computation as for the term U_1 . For the second term of the right-hand side of (3.50) we can use a similar computation as for R_1 . We obtain for $\omega \in \Gamma_t \cap \Lambda_t$ and sufficiently large t

$$P_{\omega}^{m_{2}'(1)}[\tau_{m_{2}'(2)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{m_{2}'(2)}(\xi_{2}) \le t^{a_{1}}]$$

$$\ge \left[\frac{t^{-\varepsilon} \ln^{-4K_{0}} t}{(2m_{3}(1))^{2}}\right]^{J_{2}^{1}(1) + J_{1}^{2}(1) + J_{2}^{2}(1) + 2} t^{-(r_{2}(1) - a_{1})}. \tag{3.51}$$

Finally, using (3.32), (3.41), (3.48), (3.49) and (3.51) we obtain

$$P_{\omega}[D_1 \mid D_0] \ge \frac{1}{2} \left[\frac{t^{-\varepsilon} \ln^{-4K_0} t}{(2m_2'(2))^2} \right]^{J_2^1(1) + J_1^2(1) + J_2^2(1) + 2} t^{-(r_2(1) - a_1)^+} \\
= t^{-((r_2(1) - a_1)^+ + o(1))} \tag{3.52}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.3.3. Lower bounds for $P_{\omega}[D_i \mid D_0 \cap \ldots \cap D_{i-1}]$, where $2 \leq i < N$

Using the Markov property, we first obtain

$$P_{\omega}[D_i \mid D_0 \cap \ldots \cap D_{i-1}] = \sum_{x=0}^{h_1(i-1)} P_{\omega}[D_i \mid \xi_1(\beta_i) = x, \xi_2(\beta_i) = m_2'(i)] P_{\omega}[\xi_1(\beta_i) = x \mid D_0 \cap \ldots \cap D_{i-1}].$$

Let us bound $P_{\omega}[D_i \mid \xi_1(\beta_i) = x, \xi_2(\beta_i) = m'_2(i)]$ uniformly in x for $x < h_1(i-1)$. We have by the Markov property and the independence of the two random walks

$$P_{\omega}[D_{i} \mid \xi_{1}(\beta_{i}) = x, \xi_{2}(\beta_{i}) = m'_{2}(i)]$$

$$= P_{\omega}^{x}[\tau_{h_{1}(i)}(\xi_{1}) > t^{a_{i}}]P_{\omega}^{m'_{2}(i)}[\tau_{m'_{2}(i+1)}(\xi_{2}) < \tau_{h_{1}(i)}(\xi_{2}), \tau_{m'_{2}(i+1)}(\xi_{2}) \le t^{a_{i}}]$$

$$:= V_{1}^{i} \times V_{2}^{i}. \tag{3.53}$$

Let us first treat the first term of the right-hand side of (3.53).

Lower bounds for V_1^i

If $h_1(i) > h_1(i-1)$, then by the Markov property and as $x < h_1(i-1)$ we obtain,

$$V_1^i \ge \mathbf{P}_{\omega}^{m_1(i)}[\tau_{h_1(i)}(\xi_1) > t^{a_i}].$$

As we did in (3.31) we show that for $\omega \in \Gamma_t \cap \Lambda_t$ and t large enough we have

$$P_{\omega}^{m_1(i)}\left[\tau_{h_1(i)}(\xi_1) > t^{a_i}\right] \ge \frac{1}{2}$$

If $h_1(i) = h_1(i-1)$, we use the following decomposition

$$V_1^i \ge \mathsf{P}_{\omega}^x \left[\tau_{m_1(i)}(\xi_1) < \tau_{h_1(i)}(\xi_1), \tau_{m_1(i)}(\xi_1) \le t^{a_i}\right] \mathsf{P}_{\omega}^{m_1(i)} \left[\tau_{h_1(i)}(\xi_1) > t^{a_i}\right]. \tag{3.54}$$

Again we have

$$P_{\omega}^{m_1(i)} \left[\tau_{h_1(i)}(\xi_1) > t^{a_i} \right] \ge \frac{1}{2}. \tag{3.55}$$

For the first term of the right-hand side of (3.54), we can perform exactly the same computation as in Section 3.3.2 for the term U_1 , to find that, being $J_1^1(i)$ the number of $t^{a_i-\varepsilon}$ -stable wells in the interval $[m_1(i), h_1(i)]$,

$$P_{\omega}^{x}[\tau_{m_{1}(i)}(\xi_{1}) < \tau_{h_{1}(i)}(\xi_{1}), \tau_{m_{1}(i)}(\xi_{1}) \le t^{a_{i}}] \ge \left[\frac{t^{-\varepsilon} \ln^{-2K_{0}} t}{(2h_{1}(i))^{2}}\right]^{J_{1}^{1}(i)+1}$$
(3.56)

for $\omega \in \Gamma_t \cap \Lambda_t$ and sufficiently large t.

Lower bounds for V_2^i

This term is completely similar to the term

$$\mathbf{P}_{\omega}^{m_{2}'(1)}[\tau_{m_{2}'(2)}(\xi_{2}) < \tau_{h_{1}(1)}(\xi_{2}), \tau_{m_{2}'(2)}(\xi_{2}) \leq t^{a_{1}}]$$

of Section 3.3.2. We follow step by step the method we applied in that section to obtain that

$$V_2^i \ge \left\lceil \frac{t^{-\varepsilon} \ln^{-4K_0} t}{(2m_2'(i+1))^2} \right\rceil^{J_2^1(i) + J_1^2(i) + J_2^2(i) + 2} t^{-(r_2(i) - a_i)^+}. \tag{3.57}$$

Finally using (3.55)-(3.57) we obtain

$$P_{\omega}[D_{i} \mid D_{0} \cap \dots \cap D_{i-1}] \ge \frac{1}{2} \left[\frac{t^{-\varepsilon} \ln^{-4K_{0}} t}{(2m'_{2}(i+1))^{2}} \right]^{J_{1}^{1}(i)+J_{2}^{1}(i)+J_{2}^{2}(i)+J_{2}^{2}(i)+3} t^{-(r_{2}(i)-a_{i})^{+}} \\
= t^{-((r_{2}(i)-a_{i})^{+}+o(1))} \tag{3.58}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.3.4. Lower bound for $P_{\omega}[D_N \mid D_0 \cap \ldots \cap D_{N-1}]$

As we have already done in Section 3.3.3, we have by the Markov property

$$P_{\omega}[D_N \mid D_1 \cap \ldots \cap D_{N-1}] = \sum_{x=0}^{h_1(N-1)} P_{\omega}[D_N \mid \xi_1(\beta_N) = x, \xi_2(\beta_N) = m_2'(N)]$$

$$\times P_{\omega}[\xi_1(\beta_N) = x, \xi_2(\beta_N) = m_2'(N) \mid D_0 \cap \ldots \cap D_{N-1}].$$

Then, we have by the Markov property and the independence of the two random walks

$$P_{\omega}[D_N \mid \xi_1(\beta_N) = x, \xi_2(\beta_N) = m_2'(N)] = P_{\omega}^x[\tau_{h_1(N)}(\xi_1) > t] P_{\omega}^{m_2'(N)}[\tau_{h_1(N)}(\xi_2) > t]$$

with $x < h_1(N-1)$. As we did in Section 3.3.3 for V_1^i , one can see that

$$\mathbf{P}_{\omega}^{x}[\tau_{h_{1}(N)}(\xi_{1}) > t] \geq \frac{1}{2} \left[\frac{t^{-\varepsilon} \ln^{-2K_{0}} t}{(2h_{1}(N))^{2}} \right]^{J_{1}^{1}(N)}$$

for t large enough and $\omega \in \Gamma_t \cap \Lambda_t$. Using (3.4), we show that for t large enough and $\omega \in \Gamma_t \cap \Lambda_t$,

$$P_{\omega}^{m_2'(N)}[\tau_{h_1(N)}(\xi_2) > t] \ge \frac{1}{2}$$

Therefore, we obtain

$$P_{\omega}[D_N \mid D_0 \cap \dots \cap D_{N-1}] \ge \frac{1}{4} \left[\frac{t^{-\varepsilon} \ln^{-2K_0} t}{(2m_2'(N))^2} \right]^{J_1^1(N)} = t^{o(1)}$$
(3.59)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

3.3.5. Conclusion

Using the results of the precedent subsections, it is possible to show, using the same kind of induction argument as in 3.2.4, that

$$\prod_{i=1}^{N} \mathsf{P}_{\omega}[D_i \mid D_1 \cap \ldots \cap D_{i-1}] \ge t^{-(r_1(N) + o(1))}$$
(3.60)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

By Lemma 2.4, (3.29) and (3.60), we obtain that \mathbb{P} -a.s.

$$P_{\omega}[T_2 > t] \ge t^{-(r_1(N) + o(1))} = t^{-(\zeta_2(t) + o(1))}$$
(3.61)

as $t \to \infty$. Together with (3.27), the inequality (3.61) proves Theorem 1.2.

To conclude, we deduce the second part of Theorem 1.4. Note first that by Theorem 1.3, we have for $\theta > 0$ and every $t \ge e$

$$\mathbb{P}[\zeta_2(t) < 1 + \theta] = \mathbb{P}[\zeta_2(e) < 1 + \theta] > 0 \tag{3.62}$$

the last inequality is obtained using the fact that, as W is a Brownian motion, $\zeta_2(e)$ is absolutely continuous in relation to the Lebesgue measure with density non almost surely zero in $[1, 1 + \theta]$. From (3.62) and as the sequence $(\zeta_2(n))_{n\geq 3}$ is ergodic, there exists an increasing subsequence $(n_i)_i$ such that $1 \leq \zeta_2(n_i) < 1 + \theta$, \mathbb{P} -a.s. for all $i \geq 1$. Now by Markov's inequality, we get

$$\mathbb{E}_{\omega}[T_2^c] \ge \mathbb{P}_{\omega}[T_2 > n_i]n_i^c \tag{3.63}$$

for every n_i . Thus from (3.61), (3.63) and the fact that $1 \le \zeta_2(n_i) < 1 + \theta$ for the subsequence $(n_i)_i$ we get that

$$\mathbb{E}_{\omega}[T_2^c] \ge n_i^{(c-(1+\theta)+o(1))}$$

for i sufficiently large. If c > 1, taking $\theta < c - 1$ and letting $i \to \infty$ we obtain $\mathbb{E}_{\omega}[T_2^c] = +\infty$, \mathbb{P} -a.s.

4. Proofs of Theorems 1.2 and 1.4 in the general case

4.1. Upper bound for $P_{\omega}[T_{\gamma} > t]$

First, as we did in the case $\gamma = 2$, let us find an upper bound for $P_{\omega}[T_{\gamma} > t^{1+\delta}]$ for $\delta > 0$ arbitrary instead of $P_{\omega}[T_{\gamma} > t]$. Then we write

$$\mathtt{P}_{\omega}[T_{\gamma} > t^{1+\delta}] = \mathtt{P}_{\omega}[T_{\{\gamma-1,\gamma\}} > t^{1+\delta} \mid T_{\gamma-1} > t^{1+\delta}] \mathtt{P}_{\omega}[T_{\gamma-1} > t^{1+\delta}]$$

where $T_{\{\gamma-1,\gamma\}} = \inf\{s > 0; \xi_{\gamma-1}(s) = \xi_{\gamma}(s)\}$. Let us suppose that for some $\gamma > 2$,

$$P_{\omega}[T_{\gamma-1} > t^{1+\delta}] \le t^{-(\sum_{i=1}^{\gamma-2} (\gamma - i)r_i(N) + o(1))}$$
(4.1)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. The goal of this section is to show that this implies

$$P_{\omega}[T_{\gamma} > t^{1+\delta}] \le t^{-(\sum_{i=1}^{\gamma-1} (\gamma - i)r_i(N) + o(1))}$$
(4.2)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. As by Section 3.2, (4.1) is true for $\gamma = 3$, we will conclude by induction that (4.2) is true for all $\gamma \ge 2$.

Now, we have to bound from above the term $P_{\omega}[T_{\{\gamma-1,\gamma\}} > t^{1+\delta} \mid T_{\gamma-1} > t^{1+\delta}]$ for $\gamma \geq 3$ to obtain an upper bound for $P_{\omega}[T_{\gamma} > t^{1+\delta}]$. For the sake of brevity, let us write $P_{\omega}^*[\cdot] := P_{\omega}[\cdot \mid T_{\gamma-1} > t^{1+\delta}]$.

Fix t > 1 and define, for $1 \le i \le N$, the following stopping times

$$\sigma_i = \inf\{s > 0; \xi_{\gamma}(s) = h_{\gamma - 1}(i)\},$$
(4.3)

the following events

$$B_i = \{ T_{\{\gamma - 1, \gamma\}} \notin [\sigma_{i-1}, \sigma_i] \} \tag{4.4}$$

for $1 \le i \le N-1$, with the convention $\sigma_0 = 0$ and

$$B_N = \{ T_{\{\gamma-1,\gamma\}} \notin [\sigma_{N-1}, t^{1+\delta}] \} \cap \left\{ \sigma_{N-1} < \frac{t^{1+\delta}}{2} \right\} \cdot$$

To find the upper bound for $P_{\omega}^*[T_{\{\gamma-1,\gamma\}} > t^{1+\delta}]$, we will use the following decomposition

$$\mathbf{P}_{\omega}^{*}[T_{\{\gamma-1,\gamma\}} > t^{1+\delta}] = \mathbf{P}_{\omega}^{*}\left[T_{\{\gamma-1,\gamma\}} > t^{1+\delta}, \sigma_{N-1} < \frac{t^{1+\delta}}{2}\right] + \mathbf{P}_{\omega}^{*}\left[T_{\{\gamma-1,\gamma\}} > t^{1+\delta}, \sigma_{N-1} \ge \frac{t^{1+\delta}}{2}\right]$$
(4.5)

Using the same kind of arguments as in Section 3.2, it is possible to show that the second term of the right-hand side of (4.5) is negligible in comparison with the first one as t tends to infinity (see [9] Sect. 4.1 for more details). For the upper bound of the first term of (4.5) we start by noting that

$$\mathbb{P}_{\omega}^{*} \left[T_{\{\gamma-1,\gamma\}} > t^{1+\delta}, \sigma_{N-1} < \frac{t^{1+\delta}}{2} \right] \leq \mathbb{P}_{\omega}^{*} [B_{1} \cap \ldots \cap B_{N}] \\
= \mathbb{P}_{\omega}^{*} [B_{1}] \mathbb{P}_{\omega}^{*} [B_{2} \mid B_{1}] \dots \mathbb{P}_{\omega}^{*} [B_{N} \mid B_{1} \cap B_{2} \cap \ldots \cap B_{N-1}]. \tag{4.6}$$

We now give upper bounds for the terms of the product of the right-hand side of (4.6) (see [9] Sects. 4.1.1 to 4.1.3 for more details). We have

$$P_{\omega}^*[B_1] \le t^{-(\sum_{j=1}^{\gamma-1} r_j(1) + o(1))} \tag{4.7}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

If $h_{\gamma-1}(N) = h_{\gamma-1}(N-1)$, we have $P_{\omega}^*[B_i \mid B_1 \cap \cdots \cap B_{i-1}] = 1$ for $1 \leq i \leq N$. Otherwise if $h_{\gamma-1}(N) > h_{\gamma-1}(N-1)$, we have for $1 \leq i \leq N$,

$$P_{\omega}^{*}[B_{i} \mid B_{1} \cap \dots \cap B_{i-1}] \leq 2C_{4}t^{\varepsilon}t^{-(r_{\gamma}(i-1)-a_{i-1})} = t^{-(r_{\gamma}(i-1)-a_{i-1}+o(1))}$$

$$(4.8)$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

By (4.7), (4.8) and the construction of Section 2 we can deduce by the same type of induction as that we used in Section 3.2.4 that for $\gamma \geq 3$,

$$\mathbf{P}_{\omega}^*[T_{\{\gamma-1,\gamma\}} > t] \leq t^{-\left(\frac{1}{1+\delta}\sum_{j=1}^{\gamma-1}r_j(N) + o(1)\right)}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. This leads to

$$\mathrm{P}_{\omega}[T_{\gamma} > t] \leq t^{-\left(\frac{\zeta_{\gamma}(t)}{1+\delta} + o(1)\right)}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. We can immediately deduce the first part of Theorem 1.4. Indeed, as $r_j(N) \ge 1$, for $1 \le j \le \gamma - 1$, we obtain \mathbb{P} -a.s.,

$$\mathrm{P}_{\omega}[T_{\gamma} > t] \leq t^{-\left(\frac{\gamma(\gamma-1)}{2(1+\delta)} + o(1)\right)}$$

as $t \to \infty$. To conclude, fix $c < \frac{\gamma(\gamma-1)}{2}$, as δ is arbitrary we can choose δ sufficiently small such that $\delta < \frac{\gamma(\gamma-1)-2c}{2c}$. This shows that $\mathbf{E}_{\omega}[T_{\gamma}^c] < \infty$, \mathbb{P} -a.s.

Now, in order to prove Theorem 1.2, we shall improve (3.26). That is why we shall consider convergence in \mathbb{P} -probability instead of \mathbb{P} -a.s. convergence to obtain

$$\ln P_{\omega}[T_{\gamma} > t] \le -\zeta_{\gamma}(t) \ln t + o(\ln t) \tag{4.9}$$

as $t \to \infty$. By a similar argument as that we used in Section 3.2.4 for the case $\gamma = 2$ we can obtain (4.9) in \mathbb{P} -probability.

4.2. Lower bound for $P_{\omega}[T_{\gamma} > t]$

Let $\tau_i^2(\xi_j)$ the time of second return to state i for the jth random walk.

We define the following events

$$D_0^1 = \bigcap_{i=1}^{\gamma-1} \{ \tau_{\{i-1,i+1\}}(\xi_i) > \tau_{m_{\alpha}'(1)}(\xi_{\gamma}) \} \cap \{ \xi_{\gamma}(s) \le \xi_{\gamma}(t), 0 \le s < t \le \tau_{m_{\alpha}'(1)}(\xi_{\gamma}) \}$$

and for $2 \le j \le \gamma - 1$,

$$\begin{split} D_0^j &= \cap_{i=1}^{\gamma-j} \big\{ \tau_{\{i-1,i+1\}}(\xi_i) > \tau_{m_{\gamma-j+1}(1)}(\xi_{\gamma-j+1}) \big\} \\ &\quad \cap_{i=\gamma-j+2}^{\gamma-1} \big\{ \tau_{m_i(1)-1}^2(\xi_i) \wedge \tau_{m_i(1)+1}(\xi_i) > \tau_{m_{\gamma-j+1}(1)}(\xi_{\gamma-j+1}) \big\} \\ &\quad \cap \big\{ \tau_{m_{\gamma}'(1)-1}^2(\xi_{\gamma}) \wedge \tau_{m_{\gamma}'(1)+1}(\xi_{\gamma}) > \tau_{m_{\gamma-j+1}(1)}(\xi_{N-j+1}) \big\} \\ &\quad \cap \big\{ \xi_{\gamma-j+1}(s) \le \xi_{\gamma-j+1}(t), 0 \le s < t \le \tau_{m_{\gamma-j+1}(1)}(\xi_{\gamma-j+1}) \big\} \end{split}$$

with the convention $\bigcap_{i=\gamma}^{\gamma-1} \{\ldots\} = \Omega$. We define $D_0 = \bigcap_{j=1}^{\gamma-1} D_0^j$. For $1 \leq n \leq N$, define

$$\beta(n) = \max_{2 \le j \le \gamma} \{ \tau_{m_j(n)}(\xi_j) \}.$$

Next, for $1 \le n < N$, if $a_n = l_{\gamma-1}(n)$ then define

$$D_{n} = \{ \tau_{h_{\gamma-1}(n)}^{2}(\xi_{\gamma}) > \tau_{m'_{\gamma}(n+1)}(\xi_{\gamma}), \tau_{m'_{\gamma}(n+1)}(\xi_{\gamma}) - \beta(n) \leq t^{a_{n}} \}$$
$$\cap_{j=1}^{\gamma-1} \{ (\tau_{h_{j-1}(n)}^{2}(\xi_{j}) \wedge \tau_{h_{j}(n)}(\xi_{j})) - \beta(n) > t^{a_{n}} \}$$

with the convention $\tau_{h_0(n)}^2(\xi_1) = \tau_{h_1(n)}(\xi_1)$.

If $a_n = r_i(n)$ for some $1 \le i \le \gamma - 1$ or $l_i(n)$ for some $1 \le i \le \gamma - 2$ then define

$$D_{n} = \bigcap_{j=i}^{\gamma} \{ \tau_{h_{j-1}(n)}^{2}(\xi_{j}) > \tau_{m_{j+1}(n)}(\xi_{j}), \theta_{j} t^{a_{n}} < \tau_{m_{j+1}(n)}(\xi_{j}) - \beta(n) \leq \theta_{j}' t^{a_{n}} \}$$

$$\cap_{j=1}^{i-1} \{ (\tau_{h_{j-1}(n)}^{2}(\xi_{j}) \wedge \tau_{h_{j}(n)}(\xi_{j})) - \beta(n) > t^{a_{n}} \}$$

$$\cap_{j=i}^{\gamma} \{ \tau_{h_{j+1}(n)}(\xi_{j}) - \beta(n) > t^{a_{n}} \}$$

with the conventions $\tau_{h_0(n)}^2(\xi_1) = \tau_{h_1(n)}(\xi_1), \, \cap_{j=1}^0 \{\ldots\} = \Omega$,

$$h_{\gamma+1}(n) = \min_{x > m_{\gamma+1}(n)} \{ W(x) - W(m_{\gamma+1}(n)) = a_n \ln t \}$$

and

$$\theta_j = \frac{\gamma - j}{\gamma - i + 1}, \quad \theta'_j = \frac{\gamma - j + 1}{\gamma - i + 1}.$$

Finally define

$$D_N = \{ \tau_{h_{\gamma-1}(N)}^2(\xi_\gamma) - \beta(N) \} \cap_{j=1}^{\gamma-1} \{ (\tau_{h_j(N)}^2(\xi_j) \wedge \tau_{h_{j+1}(N)}(\xi_j)) - \beta(N) > t \} \cap \{ \tau_{h_1(N)}(\xi_1) - \beta(N) > t \}.$$

On the event D_0 we place all the γ random walks at the points $m_j(1)$, for $1 \leq j \leq \gamma - 1$ and $m'_{\gamma}(1)$. Then, on the event D_n for $1 \leq n \leq N-1$, suppose that the smallest well is $I(n) = [h_{i-1}(n), h_i(n)]$, that is to say, $r_i(n) \wedge l_{i-1}(n) = a_n$. By (3.2), we know that the random walk in the well I(n) could escape from it in a time of order $t^{a_n+\varepsilon}$ with high probability. In order for the random walk which is in this well not to meet the random walks which are in the neighboring wells, we will oblige this random walk and those that are in the following wells to move to the right to occupy the first $\gamma - i + 1$ t^{a_n} -stable wells to the right of the well I(n). Furthermore, we want this to occur in a time of order less than t^{a_n} , avoiding the random walks meeting by controlling the intervals of time each random walk moves to its well to the next well. The random walks to the left of the well I(n) stay in their respective wells. Finally on D_N we oblige all the random walks to stay in their respective t-stable wells until time t.

Observe that

$$P_{\omega}[T_{\gamma} > t] \ge P_{\omega}[D_0]P_{\omega}[D_1 \mid D_0] \dots P_{\omega}[D_N \mid D_0 \cap D_1 \cap \dots \cap D_{N-1}]. \tag{4.10}$$

We now give a lower bound for each term of the right-hand side of (4.10) (see [9] Sects. 4.2.1 to 4.2.3 for more details).

For $P_{\omega}[D_0]$, we obtain

$$P_{\omega}[D_0] \ge \left(\frac{e^{-(\gamma - 1)\kappa^2}(e - 1)}{2e\kappa^2}\right)^{\gamma \ln^{1/2} t} \tag{4.11}$$

for $\omega \in \Gamma_t \cap \Lambda_t$. For $P_{\omega}[D_n \mid D_0 \cap D_1 \cap \ldots \cap D_{n-1}]$, with $1 \leq n \leq N-1$, if $a_n = l_{\gamma-1}(n)$, we have

$$P_{\omega}[D_n \mid D_0 \cap \dots \cap D_{n-1}] \ge t^{-((r_{\gamma}(n) - a_n)^+ + o(1))}$$
(4.12)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. If $a_n = r_i(n)$ for some $1 \le i \le \gamma - 1$ or $l_i(n)$ for some $1 \le i \le \gamma - 2$, we obtain

$$P_{\omega}[D_n \mid D_0 \cap \dots \cap D_{n-1}] \ge t^{-(\sum_{j=i}^{\gamma} r_{j+1}(n) - (\gamma - i + 1)a_n + o(1))}$$
(4.13)

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. Finally, it is possible to show that

$$P_{\omega}[D_N \mid D_0 \cap \cdots \cap D_{N-1}] \ge t^{o(1)} \tag{4.14}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

We now show by induction that the total cost of our strategy immediately after step n-1 is bounded from below by $t^{-(\sum_{j=1}^{\gamma-1}(\gamma-j)r_j(n)+o(1))}$. That is to say, if we denote by

$$\Pi_{n-1} = P_{\omega}[D_0]P_{\omega}[D_1 \mid D_0] \dots P_{\omega}[D_{n-1} \mid D_0 \cap D_1 \cap \dots \cap D_{n-2}]$$

for $1 \le n \le N$, we have

$$\Pi_{n-1} \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(n) + o(1))}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$.

So suppose that

$$\Pi_{n-1} \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(n) + o(1))}$$

is true for some $1 \le n < N$. We will show that it implies that

$$\Pi_n \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(n+1) + o(1))}.$$

Case 1: $a_n = l_{\gamma-1}(n)$

In this case we have by (4.12),

$$P_{\omega}[D_n \mid D_0 \cap \cdots \cap D_{n-1}] \ge t^{-((r_{\gamma}(n)-a_n)^++o(1))}$$

In the case $(r_{\gamma}(n) - a_n)^+ > 0$, let us compute the total cost using the construction of Section 2.

$$\sum_{j=1}^{\gamma-1} (\gamma - j) r_j(n) + r_{\gamma}(n) - a_n = \sum_{j=1}^{\gamma-2} (\gamma - j) r_j(n) + r_{\gamma-1}(n) + r_{\gamma}(n) - a_n$$

$$= \sum_{j=1}^{\gamma-2} (\gamma - j) r_j(n+1) + r_{\gamma-1}(n+1)$$

$$= \sum_{j=1}^{\gamma-1} (\gamma - j) r_j(n+1)$$

where we use the fact that, in this case, $r_{\gamma-1}(n+1) = r_{\gamma-1}(n) + r_{\gamma}(n) - a_n$. In the case $(r_{\gamma}(n) - a_n)^+ = 0$, we have directly

$$\sum_{j=1}^{\gamma-1} (\gamma - j) r_j(n) = \sum_{j=1}^{\gamma-1} (\gamma - j) r_j(n+1).$$

since $r_i(n) = r_i(n+1)$ for $1 \le j \le \gamma - 1$.

Therefore, in both cases $(r_{\gamma}(n) - a_n)^+ > 0$ and $(r_{\gamma}(n) - a_n)^+ = 0$ we obtain

$$\Pi_n > t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(n+1) + o(1))}.$$

Case 2: $a_n = r_i(n)$ for some $1 \le i \le \gamma - 1$ or $l_i(n)$ for some $1 \le i \le \gamma - 2$

By (4.13) we have

$$P_{\omega}[D_n \mid D_0 \cap \dots \cap D_{n-1}] \ge t^{-(\sum_{j=i}^{\gamma} r_{j+1}(n) - (\gamma - i + 1)a_n + o(1))}. \tag{4.15}$$

As an example, let us treat the case $a_n = r_i(n)$. Let us compute the total cost using the construction of Section 2. We have

$$\sum_{j=1}^{\gamma-1} (\gamma - j) r_j(n) + \sum_{j=i}^{\gamma} r_j(n) - (\gamma - i + 1) a_n = r_{\gamma}(n) + \sum_{j=i+1}^{\gamma-1} (\gamma - j + 1) r_j(n) + (\gamma - i + 1) r_i(n)$$

$$+ \sum_{j=1}^{i-2} (\gamma - j) r_j(n) + (\gamma - i + 1) r_{i-1}(n) - (\gamma - i + 1) a_n$$

$$= \sum_{j=i}^{\gamma-1} (\gamma - j) r_j(n + 1) + \sum_{j=1}^{i-2} (\gamma - j) r_j(n + 1)$$

$$+ (\gamma - i + 1) r_{i-1}(n + 1)$$

$$= \sum_{j=i}^{\gamma-1} (\gamma - j) r_j(n + 1)$$

where we used the fact that for j < i, $r_j(n+1) = r_j(n)$ and for $j \ge i$, $r_j(n+1) = r_{j+1}(n)$. Hence, we obtain

$$\Pi_n \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(n+1) + o(1))}.$$

By (4.11) and Definition 2.3 item (iii), we have

$$\Pi_0 > t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(1) + o(1))}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. Thus, the induction is right for every $1 \le 1 < N$, in particular we obtain

$$\Pi_{N-1} \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(N) + o(1))}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. Finally by (4.14), we have

$$\Pi_N \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma-j)r_j(N) + o(1))}$$

as $t \to \infty$ and $\omega \in \Gamma_t \cap \Lambda_t$. Which implies

$$P_{\omega}[T_{\gamma} > t] \ge t^{-(\sum_{j=1}^{\gamma-1} (\gamma - j)r_{j+1}(N) + o(1))}$$

or by definition of the process ζ_{γ} ,

$$P_{\omega}[T_{\gamma} > t] \ge t^{-(\zeta_{\gamma}(t) + o(1))}$$

as t tends to infinity and $\omega \in \Gamma_t \cap \Lambda_t$. Together with (4.9), this last inequality concludes the Proof of Theorem 1.2 in the general case. To obtain the second part of Theorem 1.4 we use a similar argument as in Section 3.3.5 for the case $\gamma = 2$.

5. Proof of Theorem 1.3

We will start this last section by treating the case $\gamma = 2$. From this special case, we will deduce the proof of the general case.

Consider the Brownian motions W and W' given by $W'(\cdot) = \lambda W(\cdot/\lambda^2)$ for $\lambda > 0$. Denoting respectively by $\mathcal{S}_t(W)$ and $\mathcal{S}_t(W')$ the sets of the t-stable points of W and W', it is elementary to observe that

$$\mathcal{S}_t(W') = \lambda^2 \mathcal{S}_t(W)$$

and as $\mathcal{H}_t(W) = \mathcal{S}_t(-W)$ we also have

$$\mathcal{H}_t(W') = \lambda^2 \mathcal{H}_t(W).$$

Then using these two properties and expression (1.2) of $\zeta_2(t)$ we obtain

$$\zeta_2(t, W) = \zeta_2(t^{\lambda}, W').$$

As W and W' are identically distributed we deduce $\zeta_2(t) \stackrel{\text{law}}{=} \zeta_2(t^{\lambda})$. With this scaling property, we observe that the distribution of $\zeta_2(t)$, for fixed t, does not depend on t. Let us now compute the distribution of $\zeta_2(e)$. To this end, define the following random variables

$$\begin{split} V_0 &= \inf \left\{ t > 0 : W(t) - \inf_{0 \le s \le t} W(s) = 1 \right\}, \\ M_0 &= \inf \left\{ t \in [0, V_0] : W(t) = \inf_{0 \le s \le V_0} W(s) \right\}, \\ V_1 &= \inf \left\{ t > V_0 : \sup_{V_0 \le s \le t} W(s) - W(t) = 1 \right\}, \\ H_1 &= \inf \left\{ t \in [V_0, V_1] : W(t) = \sup_{V_0 \le s \le V_1} W(s) \right\}. \end{split}$$

See Figure 5.

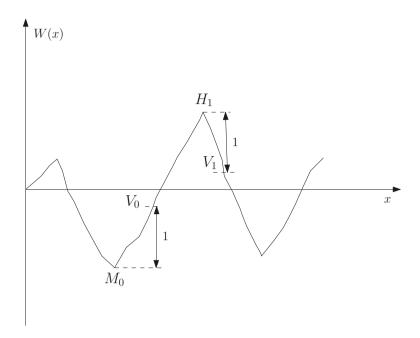


FIGURE 5. On the definition of V_0 , M_0 , V_1 and H_1 .

By definition of $\zeta_2(e)$, one can immediately see that $\zeta_2(e) = W(H_1) - W(M_0)$. Then we have

$$\mathbb{P}[\zeta_2(e) - 1 > u] = \mathbb{P}[W(H_1) - W(M_0) - 1 > u]$$

$$= \mathbb{P}[W(H_1) - W(V_0) + W(V_0) - W(M_0) - 1 > u]$$

$$= \mathbb{P}[W(H_1) - W(V_0) > u].$$

By Lemma 6.1 (ii) of [3] we know that the random variable $W(H_1) - W(V_0)$ is exponentially distributed with mean 1. So, this ends the Proof of Theorem 1.3 in the case $\gamma = 2$.

In the general case, observe that by the same argument we used for the case $\gamma=2$, we show that the distribution of $\zeta_{\gamma}(t)$ does not depend on t. Finally to obtain the distribution of $\zeta_{\gamma}(e)$ observe that by the fact that W has independent increments, $\zeta_{\gamma}(e) - \frac{\gamma(\gamma-1)}{2}$ is the sum of $\gamma-1$ independent exponential random variables with parameters $1/(\gamma-i)$, $1 \le i \le \gamma-1$. Finally after some elementary computations we get expression (1.3).

Final comments

In this last part, we discuss two other natural questions closely related to ours. Suppose that we have $\gamma \geq 2$ random walks starting at positions $x_1 < x_2 < \cdots < x_{\gamma}$.

The first question concerns the coalescing time of the γ random walks. Indeed, suppose that whenever two random walks meet they stay together forever. Thus, we can define the coalescing time T'_{γ} as the moment when all the random walks coalesce. Observe that this time is equal to the time when the first and the γ th random walks meet. Thus, we can deduce that T'_{γ} behaves like T_2 for all $\gamma \geq 2$.

The second question concerns the meeting time T'_{γ} of all the γ random walks at the same time, that is

$$T_{\gamma}'' = \inf\{s > 0; \xi_1(s) = \xi_2(s) = \dots = \xi_{\gamma}(s)\}.$$

Unlike the case of standard random walks, for which T''_{γ} is almost surely finite only for $\gamma \leq 3$, in the case of random walks in random environment, it is possible to prove, analogously to what is shown in the Appendix

of [20], that for all $\gamma \geq 2$, the γ random walks will all meet in the origin \mathbb{P} -a.s., and from this follows the finiteness of T''_{γ} .

In fact, it is possible to show that T''_{γ} also has the same behavior as the meeting time T_2 of two random walks. First, it is possible to show that when the γ random walks are in the same t-stable well they meet at the same time with high probability. Then, in the light of the Proofs of Theorems 1.2 and 1.4, for the γ random walks not to meet at the same time until time t, the strategy which has the "cheapest cost" is to keep the first $\gamma - 1$ random walks in the first t-stable well and to send the γ th random walk into the second t-stable well. This is exactly what we have done in the Proof of Theorem 1.4 for the case of two random walks.

To conclude, Theorems 1.2 and 1.4 for T_2 should also be true for T_2' and T_2'' .

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