

Selected Topics in Random Walks in Random Environment

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Abstract

Random walk in random environment (RWRE) is a fundamental model of statistical mechanics, describing the movement of a particle in a highly disordered and inhomogeneous medium as a random walk with random jump probabilities. It has been introduced in a series of papers as a model of DNA chain replication and crystal growth (see Chernov [Ch67] and Temkin [Te69, Te72]), and also as a model of turbulent behavior in fluids through a Lorentz gas description (Sinaĭ 1982 [Si82a]). It is a simple but powerful model for a variety of complex large-scale disordered phenomena arising from fields such as physics, biology and engineering. While the one-dimensional model is well-understood, in the multidimensional setting, fundamental questions about the RWRE model have resisted repeated and persistent attempts to answer them. Two major complications in this context stem from the loss of the Markov property under the averaged measure as well as the fact that in dimensions larger than one, the RWRE is not reversible anymore. In these notes we present a general overview of the model, with an emphasis on the multidimensional setting and a more detailed description of recent progress around ballisticity questions.

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Chapter 1

Preface

We present a review of random walks in random environment. The main focus evolves around several fundamental open questions concerning the existence of invariant probability measures, transience, recurrence, directional transience and ballisticity. This choice of topics is somewhat biased towards our recent research interests.

The first chapter deals with the question of the existence of an invariant probability measure of the so-called “environmental process”; such a measure is particularly useful if it is absolutely continuous with respect to the law of the environment. The existence and properties of such a measure characterize in some sense the different asymptotic behaviors of the walk, from a general law of large numbers to possibly a quenched central limit theorem, and to a variational formula for the rate function in the case of quenched large deviations. After the introduction of basic definitions and concepts, we review the one-dimensional situation, which turns out to be a controlled laboratory of several phenomena which one would expect to encounter in the multidimensional setting. Subsequently, we investigate the latter setting and give some of the corresponding (limited) results which are available in that context.

It is conjectured that for uniformly elliptic and i.i.d. environments, in dimensions $d \geq 2$, directional transience implies ballisticity. The second chapter of these notes reviews this question as well as the progress and understanding which have been achieved towards its resolution. In particular, we introduce the fundamental concept of renewal times. We then proceed to the ballisticity conditions, under which it has been possible obtain a better understanding of the so-called slowdown phenomena as well as of the ballistic and diffusive behavior in the setting of (uniformly) elliptic environments.

Chapter 2

The environmental process and its invariant measures

2.1 Definitions

Throughout these notes, for $x \in \mathbb{R}^d$, we will use the notations $|x|_\infty$, $|x|_1$ and $|x|_2$ for the L^∞ , L^1 and L^2 norms. For a subset $A \subset \mathbb{Z}^d$ we denote by ∂A its external boundary

$$\{x \in \mathbb{Z}^d \setminus A : \exists y \in A \text{ with } |x - y|_1 = 1\}, \quad (2.1)$$

and for a subset $B \subset \mathbb{R}^d$ we denote by $\overset{\circ}{B}$ its interior. We write

$$B_p(x_0, r) = \{x \in \mathbb{R}^d : |x - x_0|_p \leq r\} \quad (2.2)$$

for the closed ball centered in the x_0 with radius r in the p -norm. In addition, set $B_p(r) := B_p(0, r)$. Furthermore, the set $U := \{e \in \mathbb{Z}^d : |e|_1 = 1\}$ will serve as the set of possible jumps for the random walk to be defined. We will use C to denote constants that can change from one side of an inequality to another, and c_1, c_2, \dots for constants taking fixed values. Furthermore, if we want to emphasize the dependence of a constant on quantities, such as e.g. the dimension, we write $C(d)$. We begin with the definition of an environment.

Definition 2.1. (Environment) We define the set

$$\mathcal{P} := \left\{ (p(e))_{e \in U} \in [0, 1]^U : \sum_{e \in U} p(e) = 1 \right\} \quad (2.3)$$

of $2d$ -vectors p serving as admissible transition probabilities. An *environment* is an element ω of the *environment space* $\Omega := \mathcal{P}^{\mathbb{Z}^d}$ so that $\omega := (\omega(x))_{x \in \mathbb{Z}^d}$, where $\omega(x) \in \mathcal{P}$. We denote the components of $\omega(x)$ by $\omega(x, e)$.

Let us now define a random walk in a given environment ω .

Definition 2.2. (Random walk in an environment ω) Let $\omega \in \Omega$ be an environment and let \mathcal{G} be the σ -algebra on $(\mathbb{Z}^d)^{\mathbb{N}}$ defined by the cylinder functions. For $x \in \mathbb{Z}^d$, we define the

random walk in the environment ω starting in x as the Markov chain $(X_n)_{n \geq 0}$ on \mathbb{Z}^d whose law $P_{x,\omega}$ on $((\mathbb{Z}^d)^{\mathbb{N}}, \mathcal{G})$ is characterized by

$$P_{x,\omega}[X_0 = x] = 1, \quad \text{and}$$

$$P_{x,\omega}[X_{n+1} = y + e \mid X_n = y] = \begin{cases} \omega(y, e), & \text{if } e \in U, \\ 0, & \text{otherwise,} \end{cases}$$

whenever $P_{x,\omega}[X_n = y] > 0$, and 0 otherwise. Furthermore, we denote by

$$p^{(n)}(x, y, \omega) := P_{x,\omega}[X_n = y] \tag{2.4}$$

the n -step transition probability of the random walk in the environment ω .

We will now account for the randomness in the environment. For that purpose, let us endow the environment space Ω with the product topology and let \mathbb{P} be some probability measure defined on $(\Omega, \mathcal{B}(\Omega))$; here, \mathcal{B} denotes the corresponding Borel σ -algebra. We call \mathbb{P} the *law of the environment* and for every measurable function f defined on Ω , we denote by $\mathbb{E}[f]$ the corresponding expectation if it exists. It will frequently be useful to assume that

(IID) the coordinate maps on the product space Ω are independent and identically distributed (i.i.d.) under \mathbb{P} .

In order to give a relaxation of **(IID)** we introduce the following notation. For each $y \in \mathbb{Z}^d$, let us denote by t_y the translation defined on the environment space Ω by

$$(t_y \omega)(x, e) := \omega(x + y, e),$$

for every $x \in \mathbb{Z}^d$ and $e \in U$. It will often be useful to assume that

(ERG) the family of transformations $(t_x)_{x \in \mathbb{Z}^d}$ is an ergodic family acting on $(\Omega, \mathcal{B}(\Omega), \mathbb{P})$.

In other words, if $A \in \mathcal{B}(\Omega)$ is such that $A = t_x^{-1}(A)$ for every $x \in \mathbb{Z}^d$, then $\mathbb{P}(A) = 0$ or 1. This condition is also called *total ergodicity*. In particular, note that **(IID)** implies **(ERG)**.

For a fixed realization of ω , we now call $P_{x,\omega}$ the *quenched law* of the random walk in random environment (RWRE). Using Dynkin's theorem, it is not hard to show that for each $x \in \mathbb{Z}^d$ and $G \in \mathcal{G}$, the mapping

$$\omega \mapsto P_{x,\omega}[G]$$

is $\mathcal{B}(\Omega)$ -measurable. We can therefore define on the space $(\Omega \times (\mathbb{Z}^d)^{\mathbb{N}}, \mathcal{B}(\Omega) \otimes \mathcal{G})$ for each $x \in \mathbb{Z}^d$ the semi-direct product $P_{x,\mathbb{P}}$ of the measures \mathbb{P} and $P_{x,\omega}$ by the formula

$$P_{x,\mathbb{P}}[F \times G] := \int_F P_{x,\omega}(G) \mathbb{P}(d\omega). \tag{2.5}$$

We denote by P_x the marginal law of $P_{x,\mathbb{P}}$ on $(\mathbb{Z}^d)^{\mathbb{N}}$ and call it the *averaged* or *annealed* law of the RWRE. One of the difficulties arising in the study of RWRE is that under the averaged law it is generally not Markovian anymore.

We will need the concepts of ellipticity and uniform ellipticity.

Definition 2.3. (Ellipticity and uniform ellipticity) Let \mathbb{P} be a probability measure defined on the space of environments $(\Omega, \mathcal{B}(\Omega))$.

- We say that \mathbb{P} is *elliptic* if

(E) for every $x \in \mathbb{Z}^d$ we have that

$$\mathbb{P}\left[\min_{e \in U} \omega(x, e) > 0\right] = 1. \quad (2.6)$$

- We say that \mathbb{P} is *uniformly elliptic* if

(UE) there exists a constant $\kappa > 0$ such that for every $x \in \mathbb{Z}^d$ we have that

$$\mathbb{P}\left[\min_{e \in U} \omega(x, e) \geq \kappa\right] = 1. \quad (2.7)$$

We will usually call the environment (uniformly) elliptic in that case also.

Remark 2.4. *This labeling is motivated by operator theory where one has analogous definitions of elliptic and uniformly elliptic differential operators.*

The following auxiliary process will play a significant role in what follows.

Definition 2.5. (Environment viewed from the particle). Let (X_n) be a RWRE. We define the *environment viewed from the particle* (or also *the environmental process*) as the discrete time process

$$\bar{\omega}_n := t_{X_n} \omega,$$

for $n \geq 0$, with state space Ω .

Apart from taking values in a compact state space, another advantage of the environment viewed from the particle is that even under the averaged measure it is Markovian, as is shown in the next result following Sznitman [BS02]; however, the cost is that we now deal with an infinite dimensional state space.

Proposition 2.6. *Consider a RWRE in an environment with law \mathbb{P} . Then, under P_0 , the process $(\bar{\omega}_n)$ is Markovian with state space Ω , initial law \mathbb{P} , and transition kernel*

$$Rf(\omega) := \sum_{e \in U} \omega(0, e) f(t_e \omega), \quad (2.8)$$

defined for f bounded measurable on Ω and initial law \mathbb{P} .

PROOF. Let us first note that for every $x \in \mathbb{Z}^d$, and every bounded measurable function f on Ω ,

$$E_{x, \omega}[f(\bar{\omega}_1)] = E_{x, \omega}[f(t_{X_1} \omega)] = \sum_{e \in U} \omega(x, e) f(t_{x+e} \omega) = \sum_{e \in U} t_x \omega(0, e) f(t_e(t_x \omega)) = Rf(t_x \omega). \quad (2.9)$$

Let now $f_i, i = 0, \dots, n + 1$ be bounded measurable functions. Note that

$$\begin{aligned} E_{0,\omega}[f_{n+1}(\bar{\omega}_{n+1})(f_n(\bar{\omega}_n) \cdots f_0(\bar{\omega}_0))] &= E_{0,\omega}[f_{n+1}(t_{X_{n+1}}\omega) \cdots f_0(t_{X_0}\omega)] \\ &= E_{0,\omega}[E_{X_n,\omega}(f_{n+1}(t_{X_1}\omega))f_n(\bar{\omega}_n) \cdots f_0(\bar{\omega}_0)] \\ &= E_{0,\omega}[Rf_{n+1}(\bar{\omega}_n)f_n(\bar{\omega}_n) \cdots f_0(\bar{\omega}_0)], \end{aligned}$$

where in the second equality we took advantage of the Markov property of (X_n) under $P_{0,\omega}$, and in the last step we have used (2.9). Since $Rf_{n+1}(\bar{\omega}_n)$ is \mathcal{F}_n -measurable, where \mathcal{F}_n is the natural filtration of $(\bar{\omega}_n)$, it follows from the above that

$$E_{0,\omega}[f_{n+1}(\bar{\omega}_{n+1}) | \bar{\omega}_0, \dots, \bar{\omega}_n] = Rf_{n+1}(\bar{\omega}_n), \quad (2.10)$$

which proves the Markov property of the chain $(\bar{\omega}_n)$ under the measure $P_{0,\omega}$. It follows that the transition kernel for the quenched process is given by (2.8). Integrating $P_{0,\omega}$ with respect to \mathbb{P} we finish the proof. \square

2.2 Invariant probability measures of the environment as seen from the random walk

We now want to examine the invariant measures of the Markov chain $(\bar{\omega}_n)$. Given an arbitrary probability measure \mathbb{P} on Ω , we define the probability measure $\mathbb{P}R$ through the identity

$$\int Rf d\mathbb{P} = \int f d(\mathbb{P}R),$$

for every bounded continuous function f on Ω . Whenever $\mathbb{P} = \mathbb{P}R$, we will say that \mathbb{P} is an *invariant probability measure* for the environmental process. We will also need to consider the possibility of having invariant measures which are not necessarily probability measures: similarly to the above, we will say that a measure ν is invariant for the environmental process if for every bounded continuous function f one has that

$$\int f d\nu = \int Rf d\nu.$$

It is obvious that any degenerate probability measure which is translation invariant, is an invariant probability measure: this corresponds to any simple random walk. The following lemma is a standard result, but shows that there might be some other ways of constructing more interesting invariant probability measures. Recall that given a sequence of probability measures a limit measure is defined as the limit of any convergent subsequence.

Lemma 2.7. *Consider a RWRE and the corresponding environmental process $(\bar{\omega}_n)$. Then, if \mathbb{P} is any probability measure in Ω , there exists at least one limit measure of the Césaro means*

$$\frac{1}{n+1} \sum_{i=0}^n \mathbb{P}R^i. \quad (2.11)$$

Furthermore, every limit measure of this Césaro means is an invariant probability measure for the Markov chain $(\bar{\omega}_n)$.

PROOF. Let \mathbb{P} be an arbitrary probability measure defined on the space Ω . Denote for each $n \geq 0$ as ν_n the Césaro means of (2.11).

Since the space of probability measures defined on Ω is compact under the topology of weak convergence, we can extract a weakly convergent subsequence ν_{n_k} , so that the Césaro means has at least one limit point ν . We claim that ν is an invariant probability measure. Indeed, it is enough to prove that

$$\int Rf d\nu = \int f d\nu$$

for every bounded continuous function f . But since the transition kernel R maps bounded and continuous functions to bounded and continuous functions, we have that

$$\begin{aligned} \int Rf d\nu &= \lim_{k \rightarrow \infty} \int Rf d\nu_{n_k} = \lim_{k \rightarrow \infty} \frac{1}{n_k + 1} \sum_{i=0}^{n_k} \int f d(\nu R^{i+1}) \\ &= \lim_{k \rightarrow \infty} \left(\frac{1}{n_k + 1} \sum_{i=0}^{n_k} \int f d(\nu R^i) + \frac{1}{n_k + 1} \int f d(\nu R^{n_k+1}) - \frac{1}{n_k + 1} \int f d\nu \right) \\ &= \lim_{k \rightarrow \infty} \int f d\nu_{n_k} = \int f d\nu. \end{aligned}$$

□

Knowing only the existence of an invariant probability measure turns out not to be very helpful. We will see that what we really need is to find one which is absolutely continuous with respect to the law \mathbb{P} of the environment.

Example 2.8. *Let us consider the case $d = 1$. Assume **(E)** to be fulfilled and define*

$$\rho(x, \omega) := \frac{\omega(x, -1)}{\omega(x, 1)} \quad \text{and} \quad \rho(\omega) := \rho(0, \omega). \quad (2.12)$$

If $\mathbb{E}[\rho] < 1$ and the environment $(\omega(x))_{x \in \mathbb{Z}}$ is i.i.d. under the law \mathbb{P} , we will prove in this lecture that

$$\nu(d\omega) := f(\omega) \mathbb{P}(d\omega),$$

where

$$f(\omega) := C (1 + \rho(0, \omega)) (1 + \rho(1, \omega) + \rho(1, \omega)\rho(2, \omega) + \rho(1, \omega)\rho(2, \omega)\rho(3, \omega) + \dots) < \infty,$$

for some constant $C > 0$, is an invariant probability measure for the process $(\bar{\omega}_n)$, cf. also Theorem 2.12 below.

2.3 Transience and recurrence in the one-dimensional model

The focus of this section will be on one-dimensional RWRE under the assumption **(E)** and ergodicity properties of the law \mathbb{P} of the environment. In this context, we will derive explicit necessary and sufficient conditions in terms of the environment for the walk being transient or recurrent. It turns out that in this case the model is reversible in the following sense: for \mathbb{P} -a.a. environments ω it is possible to find a measure defined on \mathbb{Z} for which the random walk (X_n) in environment ω is reversible. This observation partly explains the fact that many explicit computations can be performed, and even explicit conditions characterizing particular behaviors of the walk can be found.

The following lemma of Kesten [Ke75] will prove useful.

Lemma 2.9. *Given any stationary sequence of random variables $(Y_n)_{n \geq 0}$ with law P such that $P[\lim_{n \rightarrow \infty} \sum_{k=0}^n Y_k = \infty] = 1$ one has $P[\liminf_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^n Y_k > 0] = 1$.*

In what follows, we will say that a function f is *Lebesgue integrable in the extended sense* if its Lebesgue integral exists, possibly taking the values ∞ or $-\infty$.

Theorem 2.10. *Consider a RWRE in dimension $d = 1$ in an environment with law \mathbb{P} such that **(E)** holds. Assume **(ERG)** and that $\mathbb{E}[\log \rho]$ is Lebesgue integrable in the extended sense. Then the following are satisfied.*

(i) *If $\mathbb{E}[\log \rho] < 0$ then the random walk is P_0 -a.s. transient to the right, i.e.,*

$$\lim_{n \rightarrow \infty} X_n = \infty, \quad P_0 - a.s.$$

(ii) *If $\mathbb{E}[\log \rho] > 0$ then the random walk is P_0 -a.s. transient to the left, i.e.,*

$$\lim_{n \rightarrow \infty} X_n = -\infty, \quad P_0 - a.s.$$

(iii) *If $\mathbb{E}[\log \rho] = 0$ then the random walk is P_0 -a.s. recurrent and*

$$\limsup_{n \rightarrow \infty} X_n = \infty \quad \text{and} \quad \liminf_{n \rightarrow \infty} X_n = -\infty \quad P_0 - a.s.$$

The above theorem was first proved within the context of branching processes in i.i.d. random environments by Smith and Wilkinson in 1969 [SW69] (see also [KZ13, Remark 8] and the references therein). In 1975 it was proved by Solomon [So75] for i.i.d. environments and afterwards extended to ergodic environments by Alili [Al99]. Here, we present a proof based on the method of Lyapunov functions (see Comets, Menshikov and Popov [CMP98] and Fayolle, Malyshev and Menshikov [FMM95]). The so-called Sinaï's regime corresponds to the recurrent case under the additional assumption that $0 < \mathbb{E}[(\log \rho)^2] < \infty$. In [Si82], Sinaï proved that under these conditions the position of the walk at time n is typically of order $(\log n)^2$ under P_0 . We will see in section 2.4, that the dichotomy expressed by Theorem 2.10 is an expression

of the different possibilities concerning the existence of an invariant measure (not necessarily a probability measure) for the environmental process which is absolutely continuous with respect to \mathbb{P} : (ii) and (i) occur when there exists such a measure; (iii) occurs when such a measure does not exist.

PROOF. We want to find a martingale defined in terms of the environment which discriminates between transience and recurrence through the use of the martingale convergence theorem. Let us furthermore try to find such a martingale of the form $f(X_n)$, where

$$f(x) = \sum_{j=0}^{x-1} \Delta_j,$$

for $x \geq 0$, and for some sequence (Δ_j) which will be chosen appropriately. In fact, using this convergence we will deduce the desired asymptotics from the properties of the limit of that martingale. Now note that with $q(x) := \omega(x, 1)$ and $p(x) := \omega(x, -1)$,

$$E_{x,\omega}[f(X_{n+1}) - f(X_n) | X_n = y] = \begin{cases} -p(y)\Delta_{y-1} + q(y)\Delta_y, & \text{if } y \geq 2, \\ p(1)\Delta_{-1} + q(1)\Delta_1, & \text{if } y = 1, \\ p(y)\Delta_{y-2} - q(y)\Delta_{y-1}, & \text{if } y \leq 0. \end{cases}$$

But if $f(X_n)$ is a martingale, the left-hand side of this display must vanish and we should have that

$$\Delta_1 = -\rho_1 \Delta_{-1},$$

and that

$$\begin{aligned} \Delta_y &= \rho_y \Delta_{y-1} & \text{for } y \geq 2, \\ \Delta_{y-2} &= \rho_y^{-1} \Delta_{y-1} & \text{for } y \leq 0, \end{aligned}$$

where we have used the shorthand notation $\rho_y := \rho(y, \omega)$. Choosing $\Delta_0 = -1$, $\Delta_1 = -\rho_1$ and $\Delta_{-1} = 1$ we deduce that

$$f(x) = \begin{cases} -\sum_{0 \leq j \leq x-1} \prod_{i=1}^j \rho_i, & \text{if } x \geq 0, \\ \sum_{x \leq j \leq -1} \prod_{i=j+1}^0 \rho_i^{-1}, & \text{if } x < 0, \end{cases}$$

serves our purposes, where $\prod_{i=1}^0 \rho_i := 1$. Hence, f is harmonic with respect to the generator of the quenched RWRE and $f(X_n)$, is an \mathcal{G}_n -martingale under the probability measure $P_{0,\omega}$, where \mathcal{G}_n is the natural σ -algebra generated by the random walk. Now, by the ergodic theorem, we have \mathbb{P} -a.s. that

$$\prod_{i=1}^x \rho_i = \exp \{x(\mathbb{E}[\log \rho] + o(1))\},$$

as $x \rightarrow \infty$, while when $x \rightarrow -\infty$, one has

$$\prod_{i=x+1}^{-1} \rho_i = \exp \{x(\mathbb{E}[\log \rho] + o(1))\}.$$

We now see that in the case $\mathbb{E}[\log \rho] < 0$, there is a constant $C > 0$ such that \mathbb{P} -a.s.

$$\lim_{x \rightarrow \infty} f(x) = -C, \quad (2.13)$$

and

$$\lim_{x \rightarrow -\infty} f(x) = \infty.$$

It follows that \mathbb{P} -a.s.

$$E_{0,\omega}[f(X_n)_-] = \sum_{x=1}^{\infty} f(x) P_{0,\omega}[X_n = x] < \infty.$$

By the martingale convergence theorem P_0 -a.s.

$$\lim_{n \rightarrow \infty} f(X_n) \text{ exists.} \quad (2.14)$$

Now, by ellipticity, it is easy to see that P_0 -a.s., only the following three possibilities can occur:

- (i) $\limsup_{n \rightarrow \infty} X_n = \infty$ and $\liminf_{n \rightarrow \infty} X_n = -\infty$.
- (ii) $\lim_{n \rightarrow \infty} X_n = \infty$.
- (iii) $\lim_{n \rightarrow \infty} X_n = -\infty$.

By (2.14) and (2.13) we conclude that necessarily case (ii) above occurs. By a similar analysis we see that if $\mathbb{E}[\log \rho] > 0$, case (iii) happens. Let us now consider the case

$$\mathbb{E}[\log \rho] = 0.$$

If ρ was almost surely constant and hence equal to 1, the above setting would be reduced to simple random walk, for which the corresponding result is canonical knowledge. Therefore, without loss of generality, we can assume that $\mathbb{E}[(\log \rho)^2] > 0$, and equally that $\mathbb{P}[\log \rho > 0] > 0$. Then, by Lemma 2.9 and the ergodicity of \mathbb{P} , we can conclude that \mathbb{P} -a.s.,

$$\limsup_{x \rightarrow \infty} \sum_{i=1}^x \log \rho_i > -\infty.$$

It follows that \mathbb{P} -a.s. one has that

$$\lim_{x \rightarrow \infty} f(x) = -\infty,$$

and similarly that

$$\lim_{x \rightarrow -\infty} f(x) = \infty.$$

If we define for $A > 0$ the stopping times $T_A := \inf\{k \geq 0 : X_k \geq A\}$ and $S_A := \inf\{k \geq 0 : X_k \leq -A\}$, we see that $f(X_{n \wedge T_A})$ and $f(X_{n \wedge S_A})$ are martingales such that $E_{0,\omega}[f(X_{n \wedge T_A})_+] < \infty$ and $E_{0,\omega}[f(X_{n \wedge S_A})_-] < \infty$, respectively. Hence, by the martingale convergence theorem we conclude that the limits

$$\lim_{n \rightarrow \infty} f(X_{n \wedge T_A}), \quad \lim_{n \rightarrow \infty} f(X_{n \wedge S_A}),$$

exist. The only possibility is that \mathbb{P} -a.s. we have that $P_{0,\omega}$ -a.s., X_n eventually hits both A and $-A$. Since A was chosen arbitrarily, this proves part (iii) of the theorem. \square

2.4 Computation of an absolutely continuous invariant measure in dimension $d = 1$

In 1999, Alili [A99] proved a one-dimensional result which establishes the existence of an invariant measure for the environment process as seen from the random walk with respect to the initial law of the environment. The proof we present here, is due to Conze and Guivarc'h [CG00] (see also [Re11]). We will say that **(B+)** is satisfied if

$$\mathbb{E} \left[(1 + \rho_0) \sum_{j=0}^{\infty} \prod_{k=0}^{j-1} \rho_{k+1} \right] < \infty,$$

while we will say that **(B-)** is satisfied if

$$\mathbb{E} \left[(1 + \rho_0^{-1}) \sum_{j=0}^{\infty} \prod_{k=-1}^{-j} \rho_k^{-1} \right] < \infty.$$

Note that in the i.i.d. case **(B+)** reduces to $\mathbb{E}[\rho_0] < 1$ while **(B-)** to $\mathbb{E}[\rho_0^{-1}] < 1$.

Theorem 2.11. (Alili) *Consider a RWRE with law \mathbb{P} fulfilling **(E)** and **(ERG)** in dimension $d = 1$. Then the following holds.*

- (i) *Assume that $\mathbb{E}[\log \rho] = 0$. If $\mathbb{E}[(\log \rho)^2] > 0$, then there are no invariant measures which are absolutely continuous with respect to \mathbb{P} . If $\mathbb{E}[(\log \rho)^2] = 0$, \mathbb{P} is the unique invariant measure of the environmental process absolutely continuous with respect to \mathbb{P} (up to multiplicative constants).*
- (ii) *If $\mathbb{E}[\log \rho] > 0$ but **(B+)** is not satisfied, or if $\mathbb{E}[\log \rho] < 0$ but **(B-)** is not satisfied, the environment viewed from the random walk has a unique invariant measure ν (up to multiplicative constants) which is absolutely continuous with respect to \mathbb{P} , but which is not a probability measure.*

(iii) If **(B+)** is satisfied, there exists a unique invariant probability measure ν which is absolutely continuous with respect to \mathbb{P} . Furthermore,

$$\frac{d\nu}{d\mathbb{P}} = C(1 + \rho_0) \sum_{j=0}^{\infty} \prod_{k=0}^{j-1} \rho_{k+1},$$

for some constant $C > 0$.

(iv) If **(B-)** is satisfied, there exists a unique invariant probability measure ν which is absolutely continuous with respect to \mathbb{P} . Furthermore

$$\frac{d\nu}{d\mathbb{P}} = C(1 + \rho_0^{-1}) \sum_{j=0}^{\infty} \prod_{k=-1}^{-j} \rho_k^{-1},$$

for some constant $C > 0$.

We will see soon how this result exhibits a relationship between the existence of an absolutely continuous invariant probability measure and the ballisticity of the random walk: in dimension $d = 1$, the existence of an absolute continuous invariant probability measure is equivalent to ballisticity. We will give more details about this soon.

Sketch of the proof. We will start proving part (ii). Note that if ν is an invariant measure, we have that for every bounded measurable function f

$$\int (q(0, \omega)f(t_1\omega) + p(0, \omega)f(t_{-1}\omega)) \nu(d\omega) = \int f(\omega) \nu(d\omega).$$

Now if ν is absolutely continuous with respect to \mathbb{P} with density ϕ , the above equation is equivalent to

$$q(0, t_{-1}\omega)\phi(t_{-1}\omega) + p(0, t_1\omega)\phi(t_1\omega) = \phi(\omega)$$

holding for ν -a.a. ω . We then have that

$$h \circ t_1^2 - \left(\frac{1}{1-q} h \right) \circ t_1 + \rho^{-1} h = 0,$$

where $h := p\phi$ and where we have written $p = p(0, \omega)$ and $q = q(0, \omega)$. If we now define

$$\tilde{h} := h \circ t_1 - \rho^{-1} h,$$

we conclude that for every $x \in \mathbb{Z}$,

$$\tilde{h} \circ t_x - \tilde{h} = 0.$$

But since \mathbb{P} is ergodic with respect to $(t_x)_{x \in \mathbb{Z}}$, we conclude that \tilde{h} is \mathbb{P} -a.s. equal to a constant C . Assume that $C = 0$. Then $h = 0$ is equivalent to

$$h(t_1\omega) = \rho^{-1}(\omega)h(\omega).$$

We claim that the only solution in this case is $h = 0$. Indeed, using induction on n we have that

$$h(t_n\omega) = h(\omega) \prod_{j=0}^{n-1} \rho^{-1}(t_j\omega).$$

If $\mathbb{E}[\log \rho] > 0$, by the ergodic theorem this would imply that a.s.

$$\lim_{n \rightarrow \infty} h(t_n\omega) = 0.$$

Now integrating with respect to \mathbb{P} , using its stationarity and the fact that $h(t_n\omega)$, $n \in \mathbb{N}$, are uniformly integrable, we conclude that

$$\int h(\omega) \mathbb{P}(d\omega) = \lim_{n \rightarrow \infty} \int h(t_n\omega) \mathbb{P}(d\omega) = 0,$$

so that

$$h = 0.$$

Using a similar argument one arrives at the same conclusion when $\mathbb{E}[\log \rho] < 0$.

So let us assume that $C \neq 0$. In this case we have that

$$h = (\rho^{-1}h) \circ t_{-1} + C. \tag{2.15}$$

Now choose a constant h_0 and define recursively

$$h_{n+1} := (\rho^{-1}h_n) \circ t_1^{-1} + C. \tag{2.16}$$

If we can prove that h_n converges \mathbb{P} -a.s. as $n \rightarrow \infty$, then the limit should be a solution to (2.15). Now from (2.16) we can deduce

$$h_n(\omega) = C \sum_{j=0}^{n-1} \prod_{k=0}^{j-1} \rho^{-1}(t_{k+1}^{-1}\omega) + \left(\prod_{k=1}^n \rho^{-1}(t_k^{-1}\omega) \right) h_0.$$

Taking the limit when $n \rightarrow \infty$, we conclude in the case in which $\mathbb{E}[\log \rho] > 0$, in combination with the ergodic theorem, that h_n converges \mathbb{P} -a.s. to

$$h(\omega) = c \sum_{j=0}^{\infty} \prod_{k=0}^{j-1} \rho^{-1}(t_{k+1}^{-1}\omega).$$

Thus,

$$\phi(\omega) = (1 + \rho^{-1}(\omega)) \sum_{j=0}^{\infty} \prod_{k=-1}^{-j} \rho^{-1}(t_k\omega).$$

This proves part (ii) of the proposition. To prove part (iii), note that Jensen's inequality and **(B-)** imply that $\mathbb{E}[\log \rho] < \infty$. Therefore, the measure with density ϕ already defined can be

normalized to define a probability measure. Similarly, one can prove part (iv). The proof of part (i) in the case $\mathbb{E}[(\log \rho)^2] > 0$ is analogous to the proof of the recurrent case of Theorem 2.10. The case $\mathbb{E}[(\log \rho)^2] = 0$ is trivial, since in this case we would be in the situation of simple random walk.

2.5 Absolutely continuous invariant measures and some implications

The existence of an invariant probability measure which is absolutely continuous with respect to the initial distribution of the environment will turn out to be crucial in the study of the model. We recall that the environmental process has been defined in Definition 2.5, which considered as a trajectory has state space $\Gamma := \Omega^{\mathbb{N}}$. Furthermore, define the law P_ω defined on its Borel σ -algebra $\mathcal{B}(\Gamma)$ through the identity

$$P_\omega[A] := P_{0,\omega}[(\bar{\omega}_n) \in A], \quad (2.17)$$

for any Borel subset A of Γ endowed with the product topology. Furthermore, for any probability measure ν defined in Ω , we define

$$P_\nu := \int P_\omega \nu(d\omega). \quad (2.18)$$

We will denote by $\theta : \Gamma \rightarrow \Gamma$ the canonical shift on Γ defined by

$$\theta(\omega_0, \omega_1, \dots) := (\omega_1, \omega_2, \dots). \quad (2.19)$$

The following result of Theorem 2.12 was proved by Kozlov in [Ko85]. For its proof, we will follow Sznitman in [BS02].

Theorem 2.12. *(Kozlov) Consider a RWRE in an environment with law \mathbb{P} fulfilling **(E)** and **(ERG)**. Assume that there exists an invariant probability measure ν for the environment seen from the random walk which is absolutely continuous with respect to \mathbb{P} . Then the following are satisfied:*

- (i) ν is equivalent to \mathbb{P} .
- (ii) The environment as seen from the random walk with initial law ν is ergodic.
- (iii) ν is the unique invariant probability measure for the environment as seen from the particle which is absolutely continuous with respect to \mathbb{P} .
- (iv) The Césaro means

$$\frac{1}{n+1} \sum_{i=0}^n \mathbb{P}R^i$$

converges weakly to ν .

Proof of part (i). Let f be the Radon-Nikodym derivative of ν with respect to \mathbb{P} and consider the event $E := \{f = 0\}$. In order to prove the desired result it will be sufficient to show $\mathbb{P}[E] = 0$.

Since ν is invariant, we have that

$$\int f \cdot (R1_E) d\mathbb{P} = (\nu R)[E] = \nu[E] = \int_{\{f=0\}} d\mathbb{P} = 0.$$

It follows that \mathbb{P} -a.s. on the event $E^c = \{f > 0\}$ one has that $R1_E = 0$. Therefore, using the fact that $R1_E \leq 1$, one has that for every $e \in U$,

$$1_E(\omega) \geq R1_E(\omega) = \sum_{e' \in U} \omega(0, e') 1_E(t_{e'}\omega) \geq \omega(0, e) 1_E(t_e\omega), \quad \mathbb{P} - a.a. \omega.$$

From the ellipticity assumption and the fact that $1_E(\omega)$ and $1_E(t_e\omega)$ for $e \in U$ only take the values 0 or 1 we have that for such e ,

$$1_E(\omega) \geq 1_E(t_e\omega), \quad \mathbb{P} - a.s.$$

Now using the fact that $\mathbb{P}[E] = \mathbb{P}[t_e^{-1}E]$ we conclude that for each $e \in U$ one has

$$1_E = 1_{t_e^{-1}E}, \quad \mathbb{P} - a.a. \omega.$$

Thus, we iteratively obtain that for each $x \in \mathbb{Z}^d$,

$$1_E = 1_{t_x^{-1}(E)}, \quad \mathbb{P} - a.s.$$

It follows that the event

$$\tilde{E} := \bigcap_{x \in \mathbb{Z}^d} t_x^{-1}(E),$$

is invariant under the action of the family $(t_y)_{y \in \mathbb{Z}^d}$ and that it differs from the event E on an event of \mathbb{P} -probability 0. Since \mathbb{P} is ergodic with respect to the family $(t_y)_{y \in \mathbb{Z}^d}$ we conclude that

$$\mathbb{P}[E] = \mathbb{P}[\tilde{E}] \in \{0, 1\}. \tag{2.20}$$

But since $\int_{E^c} f d\mathbb{P} = \int f d\mathbb{P} = 1$ we know that $\mathbb{P}[E^c] > 0$, which in combination with $\mathbb{P}[\Omega] = 1$ and (2.20) implies $\mathbb{P}[E] = 0$. Hence, \mathbb{P} is equivalent to ν .

Proof of part (ii). We will prove that if $A \in \mathcal{B}(\Gamma)$ is invariant so that $\theta^{-1}(A) = A$ then $P_\nu[A]$ (cf. (2.18) and (2.19)) is equal to 0 or 1. For $\omega \in \Omega$ define

$$\phi(\omega) := P_\omega[A].$$

We claim that

$$(\phi(\bar{\omega}_n))_{n \geq 0}$$

is a P_ν -martingale with the canonical filtration on Γ . In fact, note that since A is invariant, we have that $1_A = 1_A \circ \theta_n$ and hence,

$$E_\nu[1_A | \bar{\omega}_0, \dots, \bar{\omega}_n] = E_\nu[1_A \circ \theta_n | \bar{\omega}_0, \dots, \bar{\omega}_n] = P_{\bar{\omega}_n}[A] = \phi(\bar{\omega}_n), \quad P_\nu - a.a. (\bar{\omega}_n). \quad (2.21)$$

It follows from (2.21) and the martingale convergence theorem that

$$\lim_{n \rightarrow \infty} \phi(\bar{\omega}_n) = 1_A((\bar{\omega}_n)_{n \in \mathbb{N}}), \quad P_\nu - a.a. (\bar{\omega}_n) \quad (2.22)$$

Let us now prove that there is a set $B \in \mathcal{B}(\Omega)$ such that ν -a.s.

$$\phi = 1_B. \quad (2.23)$$

In fact, assume that (2.23) is not satisfied. Then there is an interval $[a, b] \subset (0, 1)$ with $a < b$ such that

$$\nu[\phi \in [a, b]] > 0. \quad (2.24)$$

Also, by the ergodic theorem we have that P_ν -a.s.

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} 1_{\phi^{-1}([a, b])}(\bar{\omega}_k) = \Psi := E_\nu[1_{\phi^{-1}([a, b])}(\bar{\omega}_0) | \mathcal{I}],$$

where $\mathcal{I} := \{A \in \Gamma : \theta^{-1}(A) = A\}$ is the σ -field of invariant events. Now, by (2.24),

$$E_\nu[\Psi] = P_\nu[\phi(\omega_0) \in [a, b]] = \nu[\phi \in [a, b]] > 0.$$

But this contradicts (2.22). Hence, (2.23) holds. Let us now prove that ν -a.s.

$$R1_B = 1_B. \quad (2.25)$$

Indeed, we have that P_ν -a.s. it is true that

$$1_B(\omega_0) = E_\nu[1_B(\omega_1) | \omega_0] = R1_B(\omega_0).$$

Since $P_\nu[A] = \nu[B]$, it is then enough to prove that

$$\nu[B] \in \{0, 1\}. \quad (2.26)$$

Now, \mathbb{P} -a.s. we have that

$$1_B(\omega) = R1_B(\omega) = \sum_{|e|_1=1} \omega(0, e) 1_B(t_e \omega).$$

Using ellipticity, this implies that $\mathbb{P}[B] \in \{0, 1\}$, which again by part (i) of this theorem implies (2.26).

Proof of parts (iii) and (iv). Let g be a bounded measurable function on Ω . Let ν be any invariant probability measure for the transition kernel R that is absolutely continuous with respect to \mathbb{P} . By part (ii) and the ergodic theorem we have that P_ν -a.s.

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} g(\omega_k) = \int g d\nu.$$

Now, by part (i) of this theorem, the above convergence is also occurs $P_{\mathbb{P}}$ -a.s. Hence, we have that

$$\lim_{n \rightarrow \infty} E_0 \left[\frac{1}{n} \sum_{k=0}^{n-1} g(\omega_k) \right] = \int g d\nu.$$

This proves the uniqueness of ν and part (iv). \square

An important generalization of Kozlov's theorem was obtained by Rassoul-Agha in [RA03]. There, he shows that under the assumption that the random walk is directionally transient, the environment satisfies a certain mixing and uniform ellipticity condition, and if there exists an invariant probability measure which is absolutely continuous with respect to the initial law \mathbb{P} in certain half-spaces, a conclusion analogous to Kozlov's theorem holds.

In [Le13], Lenci generalizes Kozlov's theorem to environments which are not necessarily elliptic. Lenci admits the possibility that the environment is ergodic with respect to some subgroup Γ strictly smaller than \mathbb{Z}^d , which is a stronger condition than total ergodicity, and which enables him to relax the ellipticity condition. Furthermore, in Bolthausen-Sznitman [BS02a], an example of a RWRE which does not satisfy the ellipticity condition **(E)** and for which there are no invariant probability measures for the environmental process which are absolutely continuous with respect to the initial law of the environment is presented (see also [RA03]).

2.6 The law of large numbers, directional transience and ballisticity

For the purposes of applying Kozlov's theorem, it would be important to understand how to reconstruct the random walk from the canonical environmental process. Now, let us note that if we denote by Ω_{per} the periodic environments so that

$$\Omega_{per} := \{\omega \in \Omega : \omega = t_x \omega \text{ for some } x \in \mathbb{Z}^d, x \neq 0\},$$

whenever $\omega \in \Omega \setminus \Omega_{per}$ and ω' is a translation of ω , this translation is uniquely defined. This observation would enable us to express the increments of the random walk as a function of the environmental process whenever the initial condition is not periodic. Assuming that the initial law \mathbb{P} of the environment is ergodic, and noting that the set of periodic environments is invariant under translations, we can see that $\mathbb{P}[\Omega_{per}]$ equals either 0 or 1. Nevertheless, assuming **(ERG)**, may happen that $\mathbb{P}[\Omega_{per}] = 1$, a situation where a priori we cannot perform this reconstruction (and which is impossible if we assume even **(IID)**). We will therefore prove directly the ergodicity of the increments of the random walk.

Our first application of Kozlov's theorem will relate the so-called transient regime with the ballistic one.

Definition 2.13. (Transience in a given direction) For $l \in \mathbb{S}^{d-1}$ define the event

$$A_l := \left\{ \lim_{n \rightarrow \infty} X_n \cdot l = \infty \right\} \quad (2.27)$$

of directional transience in direction l . We will call a RWRE *transient in direction l* if $P_0[A_l] = 1$.

Definition 2.14. (Ballisticity in a given direction) Let $l \in \mathbb{S}^d$. We say that a RWRE is ballistic in direction l , if P_0 -a.s.

$$\liminf_{n \rightarrow \infty} \frac{X_n \cdot l}{n} > 0. \quad (2.28)$$

We will see in Chapter 3, that in fact the limit in the left-hand side of (2.28) always exists, and is even known to be deterministic in dimensions $d = 2$.

Let us now consider for each $x \in \mathbb{Z}^d$ the *local drift* at site x defined as

$$d(x, \omega) := \sum_{e \in U} \omega(x, e) e = E_{x, \omega}[X_1 - X_0].$$

We then have the following corollary to Kozlov's theorem.

Corollary 2.15. *Consider a RWRE in an environment with law \mathbb{P} fulfilling **(E)** and **(ERG)**. Furthermore, assume that there exists an invariant probability measure for the environment seen from the particle, denoted by ν , which is absolutely continuous with respect to \mathbb{P} . Then a law of large number is satisfied so that $P_{0, \mathbb{P}}$ -a.s.*

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = \int d(0, \omega) \nu(d\omega) =: v.$$

Furthermore, if the walk is transient in a given direction l , it is necessarily ballistic in that direction so that $v \cdot l \neq 0$.

PROOF. We will follow Sabot [Sa12]. Define for $n \geq 1$,

$$\Delta X_n := X_n - X_{n-1}.$$

This is a process with state space $\mathcal{U} := U^{\mathbb{N}}$. In a slight abuse of notation to (2.19), we define the canonical shift $\theta : \mathcal{U} \rightarrow \mathcal{U}$ via

$$\theta(\Delta X_1, \Delta X_2, \dots) := (\Delta X_2, \Delta X_3, \dots). \quad (2.29)$$

Note that the process $(\Delta X_n)_{n \geq 1}$ is stationary under the law $P_{0, \nu}$. We will show that in fact the transformation θ is ergodic with respect to the space $(\mathcal{U}, \mathcal{B}(\mathcal{U}), P_{0, \nu})$, where $\mathcal{B}(\mathcal{U})$ is the Borel σ -field of \mathcal{U} . Let $A \in \mathcal{B}(\mathcal{U})$ be invariant so that $\theta^{-1}(A) = A$ and define

$$\psi(x, \omega) := P_{x, \omega}[(\Delta X_n) \in A].$$

We claim that

$$(\psi(X_n, \omega))_{n \geq 0}$$

is a martingale with respect to the canonical filtration on \mathcal{U} generated by (X_n) . Indeed,

$$P_{0,\omega}[(\Delta X_m) \in A \mid X_0, \dots, X_n] = P_{X_n, \omega}[(\Delta X_m) \in A] = \psi(X_n, \omega).$$

Therefore, taking the limit when $n \rightarrow \infty$, and for any ω , the martingale convergence theorem yields that

$$\lim_{n \rightarrow \infty} \psi(0, \bar{\omega}_n) = \lim_{n \rightarrow \infty} \psi(X_n, \omega) = 1_A((\Delta X_n)) \quad P_{0,\omega} - a.s. \quad (2.30)$$

We now have by the ergodic theorem and Kozlov's theorem that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^n \psi(0, \bar{\omega}_n) = \int \psi(0, \omega) \nu(d\omega) \quad P_{0,\nu} - a.s.$$

The limit (2.30) now implies that

$$P_{0,\nu}[(\Delta X_n) \in A] \in \{0, 1\},$$

which gives us the claimed ergodicity. We thus have that

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n \Delta X_k = \int d(0, \omega) \nu(d\omega) \quad P_{0,\nu} - a.s.$$

By Kozlov's theorem, we can conclude that the above convergence occurs $P_{0,\mathbb{P}}$ -a.s. The second claim of the corollary is immediate from Lemma 2.9 above. \square

Rassoul-Agha in [RA03], obtains a version of Corollary 2.15 where transience is replaced by the so-called Kalikow's condition [Ka81], a stronger mixing assumption than ergodicity is required, but it is necessary only to assume the existence of an invariant probability measure which is absolutely continuous with respect to the initial law only on appropriate half-spaces.

On the other hand, combining Corollary 2.15 with Theorem 2.11, we can now easily derive the following result for the one-dimensional case, originally proved by Solomon [So75] for the i.i.d. case and later extended by Alili [Al99] to the ergodic case.

Theorem 2.16. *Consider a RWRE in dimension $d = 1$ in an environment with law \mathbb{P} fulfilling **(E)** and **(ERG)**. Then, there exists a deterministic $v \in \mathbb{R}^d$ such that*

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = v, \quad P_0 - a.s.$$

Furthermore,

(i) If **(B+)** is satisfied, then

$$v = \mathbb{E} \left[(1 - \rho_0) \sum_{j=0}^{\infty} \prod_{k=0}^{j-1} \rho_{k+1} \right].$$

(ii) If $(\mathbf{B-})$ is satisfied, then

$$v = \mathbb{E} \left[(1 - \rho_0^{-1}) \sum_{j=0}^{\infty} \prod_{k=-1}^{-j} \rho_{-k}^{-1} \right].$$

(iii) If neither $(\mathbf{B+})$ nor $(\mathbf{B-})$ are satisfied, then

$$v = 0.$$

Since in case (iii) of the above one has that X_n/n converges to 0, one immediately is led to the question of the typical order of X_n in this case. The answer to this problem (and further interesting insight) has been obtained by Kesten, Kozlov and Spitzer [KKS75]: In fact, there is a direct connection between the exponent $\kappa \in (0, 1)$ characterized by

$$\mathbb{E}[\rho^\kappa] = 1,$$

and the typical order of X_n in this case, which is n^κ . We refer the reader to [KKS75] for further details.

In addition, from the above discussion we see that in dimension $d = 1$, if the family of integer shifts is ergodic with respect to the law \mathbb{P} of the environment, the walk being transient to the right or left does not ensure the existence of an invariant probability measure for the environmental process which is absolutely continuous with respect to \mathbb{P} . Let us give two examples which show that this situation could also occur for dimensions $d \geq 2$.

Example 2.17. Let $d = 2$. Consider a random walk in an environment $(\omega(x))_{x \in \mathbb{Z}^2}$ of the form $\omega(x) := (\omega(x, e))_{e \in U}$ with a law \mathbb{P} such that $\mathbb{P}[\omega(x, e) = 1/4] = 1$ for $e = e_2$ and $e = -e_2$ and $\omega(x, e_1) = q(x)$ while $\omega(x, -e_1) = p(x) = \frac{1}{2} - q(x)$, with $\mathbb{E}[\log(p(x)/q(x))] < 0$ and $\mathbb{E}[p(x)/q(x)] = 1$. Assume also that for every $x \in \mathbb{Z}^2$, $(\omega(x + ne_1))_{n \in \mathbb{Z}}$ are i.i.d. under \mathbb{P} while

$$\mathbb{P}[\omega(x + e_2) = \omega(x)] = 1$$

In other words, the environment is constant in the direction e_2 , but it is i.i.d. in the direction e_1 , see Figure 2.1 also. It is easy to check that the shifts $(\theta_x)_{x \in \mathbb{Z}^d}$ form an ergodic family with respect to \mathbb{P} . Also, the walk is transient in direction e_1 , but not ballistic in that direction and there are no invariant probability measures for the environmental process which are absolutely continuous with respect to \mathbb{P} (cf. Corollary 2.15).

Example 2.18. Let $\epsilon > 0$. Furthermore, take ϕ to be any random variable taking values on the interval $(0, 1/4)$ and such that the expected value of $\phi^{-1/2}$ is infinite, while for every $\epsilon > 0$, the expected value of $\phi^{-(1/2-\epsilon)}$ is finite. Let Z be a Bernoulli random variable of parameter $1/2$. We now define $\omega_1(0, e_1) = 2\phi$, $\omega_1(0, -e_1) = \phi$, $\omega_1(0, -e_2) = \phi$ and $\omega_1(0, e_2) = 1 - 4\phi$ and $\omega_2(0, e_1) = 2\phi$, $\omega_2(0, -e_1) = \phi$, $\omega_2(0, e_2) = \phi$ and $\omega_2(0, -e_2) = 1 - 4\phi$. We then let the environment at site 0 be given by the random variable $\omega(0, \cdot) := Z\omega_1(0, \cdot) + (1 - Z)\omega_2(0, \cdot)$,

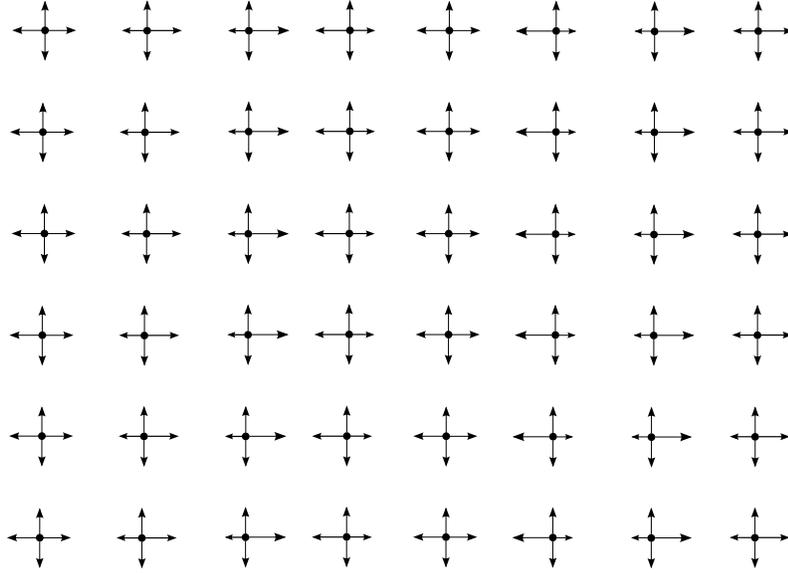


Figure 2.1: A sketch of an environment which is i.i.d. in direction e_1 and constant in direction e_2 .

and extend this to an i.i.d. environment on \mathbb{Z}^d . This environment has the property that traps can appear, where the random walk gets caught in an edge, as shown in Figure 2.2. Furthermore, as we will show, it is not difficult to check that the random walk in this random environment is transient in direction e_1 but not ballistic. Hence, due to Corollary 2.15 there exists no invariant probability measure for the environment seen from the particle, which in addition is absolutely continuous with respect to \mathbb{P} .

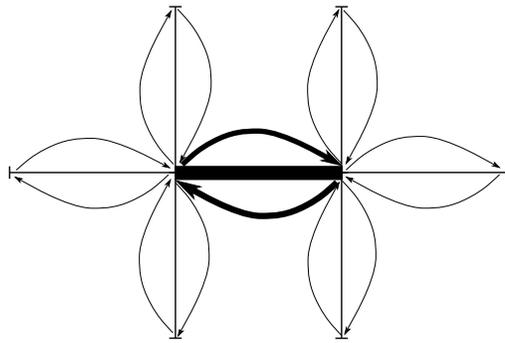


Figure 2.2: A trap produced by an elliptic environment.

These are two examples of walks which are transient in a given direction but not ballistic, and for which there is no invariant probability measure for the environmental process absolutely

continuous with respect to the initial law \mathbb{P} of the environment. It is natural hence to raise the following questions:

Open question 2.19. *Assume given a RWRE fulfilling **(ERG)** and **(E)**. Furthermore assume the RWRE is transient in a given direction. Is the existence of an invariant probability measure for the environmental process which is absolutely continuous with respect to \mathbb{P} equivalent to ballisticity in the given direction?*

Open question 2.20. *Let $d \geq 2$. Assume given a RWRE for which **(UE)** and **(IID)** are fulfilled, and which is transient in direction $l \in \mathbb{S}^{d-1}$. Is the RWRE necessarily ballistic in direction l ?*

As it is discussed above, example 2.18 shows that if the hypothesis **(UE)** is replaced by **(E)** in the open question 2.20, then its answer is negative. The following proposition gives an indication of how much ellipticity should be required.

Proposition 2.21. *Consider a random walk in an i.i.d. environment. Assume that*

$$\max_{e \in U} \mathbb{E} \left[\frac{1}{1 - \omega(0, e)\omega(0, -e)} \right] = \infty. \quad (2.31)$$

Then the walk is not ballistic in any direction.

PROOF. Fix $e \in U$ and define the first exit time of the random walk from the edge between 0 and e as

$$T_{\{0, e\}} := \min \{n \geq 0 : X_n \notin \{0, e\}\}.$$

We then have for every $k \geq 0$, using the notation $\omega_1 := \omega(0, e)$ and $\omega_2 := \omega(0, -e)$. that

$$P_{0, \omega}[T_{\{0, e\}} > 2k] = (\omega_1 \omega_2)^k$$

and

$$\sum_{k=0}^{\infty} P_{0, \omega}[T_{\{0, e\}} > 2k] = \frac{1}{1 - \omega_1 \omega_2}. \quad (2.32)$$

Using (2.31), this implies that

$$E_0[T_{\{0, e\}}] = \infty.$$

We can now show using the strong Markov property under the quenched measure and the i.i.d. nature of the environment, that for each natural $m > 0$, the time $T_m := \min\{n \geq 0 : X_n \cdot l > m\}$ can be bounded from below by the sum of a sequence of random variables $\tilde{T}_1, \dots, \tilde{T}_m$ which under the averaged measure are i.i.d. and distributed as $T_{\{0, e\}}$. This proves that P_0 -a.s. $T_m/m \rightarrow \infty$ which implies that the random walk is not ballistic in direction l . \square

Based now on Proposition 2.21 we have the following extended version of the open question 1.

Open question 2.22. *Let $d \geq 2$. Is it the case that every random walk fulfilling **(E)** and **(IID)**, and satisfying*

$$\max_{e \in U} \mathbb{E} \left[\frac{1}{1 - \omega(0, e)\omega(0, -e)} \right] < \infty,$$

and which is transient in direction $l \in \mathbb{S}^{d-1}$, is ballistic in direction l ?

For the case of an environment fulfilling **(IID)** and having a Dirichlet law, the above question was answered positively by Sabot [Sa12] in dimensions $d \geq 3$ (see also the work of Campos and Ramírez [CR13]).

2.7 Transience, recurrence and a quenched invariance principle

Similarly to the case of simple random walk, one of the most basic questions for RWRE is a classification in terms of transience and recurrence. As simple as this question is to pose, it is still far from being completely understood. In fact, a natural question is the following one.

Open question 2.23. *Is it the case that in dimensions $d \geq 3$, a RWRE fulfilling **(E)** and **(IID)** is transient?*

This question has been answered only in the case of the so-called Dirichlet environment (see Sabot [Sa11]) and essentially also for balanced environments (see Lawler [La82]). It is intimately related to the quenched central limit theorem. In this section, we will discuss how Kozlov's theorem can be used for balanced random walks to derive such a theorem, from which eventually transience in direction $d \geq 3$ can be deduced.

Consider the subset the set of environments

$$\Omega_0 := \{\omega \in \Omega : \omega(x, e) = \omega(x, -e) \text{ for all } x \in \mathbb{Z}^d, e \in U\}.$$

We will say that the law \mathbb{P} of the environment of a RWRE is *balanced* if

$$\mathbb{P}[\Omega_0] = 1,$$

where in particular we use that Ω_0 is a measurable subset of Ω . The following result was proved by Lawler in [La82].

Theorem 2.24. *Consider a random walk with an environment which has a law \mathbb{P} fulfilling **(UE)** as well as **(ERG)**, and which is balanced. Then there exists an invariant measure for the environmental process which is absolutely continuous with respect to \mathbb{P} .*

The above result is one of the few instances in which it has been possible to construct an absolutely continuous invariant measure for the environmental process in dimensions $d \geq 2$ (for non-nestling random walks at low disorder Bolthausen and Sznitman also make such a construction in [BS02a]; for random environment with Dirichlet law Sabot characterizes the cases when this happens in [Sa12]). As a corollary, Lawler can prove the following.

Corollary 2.25. *Under the conditions of Theorem 2.24 for \mathbb{P} -a.e. ω , under $P_{0,\omega}$, the sequence $X_{[n\cdot]}/\sqrt{n}$ converges in law on the Skorokhod space $D([0, \infty); \mathbb{R}^d)$ to a non-degenerate Brownian motion with a diagonal and deterministic covariance matrix $A := \{a_{i,j}\}$, $a_{i,j} = a_i \delta_{i,j}$.*

PROOF. Let us first explain how to prove the convergence of the finite-dimensional distributions. Note that for every $\theta \in \mathbb{R}^d$ sufficiently close to 0 and using $\mathbb{P}[\Omega_0] = 1$ we have that

$$e^{iX_n \cdot \theta - \sum_{k=0}^{n-1} \ln(2 \sum_{j=1}^d \cos(e_j \cdot \theta) \omega(X_k, e_j))}$$

is a martingale in n with respect to the law $P_{0,\omega}$. Therefore, rescaling θ by θ/\sqrt{n} we see that for all n large enough,

$$E_{0,\omega} \left[e^{i \frac{X_n}{\sqrt{n}} \cdot \theta - \sum_{k=0}^{n-1} \ln(2 \sum_{j=1}^d \cos\left(\frac{\theta_j}{\sqrt{n}}\right) \omega(X_k, e_j))} \right] = 1.$$

Hence, it is enough to prove that there exist constants $\{a_i : 1 \leq i \leq d\}$ such that P_0 -a.s. one has that

$$\lim_{n \rightarrow \infty} \sum_{k=0}^{n-1} \ln \left(2 \sum_{j=1}^d \cos\left(\frac{\theta_j}{\sqrt{n}}\right) \omega(X_k, e_j) \right) = - \sum_{j=1}^d \frac{a_j}{2} \theta_j^2. \quad (2.33)$$

Now, by Taylor's theorem,

$$\cos(x) = 1 - \frac{x^2}{2!} + h_1(x)x^2,$$

where $\lim_{x \rightarrow 0} h_1(x) = 0$. Hence,

$$\cos\left(\frac{\theta_j}{\sqrt{n}}\right) = 1 - \frac{\theta_j^2}{2n} + \frac{\theta_j^2}{n} h_1\left(\frac{\theta_j}{\sqrt{n}}\right),$$

and for each $k \geq 0$,

$$2 \sum_{j=1}^d \cos\left(\frac{\theta_j}{\sqrt{n}}\right) \omega(X_k, e_j) = 1 - \sum_{j=1}^d \frac{\theta_j^2}{n} \omega(X_k, e_j) + 2 \sum_{j=1}^d \frac{\theta_j^2}{n} h_1\left(\frac{\theta_j}{\sqrt{n}}\right) \omega(X_k, e_j). \quad (2.34)$$

A second application of Taylor's theorem gives that

$$\ln(1-x) = -x + h_2(x)x,$$

where $\lim_{x \rightarrow 0} h_2(x) = 0$. Thus, using (2.34) we have that,

$$\begin{aligned} \ln \left(2 \sum_{j=1}^d \cos\left(\frac{\theta_j}{\sqrt{n}}\right) \omega(X_k, e_j) \right) &= - \sum_{j=1}^d \frac{\theta_j^2}{n} \bar{\omega}_k(0, e_j) + 2 \sum_{j=1}^d \frac{\theta_j^2}{n} h_1\left(\frac{\theta_j}{\sqrt{n}}\right) \bar{\omega}_k(0, e_j) \\ &\quad + \left(\sum_{j=1}^d \frac{\theta_j^2}{n} \bar{\omega}_k(0, e_j) - 2 \sum_{j=1}^d \frac{\theta_j^2}{n} h_1\left(\frac{\theta_j}{\sqrt{n}}\right) \bar{\omega}_k(0, e_j) \right) h_2, \end{aligned}$$

where

$$h_2 = h_2 \left(\sum_{j=1}^d \frac{\theta_j^2}{n} \bar{\omega}_k(0, e_j) - 2 \sum_{j=1}^d \frac{\theta_j^2}{n} h_1 \left(\frac{\theta_j}{\sqrt{n}} \right) \bar{\omega}_k(0, e_j) \right),$$

and where we recall that the environmental process $(\bar{\omega}_n)$ has been introduced in Definition 2.5. It then follows that if we are able to prove that for each $1 \leq j \leq d$, P_0 -a.s. one has that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^n \bar{\omega}_k(0, e_j) = \frac{a_j}{2}, \quad (2.35)$$

then we have proven (2.33). To prove (2.35), by Kozlov's theorem, it is enough to use Theorem 2.24 which ensures the existence of a measure ν which is an invariant measure for the process $(\bar{\omega}_n)$ and which is absolutely continuous with respect to \mathbb{P} . To prove the convergence to Brownian motion we can use the martingale convergence theorem ([Sz04]). \square

We will now explain the main ideas in the proof of Theorem 2.24. the details of which can be found for example in Sznitman [BS02]. We will construct an invariant measure by approximating it with invariant measures with respect to the environmental processes on finite spaces. Configurations of the environment on these finite spaces will then correspond to periodic configurations on the full space. The point is to do this in such a way that the density of these invariant measures with respect to periodized versions of the measure \mathbb{P} , has an L_p norm for some $p > 1$, which is uniformly bounded in the size of the boxes.

We introduce for $x \in \mathbb{Z}^d$ the equivalence classes

$$\hat{x} := x + (2N + 1)\mathbb{Z}^d \in \mathbb{Z}^d / ((2N + 1)\mathbb{Z}^d).$$

In addition we define for $\omega \in \Omega_0$ the corresponding periodized version ω_N of ω so that $\omega_N(y) = \omega(x)$ for $y \in \mathbb{Z}^d$ and $x \in B_\infty(N)$ such that $\hat{y} = \hat{x}$, and set

$$\Omega_N := \{\omega_N : \omega \in \Omega_0\}.$$

It is straightforward to see that the random walk in the environment ω_N has an invariant measure of the form

$$m_N := \frac{1}{(2N + 1)^d} \sum_{x \in B_\infty(N)} \Phi_N(x) \delta_{\hat{x}},$$

for some function Φ_N on $B_\infty(N)$ such that $\sum_{x \in B_\infty(N)} \Phi_N(x) = (2N + 1)^d$. Now define a probability measure on Ω_N by

$$\nu_N := \frac{1}{(2N + 1)^d} \sum_{x \in B_\infty(N)} \Phi_N(x) \delta_{t_x \omega_N}.$$

Now introduce the sequence of measures

$$\mathbb{P}_N := \frac{1}{(2N+1)^d} \sum_{x \in B_\infty(N)} \delta_{t_x \omega_N}.$$

By the multidimensional ergodic theorem (see [DF88, Theorem VIII.6.9]), we have that

$$\lim_{N \rightarrow \infty} \mathbb{P}_N = \mathbb{P} \quad \mathbb{P} - a.s.$$

Also, one can see that ν_N is absolutely continuous with respect to \mathbb{P}_N ,

$$d\nu_N =: f_N d\mathbb{P}_N,$$

with

$$\int f_N^{\frac{d}{d-1}} d\mathbb{P}_N \leq \frac{1}{(2N+1)^d} \sum_{x \in B_\infty(N)} \Phi_N(x)^{\frac{d}{d-1}}.$$

Hence, for every bounded measurable function g on Ω we have that

$$\left| \int g d\nu_N \right| \leq \left(\int |g|^d d\mathbb{P}_N \right)^{\frac{1}{d}} \left(\int f_N^{\frac{d}{d-1}} d\mathbb{P}_N \right)^{\frac{d-1}{d}} \leq \|g\|_{L^d(\mathbb{P}_N)} \|\Phi_N\|_{L^{\frac{d}{d-1}}}.$$

where we write L^d for the corresponding space with respect to the normalized counting measure on $B_\infty(N)$. Now, assume that there is a constant C such that for every N ,

$$\|\Phi_N\|_{L^{\frac{d}{d-1}}} \leq C. \tag{2.36}$$

Using the compactness of Ω and Prohorov's theorem, we can extract a subsequence ν_{N_k} of ν_N which converges weakly to some limit ν as $k \rightarrow \infty$. Then we would obtain that

$$\left| \int g d\nu \right| \leq C \|g\|_{L^d(\mathbb{P})},$$

which would prove that ν is absolutely continuous with respect to \mathbb{P} . Note also that Kozlov's theorem (Theorem 2.12) ensures that ν is deterministic. Let us now prove (2.36). For that purpose, suppose that for every function $h \in L^d(\mathbb{P}_N)$,

$$\sup_{x \in B_\infty(N), \omega_N} \left| E_{x, \omega_N} \left[\sum_{k=0}^{\infty} \left(1 - \frac{1}{N^2}\right)^k h(X_k) \right] \right| \leq CN^2 \|h\|_{L^d(\mathbb{P}_N)}. \tag{2.37}$$

We claim that (2.37) implies (2.36). Indeed,

$$\begin{aligned} \|\Phi_N\|_{L^{\frac{d}{d-1}}} &= \sup_{h: \|h\|_{L^d} \leq 1} (\Phi_N, h) = \sup_{h: \|h\|_{L^d} \leq 1} \frac{1}{N^2} \sum_{k=0}^{\infty} \left(1 - \frac{1}{N^2}\right)^k \frac{1}{(2N+1)^d} \sum_{x \in B_\infty(N)} \Phi_N(x) h(x) \\ &= \sup_{h: \|h\|_{L^d} \leq 1} \sum_{k=0}^{\infty} \frac{1}{N^2} \left(1 - \frac{1}{N^2}\right)^k \frac{1}{(2N+1)^d} \sum_{x \in B_\infty(N)} \Phi_N(x) E_{x, \omega_N} [h(X_k)], \end{aligned}$$

which would yield (2.36). We now claim that (2.37) is a consequence of the inequality

$$\|Q_\omega f\|_\infty \leq CN^2 \left(\frac{1}{(2N+1)^d} \sum_{x \in B_\infty(N)} |f(x)|^d \right)^{\frac{1}{d}}, \quad (2.38)$$

where

$$Q_\omega f(x) := E_{x, \omega_N} \left[\sum_{k=0}^{S_N-1} f(X_k) \right]$$

and

$$S_N := \inf\{n \geq 0 : |X_n|_\infty \geq N\}.$$

To prove (2.37) assuming (2.38), define $\tau_0 := 0$ and

$$\tau_1 := \tau = \inf\{n \geq 0 : |X_n - X_0|_\infty \geq N\},$$

as well as recursively for $k \geq 1$, $\tau_{k+1} := \tau \circ \theta_{\tau_k} + \tau_k$. Then, for each $\rho \in [0, 1)$ we have that

$$\begin{aligned} E_{x, \omega_N} \left[\sum_{k=0}^{\infty} \rho^k f(X_k) \right] &= E_{x, \omega_N} \left[\sum_{m=0}^{\infty} \sum_{\tau_m \leq k < \tau_{m+1}} \rho^k f(X_k) \right] \\ &\leq \sum_{m=0}^{\infty} \sup_{x \in \mathbb{Z}^d} E_{x, \omega_N} [\rho^\tau]^m \sup_{x \in \mathbb{Z}^d} |(Q_{t_x \omega}(t_x f))(0)| \\ &\leq CN^2 \frac{1}{|B_\infty(N)|^{1/d}} \|f\|_{L^d} \frac{1}{1 - \sup_x E_{x, \omega_N} [\rho^\tau]}. \end{aligned}$$

Now, for every $K > 0$ we have $E_{x, \omega_N} [\rho^\tau] \leq P_{x, \omega_N}[\tau \leq K] + \rho^K P_{x, \omega_N}[\tau > K]$. But since the random walk $(X_n)_{n \geq 0}$ is a martingale, by Doob's martingale inequality we have that for every $\lambda > 0$,

$$\lambda N P_{0, t_x \omega} \left[\sup_{0 \leq k \leq K} |X_k^i| \geq \lambda N \right] \leq C' K^{1/2},$$

for some constant $C' > 0$. Choosing $K = CN^2$ for an appropriate constant C , we have that for an appropriate choice of λ ,

$$P_{x, \omega}[\tau \leq K] \leq \sum_{i=1}^d P_{0, t_x \omega} \left[\sup_{0 \leq k \leq K} |X_k|_\infty \geq \lambda N \right] \leq C' \frac{1}{\lambda} C^{1/2} N \leq \frac{1}{2}.$$

To finish the proof, it remains to establish (2.38). As explained in Sznitman [BS02], one can follow the methods developed by Kuo and Trudinger [KT90] to obtain pointwise estimates for linear elliptic difference equations with random coefficients. One uses the fact that $u = Q_\omega f$ is a solution of the equation

$$\begin{aligned}(L_\omega u)(x) &= -f(x), & \text{for } x \in B_\infty(N), \\ u(x) &= 0, & \text{for } x \in \partial B_\infty(N);\end{aligned}$$

where

$$(L_\omega g)(x) = \sum_{e \in U} \omega(x, e)(g(x + e) - g(x)). \quad (2.39)$$

and the so-called *normal mapping* (see [KT90]) defined for $x \in B_\infty(N)$ as

$$\chi_u(x) := \{p \in \mathbb{R}^d : u(z) \leq u(x) + p \cdot (z - x), \text{ for } z \in B_\infty(N) \cup \partial B_\infty(N)\}$$

to conclude that

$$\omega_d \frac{(\max u)^d}{(2N)^d} = |B_2(\max u / (2N))| \leq \sum_{x \in B_\infty(N)} |\chi_u(x)| \leq \sum_{x \in B_\infty(N)} \frac{f(x)^d}{\kappa^d},$$

where ω_d is the volume of a sphere unit radius, which proves (2.38).

Theorems 2.24 and 2.25 have recently been extended by Guo and Zeitouni in [GZ12] to the elliptic case. Further progress has been made by Berger and Deuschel in [BD12]. They introduce the following concept which is considerably weaker than ellipticity.

Definition 2.26. (Genuinely d -dimensional environment) We say that an environment $\omega \in \Omega$ is a *genuinely d -dimensional environment* if for every $e \in U$ there exists a $y \in \mathbb{Z}^d$ such that $\omega(y, e) > 0$. We say that the law \mathbb{P} of an environment is *genuinely d -dimensional* if environments are genuinely d -dimensional under \mathbb{P} with probability one.

Theorem 2.27. ([BD12]) *Consider a RWRE in an i.i.d., balanced and genuinely d -dimensional environment. Then the quenched invariance principle holds with a deterministic non-degenerate diagonal covariance matrix.*

In [Ze04], Zeitouni proves as a corollary of Lawler's quenched central limit theorem for balanced random walks the following result.

Theorem 2.28. ([Ze04, Theorem 3.3.22]) *Under the conditions of Theorem 2.24, the random walk is transient in dimensions $d \geq 3$ and recurrent in dimension $d = 2$.*

2.8 One-dimensional quenched large deviations

The following result was first derived by Greven and den Hollander [GdH94] to the case of an i.i.d. environment and then extended by Comets, Gantert and Zeitouni [CGZ00] for ergodic environments.

Theorem 2.29. (Greven-den Hollander, Comets-Gantert-Zeitouni) Consider a RWRE in dimension $d = 1$. Assume that $\mathbb{E}[\log \rho] \leq 0$ and that the environment fulfills **(E)** and is totally ergodic. Then, there exists a deterministic rate function $I : \mathbb{R} \rightarrow [0, \infty]$ such that

(i) For every open set $G \subset \mathbb{R}$ we have that

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log P_{0, \omega} \left[\frac{X_n}{n} \in G \right] \geq - \inf_{x \in G} I(x) \quad \mathbb{P} - a.s.$$

(ii) For every closed set $C \subset \mathbb{R}$ we have that

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_{0, \omega} \left[\frac{X_n}{n} \in C \right] \leq - \inf_{x \in C} I(x) \quad \mathbb{P} - a.s.$$

Furthermore, I is continuous and convex, and it is finite exactly on $[-1, 1]$.

The strategy used by Comets, Ganter and Zeitouni in [CGZ00] to prove Theorem 2.29 is based on obtaining a recursion relation for the moment generating function $\phi(\lambda) := E_{0, \omega}[e^{\lambda T_1}]$, where for $k \geq 1$, $T_k := \inf\{n \geq 0 : X_n = k\}$, which leads to a continuous fraction expansion of it. This leads to a large deviation principle for T_k/k with rate function given by the expression

$$I(t) = \sup_{\lambda \in \mathbb{R}} (\lambda t - E_0[\phi(\lambda)]).$$

As is often the case, the expression for the rate function is much more explicit in $d = 1$ than in higher dimensions (cf. also Section 2.10 for the latter). In addition to the above, in [CGZ00] the following is also shown.

Theorem 2.30. Consider a RWRE satisfying the hypotheses of Theorem 2.29. Assume that the support of the law of $\omega(0, 1)$ intersects both $(0, \frac{1}{2}]$ and $[\frac{1}{2}, 1)$. Then the rate function I of Theorem 2.29 satisfies the following properties

(i) For $x \in (0, 1]$ we have that $I(-x) = I(x) - x\mathbb{E}[\log \rho]$.

(ii) $I(x) = 0$ if and only if $x \in [0, v]$, with v denoting the limiting velocity $\lim_{n \rightarrow \infty} X_n$ (see also (3.26) below).

Part (i) of Theorem 2.30 shows that the slope of the rate function to the left of the origin does not vanish. A similar phenomenon is expected to happen for every transient random walk fulfilling **(IID)** and **(UE)** in dimensions $d \geq 2$. This behavior is expected to be connected to the resolution of a conjecture about the equivalence of two particular ballisticity conditions (see (3.29) below), which will be discussed in Chapter 3.

2.9 Multidimensional quenched large deviations

In [Va03] Varadhan presented a short proof of the quenched large deviation principle for RWRE in general ergodic environments. His method is based on the use of the superadditive ergodic theorem.

Note that by the Markov property for each environment ω the n -step transition probability of the random walk (see (2.4)) satisfies for each natural numbers n and m and $x, y \in \mathbb{Z}^d$ the inequality

$$p^{(n+m)}(0, x + y) \geq p^{(n)}(0, x)p^{(m)}(x, x + y). \quad (2.40)$$

We would like to take logarithms on both sides to obtain a superadditive quantity and then apply the subadditive ergodic theorem. Nevertheless, there are two types of degeneracy that complicate this operation:

- (i) $p^{(0)}(x, y, \omega) = 0$ for $x \neq y$;
- (ii) $p^{(n)}(x, y, \omega) = 0$ whenever n and $|x - y|_1$ do not have the same parity.

To avoid them Varadhan introduced the following *smoothed transition probabilities*, defined for each $c > 0$ and ω, x, y and non-negative real t ,

$$q_c(x, y, t) := \sup_{m \geq 0} \{p^{(m)}(x, y, \omega) e^{-c|m-t|}\}.$$

This regularization method is related to homogenization methods already developed within the context of the stochastic Hamilton-Jacobi equation (see for example Kosygina, Rezakhanlou and Varadhan [KRV06] and Rezakhanlou [Re11]).

Theorem 2.31. (*Varadhan*) *Consider a RWRE fulfilling (UE) and (ERG). Then, there exists a convex rate function $I : \mathbb{R} \rightarrow [0, \infty]$ such that*

- (i) *For every open set $G \subset \mathbb{R}^d$ we have that*

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log P_{0, \omega} \left[\frac{X_n}{n} \in G \right] \geq - \inf_{x \in G} I(x) \quad \mathbb{P} - a.s.$$

- (ii) *For every closed set $C \subset \mathbb{R}^d$ we have that*

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_{0, \omega} \left[\frac{X_n}{n} \in C \right] \leq - \inf_{x \in C} I(x) \quad \mathbb{P} - a.s.$$

Furthermore, I is continuous in $\mathring{B}_1(1)$, lower-semicontinuous in $B_1(1)$ and $I(x) = \infty$ for $x \notin B_1(1)$.

We will present here the proof of Theorem 2.31 given by Campos, Drewitz, Rassoul-Agaha, Ramírez and Seppäläinen in [CDRRS13] and which is valid also for time-dependent random environments satisfying certain ergodicity conditions — we refer the reader to [CDRRS13] for further details on the time dependent setting.

The idea is to avoid the degeneracy issues discussed related to (ii) above, by considering the random walk at even and odd times separately.

Let us begin modifying our random walk model, admitting the possibility that the walk does not move after one step, so that the set of jumps after one step is now $U' := U \cup \{0\}$ and

$$\omega(0, 0) \geq \kappa.$$

We will call this random walk the *random walk in random environment with holding times*. We will denote by $P_{x,\omega}^h$ its quenched law starting from x and by

$$p_h^{(n)}(x, y, \omega) := P_{x,\omega}^h[X_n = y]$$

its n -step transition probabilities. For $x \in \mathbb{R}^d$, we will define

$$[x] := ([x_1], \dots, [x_d]) \in \mathbb{Z}^d.$$

Let us define for $n \geq 0$, R_n as the set of sites that the random walk can visit with positive probability at time n . Thus, $R_0 := \{0\}$, $R_1 := U'$ while for $n \geq 1$,

$$R_{n+1} := \{y \in \mathbb{Z}^d : y = x + e \text{ for some } x \in R_n \text{ and } e \in U'\} = R_{n+1} + (R_n + U).$$

It is easy to check that $B_1(1)$ equals the set of limit points of the sequence of sets R_n/n . Furthermore,

$$R_n = (nB_1(1)) \cap \mathbb{Z}^d \tag{2.41}$$

(see also Lemma 3.1 in [CDRRS13]). We will now prove the following.

Proposition 2.32. *Consider a random walk in random environment with holding times, and which fulfills (UE) and (ERG) to hold. Then, for each $x \in \mathbb{Q}^d$ we have that \mathbb{P} -a.s. the limit*

$$I(x) := - \lim_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nx]) \tag{2.42}$$

exists, is convex and deterministic. Furthermore, $I(x) < \infty$ for $x \in \mathbb{Q}^d \cap \mathring{B}_1(1)$.

PROOF. Note that from Lemma 2.41 we can check that if $x \notin B_1(1)$, for every $n \geq 1$ one has that $nx \notin nB_1(1)$ so that $nx \notin R_n$, and thus $p_h^{(n)}(0, [nx]) = 0$. This proves that $I(x) = \infty$ if $x \notin B_1(1)$.

Let us now consider an $x \in \mathbb{Q}^d \cap \mathring{B}_1(1)$. Note that there exists a $k \in \mathbb{N}$ and a $y \in \mathbb{Z}^d \cap k\mathring{B}_1(1)$ such that $x = k^{-1}y$; in addition, $y \in R_k$.

We will now introduce an auxiliary function \tilde{I} and then show that it in fact equals the expression given for I in (2.42). Indeed, by the convexity of $B_1(1)$, the subadditive ergodic theorem [Li85] and (2.40), we have that

$$\tilde{I}(k^{-1}y) := - \lim_{m \rightarrow \infty} \frac{1}{mk} \log p_h^{(mk)}(0, my)$$

exists \mathbb{P} -a.s. Furthermore, this definition is independent of the representation of x . Indeed, if $x = k^{-1}y_1 = l^{-1}y_2$ for some $k, l \in \mathbb{N}$, $y_1 \in \mathbb{Z}^d \cap k\mathring{B}_1(1)$ and $y_2 \in \mathbb{Z}^d \cap l\mathring{B}_1(1)$, we have that

$$\tilde{I}(k^{-1}y_1) = - \lim_{n \rightarrow \infty} \frac{1}{nlk} \log p_h^{(nlk)}(0, nly_1) = - \lim_{n \rightarrow \infty} \frac{1}{nlk} \log p_h^{(nlk)}(0, nky_2) = \tilde{I}(l^{-1}y_2).$$

We will next prove that \tilde{I} is deterministic on $\mathbb{Q}^d \cap \mathbb{Z}^d$. Let $x \in \mathbb{Q}^d \cap \mathring{B}_1(1)$. There exists a $k \in \mathbb{N}$ and a $y \in \mathbb{Z}^d \cap k\mathring{B}_1(1)$ such that $x = k^{-1}y$. Now it is enough to prove that for each $z \in U$ one has that

$$\tilde{I}(x, \omega) \leq \tilde{I}(x, t_z \omega) = - \lim_{m \rightarrow \infty} \frac{1}{mk} \log p_h^{(mk)}(z, my + z).$$

But for each $n \in \mathbb{N}$, we have that

$$- \frac{1}{mnk} \log p_h^{(mnk)}(0, mny) \leq - \frac{1}{mnk} \log p_h^{(mnk)}(0, z) - \frac{1}{mnk} \log p_h^{(mnk)}(z, mny).$$

By uniform ellipticity, the first term in the right-hand side of the above inequality tends to 0 as $m \rightarrow \infty$. Therefore,

$$\tilde{I}(x, \omega) = - \lim_{m \rightarrow \infty} \frac{1}{mnk} \log p_h^{(mnk)}(0, mny) \leq - \liminf_{m \rightarrow \infty} \frac{1}{mnk} \log p_h^{(mnk)}(z, mny).$$

On the other hand,

$$\begin{aligned} & - \frac{1}{mnk} \log p_h^{(mnk)}(z, mny) \\ & \leq - \frac{1}{mnk} \log p_h^{((m-1)nk)}(z, (m-1)ny + z) - \frac{1}{mnk} \log p_h^{(nk-1)}((m-1)ny + z, mny). \end{aligned}$$

Now, since $z \in U$, one can check that $p_h^{(nk-1)}((m-1)ny + z, mny) \geq \kappa^{nk-1}$, so that the last term of the above inequality tends to 0 when $m \rightarrow \infty$. We can then conclude that $\tilde{I}(x, \omega) \leq \tilde{I}(x, t_z \omega)$.

We will now prove that I is well defined in $\mathbb{Q}^d \cap \mathring{B}_1(1)$ and that it equals \tilde{I} there. Let $x \in \mathbb{Q}^d \cap \mathring{B}_1(1)$. Furthermore, choose k such that $kx \in \mathbb{Z}^d$ and given $n \in \mathbb{N}$ define

$$m := \left[\frac{n}{k} \right].$$

Necessarily, we can find a sequence $z_1, \dots, z_{n-mk} \in U$ such that

$$[nx] = mkx + z_1 + \dots + z_{n-mk}.$$

Hence, by superadditivity and uniform ellipticity we have that

$$- \frac{1}{n} \log p_h^{(n)}(0, [nx]) \leq - \frac{1}{n} \log p_h^{(mk)}(0, mkx) - \frac{1}{n} \log \kappa^{n-mk}.$$

Therefore

$$-\limsup_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nx]) \leq \tilde{I}(x).$$

Using a similar argument we can establish that

$$-\liminf_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nx]) \geq \tilde{I}(x).$$

□

We want now to extend Proposition 2.32 to $x \in \mathbb{R}^d$. To do this, we will need to establish a lemma which in some sense shows that the quantity $-\log p^{(n)}(0, [nx])$ is continuous as a function of x . For each $x \in \mathbb{Z}^d$ we define $s(x)$ as the minimum number n of steps required for the random walk to move from 0 to x . so that

$$s(x) := \min\{n \geq 0 : x \in R_n\}.$$

We will now define a norm in \mathbb{R}^d as follows. For each $y \in \partial B_1(1)$ we set $\|y\| := 1$. Then, for each $x \in \mathbb{R}^d$ of the form $x = ay$ for some $a \geq 0$, we define $\|x\| := a$. Since $B_1(1)$ is convex, symmetric (in the sense that $x \in B_1(1)$ implies that $-x \in B_1(1)$), this implies that this defines a norm. It is easy to check that for every $x \in \mathbb{Z}^d$,

$$\|x\| \leq s(x) \leq \|x\| + 1. \quad (2.43)$$

Lemma 2.33. *Let $z \in B_1(1)$ and $x \in \overset{\circ}{B}_1(1)$.*

(i) *For each natural n there exists an n_2 such that*

$$n \leq n_2 \leq n + \frac{4d+1}{1-\|x\|} + n \frac{\|x-z\|}{1-\|x\|} + 1. \quad (2.44)$$

and such that

$$-\log p_h^{(n_2)}(0, [n_2x]) \leq -\log p_h^{(n)}(0, [nz]) - \log \kappa^{n_2-n}.$$

(ii) *Similarly, whenever $\|x-z\| < 1 - \|x\|$, there exists an n_0 such that for each natural $n \geq n_0$ there exists an n_1 such that*

$$n - \frac{4d+1}{1-\|x\|} - n \frac{\|x-z\|}{1-\|x\|} - 1 \leq n_1 \leq n \quad (2.45)$$

and such that

$$-\log p_h^{(n)}(0, [nz]) \leq -\log p_h^{(n_1)}(0, [n_1x]) - \log \kappa^{n-n_1}.$$

PROOF. To prove part (i) of the lemma, it is enough to show that there exists an $n_2 \geq n$ satisfying (2.44) and such that

$$s([n_2x] - [nz]) \leq n_2 - n. \quad (2.46)$$

But by (2.43) and the fact that $\|x - [x]\| \leq d$ we see that

$$\begin{aligned} s([n_2x] - [nz]) &\leq \|[n_2x] - [nz]\| + 1 \leq \|[n_2x] - [nx]\| + \|[nx] - [nz]\| + 1 \\ &\leq \|(n_2 - n)x\| + \|n(x - z)\| + 4d + 1 = (n_2 - n)\|x\| + n\|x - z\| + 4d + 1. \end{aligned}$$

This shows that (2.46) is satisfied whenever

$$n_2 \geq n + \frac{4d + 1}{1 - \|x\|} + n \frac{\|x - z\|}{1 - \|x\|}.$$

To prove part (ii) of the lemma, note that it is enough to show that there exists an $n_1 \leq n$ satisfying (2.45) and

$$s([nz] - [n_1x]) \leq n - n_1.$$

But,

$$s([nz] - [n_1x]) \leq n\|z - x\| + (n - n_1)\|x\| + 4d + 1$$

which is equivalent to

$$n_1 \leq n - \frac{4d + 1}{1 - \|x\|} - n \frac{\|z - x\|}{1 - \|x\|}.$$

□

We are now in a position to extend Proposition 2.32 to the following.

Proposition 2.34. *Consider a random walk in random environment with holding times, where the law \mathbb{P} of the environment is totally ergodic. Then, for each $x \in \mathbb{R}^d$ we have that \mathbb{P} -a.s. the limit*

$$I(x) := - \lim_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nx])$$

exists, is convex and deterministic. Furthermore, $I(x) < \infty$ if and only if $x \in B_1(1)$.

PROOF. Let $z \in \mathbb{R}^d \cap \overset{\circ}{B}_1(1)$. Choose a point x with rational coordinates such that $\|z - x\| < 1 - \|x\|$ and $\frac{1}{1 - \|x\|} \leq 2 \frac{1}{1 - \|z\|}$. By Lemma 2.33, for each $n \geq n_0$ we can find n_1 and n_2 satisfying (2.44) and (2.45) and such that

$$-\frac{n_2}{n} \frac{1}{n_2} \log p_h^{(n_2)}(0, [n_2x]) \leq -\frac{1}{n} \log p_h^{(n)}(0, [nz]) + b \left(\frac{n_2}{n} - 1 \right)$$

and

$$-\frac{1}{n} \log p_h^{(n)}(0, [nz]) \leq -\frac{n_1}{n} \frac{1}{n_1} \log p_h^{(n_1)}(0, [n_1x]) + b \left(1 - \frac{n_1}{n}\right),$$

where $b := -\log \kappa$. From inequalities (2.44) and (2.45) of Lemma 2.33 and by Proposition 2.32 we can then conclude that

$$I(x) \leq -\liminf_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nz]) + C(z)b\|x - z\|$$

and

$$-\liminf_{n \rightarrow \infty} \frac{1}{n} \log p_h^{(n)}(0, [nz]) \leq I(x) + C(z)b\|x - z\|,$$

where $C(z) := 2\frac{1}{1-\|z\|}$. Letting $x \rightarrow z$ we conclude that I is well defined on $\mathbb{R}^d \cap \mathring{B}_1(1)$. \square

We are now in a position to extend the function I of Proposition 2.34 from $\mathring{B}_1(1)$ to $B_1(1)$ as

$$I(x) := \begin{cases} I(x), & \text{if } x \in \mathring{B}_1(1), \\ \liminf_{\mathring{B}_1(1) \ni y \rightarrow x} I(y), & \text{if } x \in \partial B_1(1). \end{cases}$$

We will show that this is in fact the rate function of Theorem 2.31, but of a RWRE with holding times. Let us first show that I satisfies the requirements of Theorem 2.31. By uniform ellipticity, it is clear that $I(x) \leq |\log \kappa|$ whenever $x \in B_1(1)$. Also, the proof of Proposition 2.34 shows that I is continuous in $\mathring{B}_1(1)$. Furthermore, it is obvious that I is convex and lower-semicontinuous in $B_1(1)$.

Now, note that if G is an open subset of \mathbb{R}^d and $x \in G$, the sequence $[nx]$ is in $nG \cap \mathbb{Z}^d$ and

$$P_{0,\omega}^h \left[\frac{X_n}{n} \in G \right] \geq P_{0,\omega}^h [X_n = [nx]].$$

In combination with Proposition 2.34 we therefore conclude that

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \log P_{0,\omega}^h \left[\frac{X_n}{n} \in G \right] \geq -\inf_{x \in G} I(x).$$

Let us now consider a compact set $C \subset \mathring{B}_1(1)$. We then have that

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_{0,\omega}^h \left[\frac{X_n}{n} \in C \right] \leq \limsup_{n \rightarrow \infty} \sup_{x \in C} \frac{1}{n} \log p_h^{(n)}(0, [nx]) = \inf_n \sup_{x \in C} \sup_{m \geq n} \frac{1}{m} \log p_h^{(n)}(0, [mx]).$$

Now, through a contradiction argument and an application of Lemma 2.33, one can prove that

$$\sup_{x \in C} \sup_{m \geq n} \frac{1}{m} \log p_h^{(n)}(0, [mx]) \leq -\inf_{x \in C} I(x).$$

This shows that

$$\limsup_{n \rightarrow \infty} \frac{1}{n} \log P_{0,\omega}^h \left[\frac{X_n}{n} \in C \right] \leq - \inf_{x \in C} I(x). \quad (2.47)$$

Standard arguments using uniform ellipticity enable us now to extend (2.47) from compact sets to closed sets.

One can now derive Theorem 2.31 for the plain RWRE from the RWRE with holding times as follows. Define the *even lattice* as $\mathbb{Z}_{\text{even}}^d := \{x \in \mathbb{Z}^d : |x|_1 \text{ is even}\}$. Using the fact that since $\mathbb{Z}_{\text{even}}^d$ is a free Abelian group it is isomorphic to \mathbb{Z}^d , we can apply Proposition 2.34 for the RWRE with holding times to deduce an analogous result for the random walk $Y_n := X_{2n}$ at even times. On the other hand, using the equality

$$P_{0,\omega} \left[\frac{X_{2n+1}}{2n+1} \in A \right] = \sum_{i=1}^{2d} \omega(0, e_i) P_{e_i, \omega} \left[\frac{X_{2n}}{2n} \in A \right],$$

and the asymptotic behavior previously proved at even times, in combination with the assumption of uniform ellipticity, we can deduce the large deviation principle of Theorem 2.31.

2.10 Rosenbluth's variational formula for the multidimensional quenched rate function

The drawback of Theorem 2.31 is that it gives very little information about the rate function of the quenched large deviations of the random walk. A partial remedy to this was obtained by Rosenbluth [Ro06] in his Ph.D. thesis in 2006, where he derived a variational expression for the rate function. To state Rosenbluth's result, it is more natural to define the RWRE in an abstract setting, where we first define the dynamics of the environmental process. In analogy to the set of admissible transition kernels \mathcal{P} defined in (2.3), we denote by \mathcal{Q} the set of measurable functions $q : \Omega \times U \mapsto [0, 1]$ such that $\sum_{e \in U} q(\omega, e) = 1$ for all $\omega \in \Omega$. Define the function $p \in \mathcal{Q}$ via $p(\omega, e) := \omega(0, e)$, corresponding to the transition probabilities of the canonical RWRE. Let us call \mathcal{D} the set of measurable functions $\phi : \Omega \rightarrow [0, \infty)$ such that $\int \phi d\mathbb{P} = 1$.

Theorem 2.35. *Assume that (ERG) is fulfilled and that there is an $\alpha > 0$ such that*

$$\max_{e \in U} \int |\ln p(\omega, e)|^{d+\alpha} \mathbb{P}(d\omega) < \infty.$$

Then the RWRE satisfies a large deviation principle with rate function

$$I(x) := \sup_{\lambda \in \mathbb{R}^d} \{\lambda \cdot x - \Lambda(\lambda)\},$$

where

$$\Lambda(\lambda) := \sup_{q \in \mathcal{Q}} \sup_{\phi \in \mathcal{D}} \inf_h \sum_{e \in U} \int \left(\lambda \cdot e - \ln \frac{q(\omega, e)}{p(\omega, e)} + h(\omega) - h(T_e \omega) \right) q(\omega, e) \phi(\omega) \mathbb{P}(d\omega).$$

Remark 2.36. *The integrability assumption in the above theorem is fulfilled if **(UE)** holds true, for example.*

Note that using canonical LDP machinery, one can show that it is enough to prove that

$$\lim_{n \rightarrow \infty} \log E_{P_\omega} [e^{\lambda \cdot X_n}] = \Lambda(\lambda).$$

We will just give an idea of the proof of the above theorem deriving the lower bound in the above limit. In analogy to the definition of P_ω in (2.17), given $q \in \mathcal{Q}$, we denote by Q_ω the law of the corresponding Markov chain $(\bar{\omega}_n)_{n \geq 0}$ starting from ω . We then have

$$E_{P_\omega} [e^{\lambda \cdot X_n}] = E_{Q_\omega} \left[e^{\lambda \cdot X_n} \frac{dP_\omega}{dQ_\omega} \right] = E_{Q_\omega} \left[\exp \left\{ \lambda \cdot X_n - \sum_{k=0}^{n-1} \ln \frac{q(t_{X_k} \omega, X_{k+1} - X_k)}{p(t_{X_k} \omega, X_{k+1} - X_k)} \right\} \right].$$

By Jensen's inequality it follows that

$$\liminf_{n \rightarrow \infty} \frac{1}{n} \ln E_{P_\omega} [e^{\lambda \cdot X_n}] \geq \lim_{n \rightarrow \infty} E_{Q_\omega} \left[\frac{1}{n} \lambda \cdot X_n - \frac{1}{n} \sum_{k=0}^{n-1} \sum_{e \in U} \ln \frac{q(t_{X_k} \omega, e)}{p(t_{X_k} \omega, e)} \right]. \quad (2.48)$$

Now note that the expectation of the second term of (2.48) can be written as

$$E_{Q_\omega} \left[\frac{1}{n} \sum_{k=0}^{n-1} \sum_{e \in U} \ln \frac{q(t_{X_k} \omega, e)}{p(t_{X_k} \omega, e)} \right] = \frac{1}{n} \sum_{k=0}^{n-1} \sum_{e \in U} \ln \frac{q(t_{X_k} \omega, e)}{p(t_{X_k} \omega, e)} q(t_{X_k} \omega, e).$$

Let us now assume that the chain $(\bar{\omega}_n)_{n \geq 0}$ under Q_ω has an invariant measure ν which is absolutely continuous with respect to \mathbb{P} . Let us call ϕ the Radon-Nikodym derivative of ν with respect to \mathbb{P} . By Kozlov's theorem (Theorem 2.12), we know that the measure ν is such that $Q_\nu := \int Q_\omega \nu(d\omega)$ is ergodic (with respect to the time shifts). It follows that

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=0}^{n-1} \sum_{e \in U} \ln \frac{q(t_{X_k} \omega, e)}{p(t_{X_k} \omega, e)} q(t_{X_k} \omega, e) = \int \sum_{e \in U} \ln \frac{q(\omega, e)}{p(\omega, e)} q(\omega, e) \phi(\omega) \mathbb{P}(d\omega) \quad Q_\nu - a.a. \omega,$$

and hence that

$$\lim_{n \rightarrow \infty} E_{Q_\omega} \left[\frac{1}{n} \sum_{k=0}^{n-1} \sum_{e \in U} \ln \frac{q(t_{X_k} \omega, e)}{p(t_{X_k} \omega, e)} q(t_{X_k} \omega, e) \right] = \int \sum_{e \in U} \ln \frac{q(\omega, e)}{p(\omega, e)} q(\omega, e) \phi(\omega) \mathbb{P}(d\omega) \quad \mathbb{P} - a.a. \omega.$$

On the other hand, by the law of large numbers, we have that the behavior of the first term on the right-hand side of (2.48) is characterized by

$$\lim_{n \rightarrow \infty} E_{Q_\omega} \left[\frac{1}{n} \lambda \cdot X_n \right] = \int \sum_{e \in U} \lambda \cdot e q(\omega, e) \phi(\omega) \mathbb{P}(d\omega) \quad \mathbb{P} - a.a. \omega.$$

It follows that if we call \mathcal{Q}_0 the set of transition probabilities q for which there is an invariant measure ν_q which is absolutely continuous with respect to \mathbb{P} (and which is unique, by part (iii) of Kozlov's theorem), with $\phi_q = \frac{d\nu_q}{d\mathbb{P}}$ we have by (2.48) that

$$\Lambda(\lambda) \geq \sup_{q \in \mathcal{Q}_0} \sum_{e \in U} \int \left((\lambda, e) - \ln \frac{q(\omega, e)}{p(\omega, e)} \right) q(\omega, e) \phi_q(\omega) \mathbb{P}(d\omega).$$

Now note that for $\phi \in \mathcal{D}$, the following are equivalent

$$\phi = \phi_q$$

and

$$\inf_h \int \sum_{e \in U} (h(\omega) - h(T_e \omega)) q(\omega, e) \phi(\omega) \mathbb{P}(d\omega) = 0.$$

Similarly,

$$\phi \neq \phi_q$$

and

$$\inf_h \int \sum_{e \in U} (h(\omega) - h(T_e \omega)) q(\omega, e) \phi(\omega) \mathbb{P}(d\omega) = -\infty.$$

Therefore, we conclude that

$$\begin{aligned} & \sup_{q \in \mathcal{Q}_0} \sum_{e \in U} \int \left(\lambda \cdot e - \ln \frac{q(\omega, e)}{p(\omega, e)} \right) q(\omega, e) \phi_q(\omega) \mathbb{P}(d\omega) \\ &= \sup_{q \in \mathcal{Q}, \phi \in \mathcal{D}} \inf_h \sum_{e \in U} \int \left(\lambda \cdot e - \ln \frac{q(\omega, e)}{p(\omega, e)} + h(\omega) - h(T_e \omega) \right) q(\omega, e) \phi_q(\omega) \mathbb{P}(d\omega), \end{aligned}$$

which finishes the sketch of the proof for the lower bound.

A level 2 large deviation principle version of Rosenbluth's variational formula was derived by Yilmaz in [Yi09]. Subsequently, a level 3 version was derived by Rassoul-Agha and Seppäläinen in [RAS11].

Chapter 3

Trapping and ballistic behavior in higher dimensions

In Chapter 2 we have already considered some situations in which one has been able to obtain information not only on transience and ballisticity, but also on the diffusive behavior of RWRE as well as its large deviations; in these situations, this supplied us with a rather precise understanding of the asymptotic behavior. The content of this chapter is a more general analysis of RWRE in terms of the coarser scales of (directional) transience and ballistic behavior.

3.1 Directional transience

As we have seen in Chapter 2, the question of whether under appropriate conditions a RWRE in dimension $d \geq 3$ is transient, remains essentially unsolved. More is known, however, about “transience in a given direction” which has been introduced in Definition 2.13, and we will see how this concept plays a role in the investigation of ballistic behavior of RWRE also. In fact, some quite challenging questions concerning RWRE are related to that notion, too, as we will see in this chapter.

In the following, we will tacitly use for $x \in \mathbb{Z}^d$ the equivalence of the conditions

$$\begin{aligned} & “P_x[A_l] = 1”, \\ & \text{and} \\ & “\text{for } \mathbb{P}\text{-almost all } \omega \text{ one has } P_{x,\omega}[A_l] = 1”. \end{aligned} \tag{3.1}$$

Note that this equivalence is a direct consequence of the definition of the averaged measure below (2.5).

The following result has essentially been proven by Kalikow [Ka81] and has been refined in [SZ99, ZM01].

Lemma 3.1. *Consider a RWRE satisfying **(E)** and **(IID)**. Then for every $l \in \mathbb{S}^{d-1}$ we have that*

$$P_0[A_l \cup A_{-l}] \in \{0, 1\}.$$

Of course, the above zero-one law seems incomplete and one would like to have a zero-one law for the event A_l already. Intriguingly, however, it is still not known if such a statement holds in full generality.

Open question 3.2. *Consider a RWRE satisfying the assumptions **(E)** and **(IID)**. Is it true that for every $l \in \mathbb{S}^{d-1}$ one has*

$$P_0[A_l] \in \{0, 1\}? \tag{3.2}$$

As we have seen in Theorem 2.10, statement 3.2 holds true for $d = 1$. In dimension two, it has been proven to hold true by Zerner and Merkl [ZM01]. In fact, it is also shown in that source that if one assumes the environment to be stationary and ergodic with respect to lattice translations only, it can indeed happen that $P_0[A_l] \notin \{0, 1\}$.

Apart from leading to interesting problems on its own, the events A_l also play a key role in the next section in order to define a renewal structure for RWRE.

3.2 Renewal structure

In order to prove some of the main asymptotic results for RWRE in the directionally transient regime, we will define a renewal structure which will help us to decompose the RWRE in terms of finite i.i.d. (apart from its initial part; see Corollary 3.6) trajectories. The first use of this renewal structure in the context of RWRE is due to Kesten, Kozlov and Spitzer [KKS75] in the one-dimensional case, and it has then been generalized to the higher-dimensional case by Sznitman and Zerner [SZ99]. It can be introduced as follows: given a direction $l \in \mathbb{S}^{d-1}$, it is the first time that the random walk reaches a new maximum level in direction l and such that after this time it never goes below this maximum in direction l . Thus, an easy way to define the renewal time τ_1 is via

$$\tau_1 := \min \left\{ n \geq 1 : \max_{0 \leq m \leq n-1} X_m \cdot l < X_n \cdot l \leq \inf_{m \geq n} X_m \cdot l \right\}. \tag{3.3}$$

Another way to put it is that τ_1 is the first time that the last exit time from a half space of the form $\{x \cdot l < r\}$, some $r \in \mathbb{R}$, coincides with the first entrance time into its complement.

In order to introduce notation which is used in the computations below, we give another definition of τ_1 in terms of a sequence of stopping times; it is slightly more involved. Consider

$$H_u^l := \inf \{ n \geq 1 : X_n \cdot l > u \}$$

for $u \in \mathbb{R}$ as well as

$$D := \inf \{ n \geq 0 : X_n \cdot l < X_0 \cdot l \}$$

which are stopping times with respect to the canonical filtration. Furthermore, set

$$S_0 := 0, \quad R_0 := X_0 \cdot l.$$

In a slight abuse of notation and similarly to (2.29), we will now use θ to denote the canonical shift on $(\mathbb{Z}^d)^\mathbb{N}$, i.e.,

$$\theta : (x_0, x_1, x_2, \dots) \mapsto (x_1, x_2, x_3, \dots),$$

and for $n \geq 1$, we define θ_n to be the n -fold composition of θ . Using this notation, for $k \geq 1$ we now introduce the stopping times

$$S_k := H_{R_{k-1}}^l, \quad D_k := \begin{cases} D \circ \theta_{S_k} + S_k, & \text{if } S_k < \infty, \\ \infty, & \text{otherwise,} \end{cases} \quad (3.4)$$

$$R_k := \sup\{X_m \cdot l : 0 \leq m \leq D_k\}.$$

We then define

$$K := \inf\{k \geq 0 : S_k < \infty, D_k = \infty\}, \quad (3.5)$$

and the first *renewal time*,

$$\tau_1 := S_K.$$

Note that τ_1 is not a stopping time with respect to the canonical filtration anymore, since in order to determine whether $\{S_K = m\}$ occurs one has to “see into the future” of (X_n) after time m . One can then recursively define the sequence of regeneration times $(\tau_k)_{k \in \mathbb{N}}$ via

$$\tau_{k+1} = \tau_1 \circ \theta_{\tau_k} + \tau_k, \quad k \geq 1,$$

and set $\tau_0 = 0$. See Figure 3.1 for an illustration of the above renewal structure.

Remark 3.3. • *Note here that, although not emphasized explicitly in the notation, the definition of the sequence (τ_n) depends on the choice of the direction l ; if the very choice of l matters, it will usually be clear from the context.*

- *If working with directions l having rational coordinates, Definition 3.3 works fine. However, for general directions $l \in \mathbb{S}^{d-1}$, one might under some circumstances run into slightly more technical argumentations — e.g., for guaranteeing that each time a renewal time occurs, the walker has gained some height bounded away from 0 in direction l (see for example [Sz00, (1.63)]); however, these complications do not pose any serious problems.*

Note, on the one hand, that one way to avoid this kind of technicalities is to replace $H_{R_{k-1}}^l$ in (3.4) by $H_{R_{k-1}+a}^l$ for some $a > 0$, as is done for example in [SZ99]. On the other hand, however, formulas such as in Lemma 3.10 would result to be more complicated, and therefore we stick to the definition given above.

The following lemma illustrates the role of the events A_l from (2.27) in the definition of the renewal structure described above.

Lemma 3.4. *Assume (E) and (IID) to hold. Let furthermore $l \in \mathbb{S}^{d-1}$ and assume that*

$$P_0[A_l] > 0. \quad (3.6)$$

Then the following are satisfied:

$$(i) \quad P_0[D = \infty] > 0; \quad (3.7)$$

$$(ii) \quad P_0[A_l \Delta \{K < \infty\}] = 0.$$

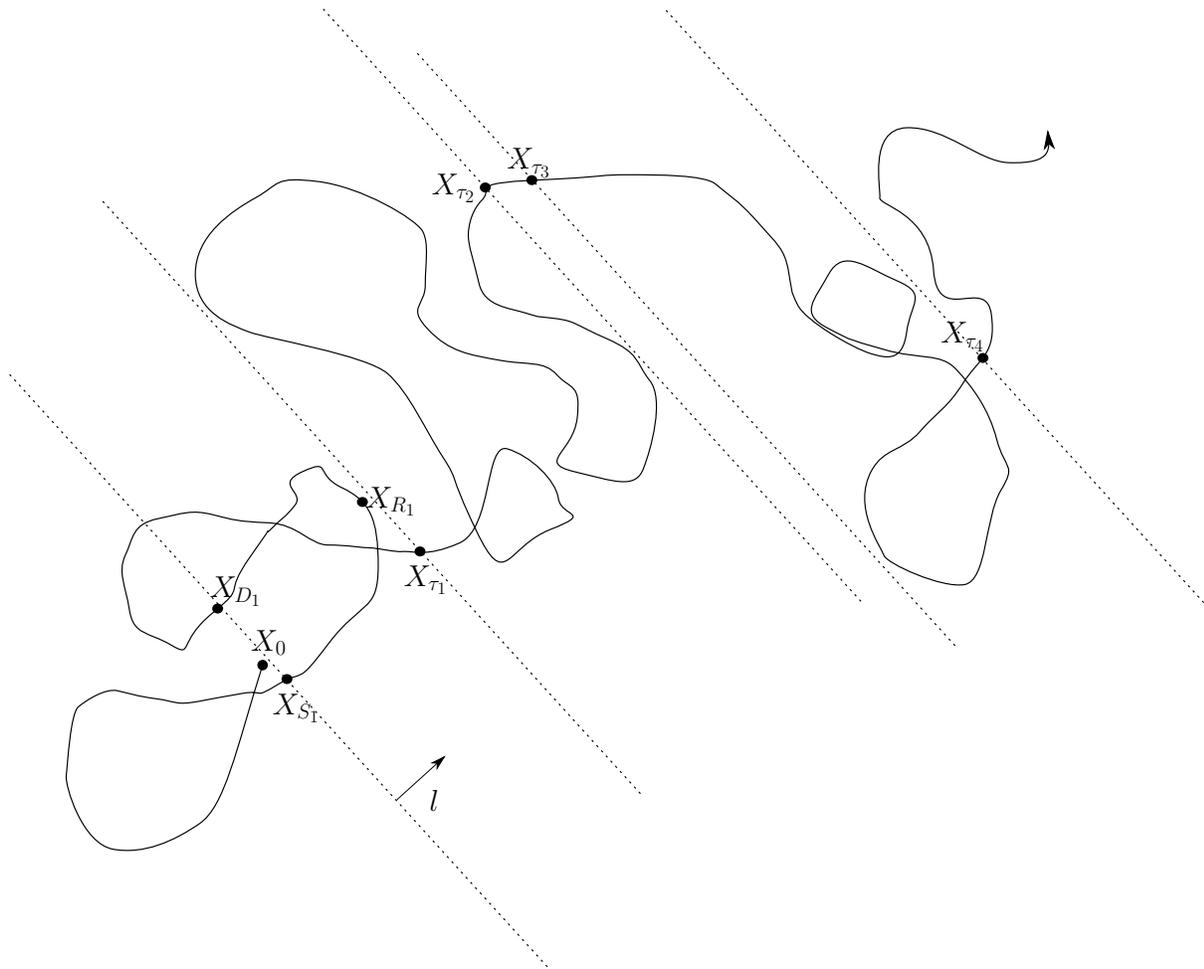


Figure 3.1: *Sketch of the renewal structure*

In words, Lemma 3.4 (i) states that if the walk has a positive probability of finally escaping to infinity in direction l , then it must have a positive probability of doing so “at once”, i.e., without entering the half-space $\{x \in \mathbb{Z}^d : x \cdot l < 0\}$. Part (ii) then ensures that on A_l , the above renewal structure is a.s. well-defined.

PROOF. Let us first prove part (i). Let (3.6) be fulfilled and assume that

$$P_0[D = \infty] = 0, \quad \text{i.e.,} \quad P_0[D < \infty] = 1.$$

From the invariance of \mathbb{P} under spatial translations, it follows that for all $x \in \mathbb{Z}^d$ we have

$$P_x[D < \infty] = 1.$$

Using (3.1), we deduce that for \mathbb{P} -almost all ω we would get that for all $x \in \mathbb{Z}^d$,

$$P_{x,\omega}[D < \infty] = 1.$$

Therefore, iteratively applying the strong Markov property at the return times of the walk to the half-space $\{x \in \mathbb{Z}^d : x \cdot l \leq 0\}$, we obtain that P_0 -a.s.,

$$\liminf_{n \rightarrow \infty} X_n \cdot l \leq 0,$$

which is a contradiction to (3.6).

We now prove part (ii). Recalling the definition of K from (3.5), we note that

$$\{K < \infty\} \subset A_{-l}^c.$$

In combination with the zero-one law of Lemma 3.1, we therefore infer that

$$P_0[\{K < \infty\} \setminus A_l] = 0. \quad (3.8)$$

On the other hand, observe that for $k \geq 1$,

$$\begin{aligned} P_0[R_k < \infty] &= P_0[S_k < \infty, R_k < \infty] = \mathbb{E}[E_{0,\omega}[S_k < \infty, P_{X_{S_k},\omega}[D < \infty]]] \\ &= \sum_{x \in \mathbb{Z}^d} \mathbb{E}[P_{0,\omega}[S_k < \infty, X_{S_k} = x] P_{x,\omega}[D < \infty]] \\ &= \sum_{x \in \mathbb{Z}^d} P_0[S_k < \infty, X_{S_k} = x] P_0[D < \infty] \\ &= P_0[S_k < \infty] P_0[D < \infty] \leq P_0[S_{k-1} < \infty, R_{k-1} < \infty] P_0[D < \infty], \end{aligned}$$

where to obtain the penultimate equality we used assumption **(IID)** in combination with the fact that $P_{0,\omega}[S_k < \infty, X_{S_k} = x]$ and $P_{x,\omega}[D < \infty]$ are measurable with respect to a disjoint set of coordinates in Ω .

It follows that

$$P_0[R_k < \infty] \leq P_0[D < \infty]^k.$$

Using part (i) of this lemma, this again implies $P_0[K < \infty \mid A_l] = 1$, which again yields

$$P_0[A_l \setminus \{K < \infty\}] = 0$$

and hence in combination with (3.8) finishes the proof. □

The next result is contained in [SZ99, Prop. 1.4]

Proposition 3.5. *Denote*

$$\mathcal{G}_1 := \sigma(\tau_1, (X_k)_{0 \leq k \leq \tau_1}, (\omega(y, \cdot))_{\{y : y \cdot l < X_{\tau_1} \cdot l\}}).$$

Then the joint distribution of

$$((X_n - X_{\tau_1})_{n \geq \tau_1}, (\omega(y, \cdot))_{y \cdot l \geq X_{\tau_1} \cdot l})$$

under $P_0[\cdot \mid A_l, \mathcal{G}_1]$ equals the joint distribution of

$$((X_n)_{n \geq 0}, (\omega(y, \cdot))_{y \cdot l \geq 0})$$

under $P_0[\cdot \mid D = \infty]$.

In particular, one can infer inductively that on A_l , the sequence of renewal times (τ_n) is well-defined.

As a corollary of a slight generalization of the above result, Sznitman and Zerner [SZ99] obtain the following.

Corollary 3.6. *Under $P_0[\cdot | A_l]$, the variables $(X_{\tau_k} - X_{\tau_{k-1}}, \tau_k)_{k \geq 1}$ are an independent family. Furthermore, $(X_{\tau_k} - X_{\tau_{k-1}}, \tau_k)_{k \geq 2}$, under $P_0[\cdot | A_l]$ are identically distributed as $(X_{\tau_1} - X_0, \tau_1)$ under $P_0[\cdot | A_l]$.*

On an intuitive level, the idea behind the proof of Corollary 3.6 is that the environments that the walk sees between different renewal times are i.i.d., which can then be transferred to the behavior of the walk itself.

3.3 A general law of large numbers

Recall that we have already seen a law of large numbers in Corollary 2.15; however, the assumptions for that result included the existence of an invariant measure ν for the environmental process such that ν was absolutely continuous with respect to \mathbb{P} . We have seen that in some special cases (cf. e.g. Theorem 2.24), one can ensure the existence of such a measure ν . On the other hand, however, not much is known about when such ν exists, and it would be desirable to have a law of large numbers that holds without this assumption.

The following theorem is such a result and constitutes a slight refinement of the directional laws of large numbers by Zerner [Ze02, Theorem 1] and Zeitouni [Ze04, Theorem 3.2.2].

Theorem 3.7. *Assume (IID) and (UE) to hold. Then in dimensions $d \geq 2$, there exists a direction $\nu \in \mathbb{S}^{d-1}$, and $v_1, v_2 \in [0, 1]$ (all deterministic) such that P_0 -a.s.*

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = v_1 \nu 1_{A_\nu} - v_2 \nu 1_{A_{-\nu}}. \quad (3.9)$$

Remark 3.8. *Let us remark here that on the level of the law of large numbers (in contrast to the central limit theorem or large deviation results), the averaged result directly implies the \mathbb{P} -a.s. quenched result due to (3.1).*

Since the conjectured zero-one law of open question 3.2 is still eluding its complete resolution, the right-hand side of (3.9) might be a non-degenerate random variable. In dimensions larger or equal to five, Berger [Be08] has shown that at least one of the velocities v_1 and v_2 must vanish. In dimension two, the zero-one law of Zerner and Merkl [ZM01] mentioned after open question 3.2 leads to the following corollary of Theorem 3.7.

Corollary 3.9. *Assume (IID) and (UE) to hold. Then in dimension $d = 2$, there exists a direction $\nu \in \mathbb{S}^{d-1}$, and $v_1 \in [0, 1]$ (all deterministic) such that P_0 -a.s.*

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = v_1 \nu.$$

To prove Theorem 3.7, we need the following lemma.

Lemma 3.10. *Assume (IID) and (E) to be fulfilled. Then for $l = (l_1, \dots, l_d) \in \mathbb{Z}^d$ such that $\gcd(l_1, \dots, l_d) = 1$, one has*

$$E_0[X_{\tau_1} \cdot l \mid D = \infty] = \frac{1}{P_0[D = \infty \mid A_l] \lim_{i \rightarrow \infty} P_0[H_{i-1}^l < \infty, X_{H_i^l} \cdot l = i]} < \infty. \quad (3.10)$$

(Note that in a slight abuse of notation we use $l \in \mathbb{Z}^d$ instead of $l \in \mathbb{S}^{d-1}$ here.)

In the case $l = (1, 0, \dots, 0)$, the proof of Lemma 3.10 can be found in [TZ04, Lemma 3.2.5] and is based on an argument by Zerner. See [DR10, Lemma 2.5] for how (in the context of a different renewal structure) the generalization to l as in Lemma 3.10 works and how to obtain the finiteness of (3.10).

Proof of Theorem 3.7. The proof is split into several pieces.

- (i) We start with proving the following version of a directional law of large numbers, which can be found in [TZ04, Theorem 3.2.2]. It states that for $l \in \mathbb{S}^{d-1}$ with

$$P_0[A_l \cup A_{-l}] = 1 \quad (3.11)$$

there exist $v_l, v_{-l} \in [0, 1]$ such that P_0 -a.s.

$$\lim_{n \rightarrow \infty} \frac{X_n \cdot l}{n} = v_l 1_{A_l} - v_{-l} 1_{A_{-l}}. \quad (3.12)$$

We will prove this result here for $l \in \mathbb{Z}^d$, which is slightly easier notationwise. Without loss of generality, assume that $P_0[A_l] > 0$. Then, by the standard law of large numbers in combination with Corollary 3.6, $P_0[\cdot \mid A_l]$ -a.s. we have that

$$\lim_{k \rightarrow \infty} \frac{\tau_k}{k} = E_0[\tau_1 \mid D = \infty],$$

and

$$\lim_{k \rightarrow \infty} \frac{X_{\tau_k} \cdot l}{k} = E_0[X_{\tau_1} \cdot l \mid D = \infty].$$

From this we conclude that $P_0[\cdot \mid A_l]$ -a.s.

$$\lim_{k \rightarrow \infty} \frac{X_{\tau_k} \cdot l}{\tau_k} = \frac{E_0[X_{\tau_1} \cdot l \mid D = \infty]}{E_0[\tau_1 \mid D = \infty]} =: v_l, \quad (3.13)$$

which due to Lemma 3.10 is a finite quantity. Using the fact that the τ_k and $X_{\tau_k} \cdot l$ are increasing in k , one obtains the sandwiching

$$\frac{X_{\tau_k} \cdot l}{\tau_{k+1}} \leq \frac{X_n \cdot l}{n} \leq \frac{X_{\tau_{k+1}} \cdot l}{\tau_k}$$

for $\tau_k \leq n < \tau_{k+1}$. In combination with (3.13) we infer that

$$\lim_{n \rightarrow \infty} \frac{X_n \cdot l}{n} = v_l$$

$P_0[\cdot \mid A_l]$ -a.s. By exchanging l for $-l$ in the above, in combination with (3.11) we therefore obtain (3.12).

- (ii) Next, we will use [Ze02, Theorem 1] which states that assuming **(IID)**, **(E)** and $P_0[A_e \cup A_{-e}] = 0$, one has for any $e \in U$, that

$$\lim_{n \rightarrow \infty} \frac{X_n \cdot e}{n} = 0, \quad P_0 - a.s. \quad (3.14)$$

On a very coarse heuristic level, the proof of that result is as follows by contradiction: Let

$$P_0[A_l \cup A_{-l}] = 0, \quad (3.15)$$

and assume that

$$\limsup_{n \rightarrow \infty} \frac{X_n \cdot l}{n} > 0$$

with positive probability. Then, if one partitions \mathbb{Z}^d into slabs orthogonal to l which are of positive finite thickness, there exists a constant C such that with positive probability, the walk visits each of a positive fraction of the slabs for at most C time steps. One can next deduce that, denoting the first entrance position of the walk in such a slab by x , there exists a positive number r and a vector z such that with positive probability, the walk visits the slab for the last time at its r -th visit to $x + z$. From this one is then able to deduce that one must have $P_0[A_l] > 0$, a contradiction to (3.15). We refer the reader to [Ze02] for more details.

An inspection of the proof in [Ze02] yields that by slightly modifying it, one obtains (3.14) for e replaced by arbitrary $l \in \mathbb{S}^{d-1}$. In combination with the result of (3.12), and due to the zero-one law of Lemma 3.1, we may therefore omit assumption (3.11) and still obtain that (3.12) holds true.

- (iii) Using (3.12), we obtain that $\lim_{n \rightarrow \infty} X_n/n$ exists P_0 -a.s. and, also P_0 -a.s., takes values in a set of cardinality at most 2^d . One can then take advantage of similar arguments as Goergen on page 1112 of [Go06] in order to show that P_0 -a.s. $\lim_{n \rightarrow \infty} X_n/n$ takes values in a set of two elements which are collinear, which finishes the proof. Indeed, assume there were v_1, v_2 not collinear such that $P_0[\lim_{n \rightarrow \infty} X_n/n = v_i] > 0$ for $i = 1, 2$. Then for any l such that

$$l \cdot v_1, l \cdot v_2 > 0 \quad (3.16)$$

one obtains by (3.12) and the fact that

$$\left\{ \lim_{n \rightarrow \infty} X_n/n = v_1 \right\} \cup \left\{ \lim_{n \rightarrow \infty} X_n/n = v_2 \right\} \subset A_l,$$

that

$$l \cdot v_1 = v_l = l \cdot v_2. \quad (3.17)$$

Since the set of vectors l fulfilling (3.16) is open, we can let l vary along a set of basis vectors fulfilling (3.16) and hence conclude that (3.17) holds for a set of vectors l which form a basis. This implies $v_1 = v_2$, a contradiction to the assumption that v_1 and v_2 were collinear. This yields Theorem 3.7.

□

Remark 3.11. *It is useful to observe from part (i) of the proof of Lemma 3.10 that*

$$v_l \neq 0 \text{ if and only if } E_0[\tau_1 \mid D = \infty] < \infty. \quad (3.18)$$

This condition is in general hard to check — it will be one of the principal goals of the remaining part of these notes to investigate conditions that ensure $v_l \neq 0$.

3.4 Ballisticity

We have seen in Theorem 3.7 that a version of a law of large numbers is valid. This, however, did not tell us anything practical about the fundamental question of whether v_1 and v_2 are equal to or different from 0 (except for the one-dimensional setting of Theorem 2.16, Remark 3.11, and the result of Berger [Be08] alluded to above). Here, we will address this question and for this purpose recall the concept of ballisticity in a given direction (see Definition 2.14).

Remark 3.12. *If a RWRE is ballistic in a direction l according to Definition 2.14, then one can deduce that P_0 -a.s., the limit*

$$\lim_{n \rightarrow \infty} \frac{X_n \cdot l}{n} \text{ exists, is positive, and is } P_0\text{-a.s. constant.} \quad (3.19)$$

Indeed, if (2.28) is fulfilled, then $P_0[A_l] = 1$ and hence the renewal structure as introduced in Section 3.2 is P_0 -a.s. well-defined (cf. Lemma 3.4). Similarly to the proof of Theorem 3.7 one obtains that P_0 -a.s.,

$$\lim_{k \rightarrow \infty} \frac{X_{\tau_k} \cdot l}{\tau_k} = \frac{E_0[X_{\tau_1} \cdot l \mid D = \infty]}{E_0[\tau_1 \mid D = \infty]} \quad (3.20)$$

exists; using (2.28) we then infer that the expression in (3.20) must be positive, which implies (3.19).

If one wants to investigate the occurrence of ballistic behavior in higher dimensions, it is obvious that one cannot expect as simple conditions as in the one-dimensional case (cf. Theorem 2.16). As a partial remedy, Sznitman [Sz02] has introduced conditions which in some sense can be considered a higher-dimensional analog to the conditions given in Theorem 2.16 for dimension one. These conditions have turned out to be useful in a plethora of different contexts of RWRE.

Definition 3.13. (Conditions $(T)_\gamma$, (T') and (T)). Assume $l \in \mathbb{S}^{d-1}$ and $\gamma \in (0, 1]$. We say that condition $(T)_\gamma|l$ is satisfied if there exists a neighborhood V_l of l such that for every $l' \in V_l$ one has that

$$\limsup_{L \rightarrow \infty} \frac{1}{L^\gamma} \log P_0[H_{bL}^{-l'} < H_L^{l'}] < 0. \quad (3.21)$$

We say that condition $(T)|l$ is satisfied if condition $(T)_1|l$ holds. Finally, we say that condition $(T')|l$ is satisfied if for every $\gamma \in (0, 1)$, condition $(T)_\gamma|l$ is satisfied. Also, if the precise value of l is irrelevant, then we often write $(T)_\gamma$ instead of $(T)_\gamma|l$, and analogously for the remaining conditions.

Intuitively, if the walk escapes in direction l' and is “well-behaved”, then the probability in (3.21) corresponds to that of a rare event and, due to the independence structure of the environment, should decay reasonably fast.

Example 3.14. *Zerner and Sznitman [Ze98, Sz00] have introduced a classification of RWREs in terms of the support of the law of the random variable*

$$d(0, \omega) = \sum_{e \in U} \omega(0, e) \cdot e; \quad (3.22)$$

The random variable $d(0, \cdot)$ is the local drift at the origin. Denote by $C \subset D$ (cf. (2.2)) the convex hull of the support of the law of $d(0, \omega)$. An RWRE is called

(i) non-nestling if

$$0 \notin C;$$

(ii) marginally nestling if

$$0 \in \partial C;$$

(iii) plain nestling if

$$0 \in \overset{\circ}{C}.$$

In terms of investigating their ballistic behavior, the non-nestling and marginally nestling RWREs are easier to handle than the nestling ones. This is due to the fact that their behavior “dominates” that of i.i.d. variables with positive expectation. We leave it to the reader to prove that non-nestling RWRE satisfy condition (T).

For future purposes it will be helpful to also consider the corresponding polynomial analogues.

Definition 3.15. (Conditions $(\mathcal{P}^*)_M, (\mathcal{P}^*)_0$). Assume $M > 0$ and $l \in \mathbb{S}^{d-1}$ to be given. We say that condition $(\mathcal{P}^*)_M|l$ (sometimes referred to as $(\mathcal{P}^*)_M$ or (\mathcal{P}^*) also) is fulfilled, if there exists a neighborhood V_l of l such that for all $l' \in V_l$ and for all $b > 0$ we have

$$\lim_{L \rightarrow \infty} L^M P_0[H_{bL}^{-l'} < H_L^{l'}] = 0. \quad (3.23)$$

In addition, we define $(\mathcal{P}^*)_0$ to hold if for all l' in a neighborhood of l and for all $b > 0$ we have

$$\lim_{L \rightarrow \infty} P_0[H_{bL}^{-l'} < H_L^{l'}] = 0. \quad (3.24)$$

Remark 3.16. • *In the following we will give some fundamental results that were mostly proven under the assumption of condition (T'). However, in anticipation of Theorem 3.29 below, we will instead formulate them assuming $(\mathcal{P})_M$ for $M > 15d + 5$ only.*

- *Also, note that due to Theorem 3.29 it is actually sufficient to assume $(\mathcal{P})_M$ (see Definition 3.31) instead of $(\mathcal{P}^*)_M$, both for $M > 15d + 5$, in what follows. This condition is a priori weaker and has the advantage that it can be checked on finite boxes already. However, since it is more complicated to state and needs notation introduced only later on, we will not give its exact definition here yet.*

There is an alternative formulation for the conditions $(T)_\gamma$, which instead of considering slab exit estimates involves transience and the (stretched) exponential integrability of the renewal radii.

Theorem 3.17. ([Sz02, Cor. 1.5]) *Assume **(IID)** and **(UE)** to hold, and let furthermore $d \geq 1$ and $\gamma \in (0, 1]$. Then the following are equivalent.*

- (i) *Condition $(T)_\gamma|l$ is satisfied.*
- (ii) *One has $P_0[A_l] = 1$ (note that this ensures that τ_1 is well-defined) and there exists a constant $C > 0$ such that*

$$E_0 \left[\exp \left\{ C^{-1} \max_{0 \leq i \leq \tau_1} |X_i|_1^\gamma \right\} \right] < \infty. \quad (3.25)$$

Note that the first part of the condition (ii) in Theorem 3.17 in combination with the law of large numbers of Theorem 3.7 already supplies us with the fact that P_0 -a.s., $\lim_n X_n/n$ converges to a deterministic vector. Therefore, due to Theorem 3.18 below, the second part of condition (ii) in Theorem 3.17 can be seen as guaranteeing that this deterministic limit is different from 0. Note, however, that an affirmative answer to the open question 2.20 would imply that the transience assumption $P_0[A_l]$ is already sufficient and the integrability condition of (3.25) not needed for having a non-zero limiting velocity, i.e., ballisticity.

These stretched exponential integrability assumptions on the renewal radii have been used by Sznitman (see [Sz02]) to deduce the following: In dimensions larger than or equal to two, (T') implies a law of large numbers with non-zero limiting velocity as well as an invariance principle for the RWRE, so that diffusively rescaled it converges to Brownian motion under the averaged measure.

Theorem 3.18. ([Sz02, Thm. 3.3]) *Assume **(IID)** and **(UE)** to hold. Furthermore, assume $d \geq 2$ and let $(\mathcal{P})_M|l$ is fulfilled for some $l \in \mathbb{S}^{d-1}$ and $M > 15d + 5$. Then:*

- (i) *The RWRE is ballistic, i.e., one has P_0 -a.s. that*

$$\lim_{n \rightarrow \infty} \frac{X_n}{n} = v \neq 0, \quad (3.26)$$

and where v is deterministic.

- (ii) *Under P_0 and with*

$$B_t^n := \frac{1}{\sqrt{n}}(X_{[nt]} - [nt]v), \quad t \geq 0,$$

the sequence of processes $((B_t^n)_{t \geq 0})_{n \in \mathbb{N}}$ converges in law on the Skorokhod space $D([0, \infty); \mathbb{R}^d)$ to Brownian motion with non-degenerate covariance matrix as $n \rightarrow \infty$.

Remark 3.19. *Recently, there has also been initiated the investigation of ballisticity and related topics for the situation where **(IID)** holds, but the condition **(UE)** has been replaced by the weaker **(E)**. In this context, in order to obtain results comparable to the ones above, one then*

has to make assumptions on the decay of the random variables $\omega(0, e)$ at 0. These assumptions can be used to apply large deviations estimates in order to obtain that with high \mathbb{P} -probability, for sufficiently long paths, the probability of following them is comparable at least to a situation where one has uniform ellipticity; see Section 3.12 as well as Campos and Ramírez [CR13] for further details.

Open question 3.20. *Theorem 3.18 states that the conditions $(\mathcal{P})_M$ for $M > 15d+5$ do imply a ballistic behavior. Vice versa, one can ask if (3.26) already implies the validity of $(\mathcal{P})_M$ for $M > 15d+5$. This question is intimately linked to the slope of the large deviation principle rate function in the origin.*

As observed in Remark 3.11, in order to guarantee a positive limiting velocity, and therefore to prove Theorem 3.18 (i), it is enough to show the integrability of τ_1 with respect to $P[\cdot | D = \infty]$. On the other hand, in order to deduce Theorem 3.18 (ii), an essential part of the proof is to establish the square integrability of τ_1 with respect to $P[\cdot | D = \infty]$ (see also Theorem 4.1 in [Sz00]). Both of these integrability conditions are a direct consequence of the following recent result of Berger.

Theorem 3.21. *([Be12, Prop. 2.2]) Let (\mathbf{UE}) , (\mathbf{IID}) , and $(\mathcal{P})_M|l$ be fulfilled for some $l \in \mathbb{S}^{d-1}$ and $M > 15d+5$. Then, for $d \geq 4$ and every $\alpha < d$ one has that*

$$P_0[\tau_1 \geq u] \leq \exp\{-(\log u)^\alpha\}$$

for all u large enough.

In the plain nestling case, this asymptotics is very close to being optimal as can be seen by the use of so-called naïve traps (see proof of [Sz00, Thm. 2.7] for a more restricted version of these traps and [Sz04] also). These correspond to balls within which the local drift points in the direction of the origin, see Figure 3.2 as well. Using such traps one gets the following.

Theorem 3.22. *([Sz00, Thm. 2.7], [Sz04]) Assume (\mathbf{UE}) , (\mathbf{IID}) , and $(\mathcal{P})_M|l$ to be fulfilled for some $l \in \mathbb{S}^{d-1}$ and $M > 15d+5$. Then, for $d \geq 2$ there exists a constant C such that one has*

$$P_0[\tau_1 \geq u] \geq \exp\{-C(\log u)^d\}$$

for all u large enough.

As a corollary to Theorem 3.21, Berger obtained the following large deviations upper bound, essentially matching Sznitman's lower bound for the nestling case in [Sz00, Section 5]. In this result, we write v for the P_0 -a.s. non-zero limit of X_n/n , cf. Theorem 3.7 (i).

Theorem 3.23. *([Be12]) Let (\mathbf{UE}) and (\mathbf{IID}) be fulfilled. Assume furthermore that $d \geq 4$ and that $(\mathcal{P})_M|l$ is fulfilled for some $l \in \mathbb{S}^{d-1}$ and $M > 15d+5$. Then for $\alpha \in (0, d)$, $y \in \{tv : t \in [0, 1)\}$, and $\varepsilon \in (0, |y - v|_2)$, one has*

$$P_0[|X_n/n - y|_2 < \varepsilon] < \exp\{-(\log n)^\alpha\}$$

for all n large enough.

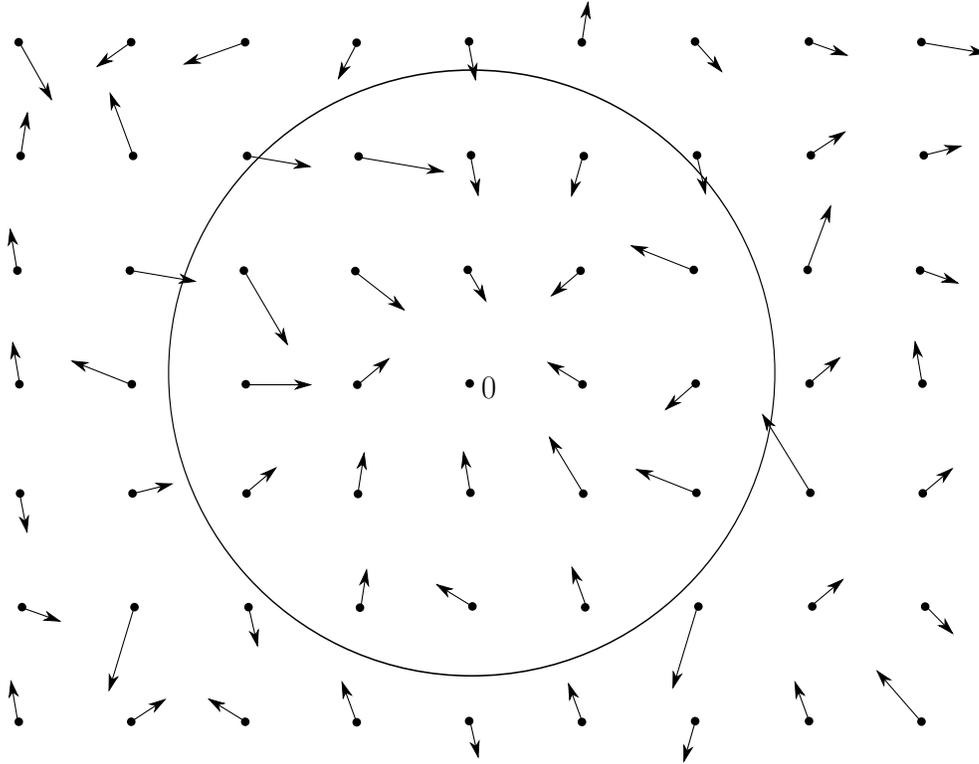


Figure 3.2: A realization of a naive trap: The local drifts within the ball point towards the origin whereas the local drifts outside the ball are arbitrary.

Remark 3.24. *In the results of [Be12], one has the standing assumption that $d \geq 4$. This assumption (in combination with (T')) is used in order to deduce that, on \mathbb{P} -average, two independent random walks in the same environment do not meet too often. It is plausible that a refinement of the methods in [Be12] might still yield corresponding results in $d = 3$; however, it seems that for the case $d = 2$ one essentially needs some further new ideas.*

3.5 How to check (T') on finite boxes

The conditions $(T)_\gamma$ in any of the formulations of Theorem 3.17, as well as the condition (\mathcal{P}^*) , are asymptotic in nature and therefore generally not easy to check. In this context, the effective criterion introduced by Sznitman [Sz02] proves to be a helpful tool for checking these conditions on finite boxes already. It can be seen as an analog to the ballisticity conditions of Solomon (cf. Theorem 2.16) in higher dimensions.¹

In order to introduce this criterion, for positive numbers L , L' and \tilde{L} as well as a space

¹Note that, while the condition $(\mathcal{P})_M$ of Definition 3.31 also is effective in the sense that it can be checked on finite boxes, the proof that it implies (T') takes advantage of the effective criterion (cf. Definition 3.31 and Theorem 3.29) — we therefore do introduce this criterion here.

rotation R around the origin we define the

box specification $\mathcal{B}(R, L, L', \tilde{L})$ as the box $B := \{x \in \mathbb{Z}^d : x \in R((-L, L') \times (-\tilde{L}, \tilde{L})^{d-1})\}$.

Recalling the notation of (2.1), we introduce

$$\rho_{\mathcal{B}}(\omega) := \frac{P_{0,\omega}[H_{\partial B} \neq H_{\partial_+ B}]}{P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}]},$$

where for a subset $A \subset \mathbb{Z}^d$, we use the notation

$$H_A := \inf\{n \geq 0 : X_n \in A\},$$

as well as

$$\partial_+ B := \{x \in \partial B : R(e_1) \cdot x \geq L', |R(e_j) \cdot x|_2 < \tilde{L} \forall j \in \{2, \dots, d\}\}.$$

We will sometimes write ρ instead of $\rho_{\mathcal{B}}$ if the box we refer to is clear from the context.

Definition 3.25. Given $l \in \mathbb{S}^{d-1}$, the *effective criterion with respect to l* is satisfied if for some $L > c_1$ and $\tilde{L} \in [3\sqrt{d}, L^3)$, we have that

$$\inf_{\mathcal{B}, a} \left\{ c_2 \left(\ln \frac{1}{\kappa} \right)^{3(d-1)} \tilde{L}^{d-1} L^{3(d-1)+1} \mathbb{E}[\rho_{\mathcal{B}}^a] \right\} < 1. \quad (3.27)$$

Here, when taking the infimum, a runs over $[0, 1]$ while \mathcal{B} runs over the

box specifications $\mathcal{B}(R, L-2, L+2, \tilde{L})$ with R a rotation around the origin such that $R(e_1) = l$. (3.28)

Furthermore, c_1 and c_2 are dimension dependent constants.

The effective criterion is of significant importance due to the combination of the facts that it can be checked on finite boxes (in comparison to (T') which is asymptotic in nature) and that it is equivalent to (T') , cf. Theorem 3.26 below.

Theorem 3.26 ([Sz02]). *Let (IID) and (UE) be fulfilled. Then for each $l \in \mathbb{S}^{d-1}$ the following conditions are equivalent.*

- (i) *The effective criterion with respect to l is satisfied.*
- (ii) *$(T')|l$ is satisfied.*

In the proof of Theorem 3.26, the estimate (3.27) serves as a seed estimate for an involved multi-scale renormalization scheme. We refer to the original source for the lengthy proof of this fundamental result, and to p. 239 ff. of [Sz04] for a reasonably detailed proof sketch.

3.6 Interrelation of stretched exponential ballisticity conditions

While a priori $(T)_\gamma$ is a weaker condition the smaller γ is, Sznitman [Sz02] showed that for each $\gamma \in (0.5, 1)$, the conditions $(T)_\gamma$ and (T') are equivalent. This equivalence has been extended by Drewitz and Ramírez [DR11] to some dimension dependant interval $(\gamma_d, 1)$, with $\gamma_d \in (0.366, 0.388)$, for all $d \geq 2$. Furthermore, it has been conjectured (see p. 227 in [Sz04]) that

$$\text{the conditions } (T)_\gamma|l \text{ are equivalent for all } \gamma \in (0, 1]. \quad (3.29)$$

Theorem 3.27 ([DR12, BDR12]). *Assume $d \geq 2$, **(UE)** and **(IID)** to hold. Then, for $l \in \mathbb{S}^{d-1}$, the conditions $(T)_\gamma|l$, $\gamma \in (0, 1)$, are all equivalent.*

Open question 3.28. *It is still not known if (T') is actually equivalent to condition (T) ; however, in some sense there is not missing “too much” in some sense (see [Sz02, Prop. 2.3]).*

According to Theorem 3.27, in order to check (T') , it is sufficient to check $(T)_\gamma$ for any γ small enough but positive. As alluded to before already, we will see in the next section that it is sufficient to establish the polynomial conditions $(\mathcal{P}^*)_M$ or $(\mathcal{P})_M$ for M large enough.

3.7 The condition $(\mathcal{P})_M$

The main result of this section will be that of [BDR12], namely that for M large enough, $(\mathcal{P})_M$ already implies the conditions $(T)_\gamma$ and hence all its consequences such as ballistic behavior and an invariance principle.

We will be guided by the presentation in [BDR12] — however, we will omit a significant share of the more technical parts of the proof and try to give a less rigorous and more intuitive description instead.

The main result of this section is the following.

Theorem 3.29 ([BDR12]). *Assume $d \geq 2$, **(IID)** and **(UE)** to be fulfilled. Let $l \in \mathbb{S}^{d-1}$ and assume that $(\mathcal{P}^*)_M|l$ or $(\mathcal{P})_M|l$ holds for some $M > 15d + 5$. Then $(T')|l$ holds.*

Remark 3.30. *The condition $M > 15d + 5$ looks quite arbitrary, and is indeed not the weakest condition possible. However, since with the methods we used it does not seem possible to significantly weaken this condition, we refrain from trying to do so.*

We are going to introduce some of the notation needed for the proof of Theorem 3.29 as well as give two propositions that play a fundamental role in the proof.

Let

$$c_3 = \exp \left\{ 100 + 4d(\ln \kappa)^2 \right\}, \quad (3.30)$$

let $N_0 \geq c_3$ be an even integer, and set $N_{-1} := 2N_0/3$. Using the notation

$$\pi_l : \mathbb{R}^d \ni x \mapsto (x \cdot l)l \in \mathbb{R}^d \quad (3.31)$$

to denote the orthogonal projection on the space $\{\lambda l : \lambda \in \mathbb{R}\}$, we introduce the box

$$B := \left\{ y \in \mathbb{Z}^d : -\frac{N_0}{2} < (y-x) \cdot l < N_0, |\pi_{l^\perp}(y-x)|_\infty < 25N_0^3 \right\}, \quad (3.32)$$

as well as their frontal parts

$$\tilde{B} := \left\{ y \in \mathbb{Z}^d : N_0 - N_{-1} \leq (y-x) \cdot l < N_0, |\pi_{l^\perp}(y-x)|_\infty < N_0^3 \right\}. \quad (3.33)$$

In addition, we define

$$\partial_+ B := \{y \in \partial B : (y-x) \cdot l \geq N_0\}. \quad (3.34)$$

To simplify notation, throughout we will denote a typical box of scale k by B_k , and its middle frontal part by \tilde{B}_k .

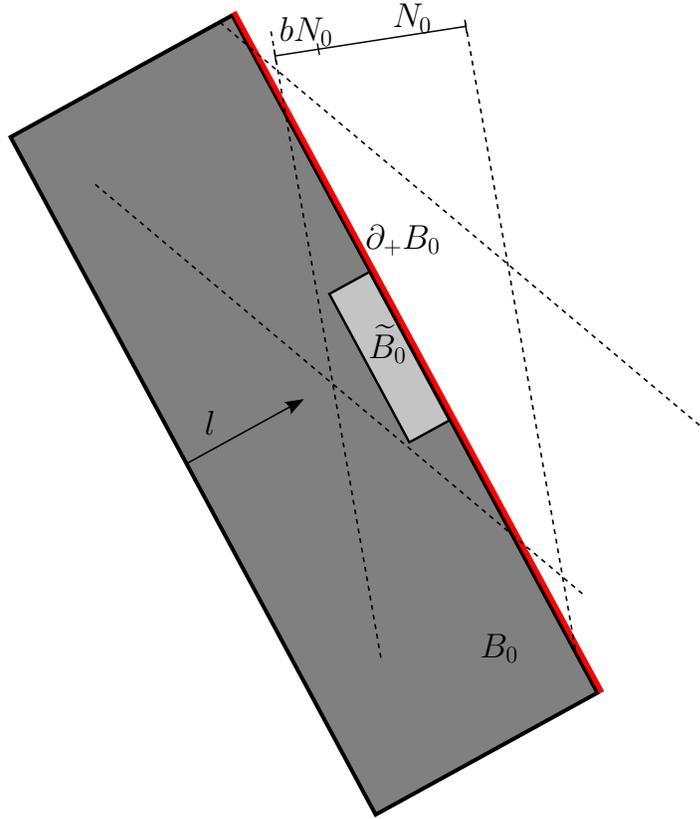


Figure 3.3: A box B_0 and the middle frontal part \tilde{B}_0 ; the dashed lines illustrate the slabs from the definition of $(\mathcal{P}^*)_M|l$, shifted by some $x \in \tilde{B}_0$; it is visually apparent here how condition (\mathcal{P}^*) implies condition (\mathcal{P}) .

Definition 3.31. Let $l \in \mathbb{S}^{d-1}$ and $M > 0$. We say that $(\mathcal{P})_M|l$ is fulfilled if

$$\sup_{x \in \tilde{B}_0} P_x [H_{\partial B_0} \neq H_{\partial_+ B_0}] < N_0^{-M} \quad (3.35)$$

holds for some $N_0 \geq c_3$.

3.8 An intermediate condition between $(\mathcal{P})_M$ and $(T)_\gamma$

We need a little further notation for stating this result in particular. To start with, for a given generic $l = l_1 \in \mathbb{S}^{d-1}$, we choose l_2, \dots, l_d arbitrarily in such a way that l_1, \dots, l_d forms an orthonormal basis of \mathbb{R}^d .

For $L > 0$, define

$$\mathcal{D}_L^l := \left\{ x \in \mathbb{Z}^d : -L \leq x \cdot l \leq 10L, |x \cdot l_k| \leq \frac{L^3 \ln \ln L}{\ln L} \forall k \in \{2, \dots, d\} \right\}$$

as well as its *frontal boundary part*

$$\partial_+ \mathcal{D}_L^l := \left\{ x \in \partial \mathcal{D}_L^l : \pi_l(x) \cdot l > 10L, |x \cdot l_k| \leq \frac{L^3 \ln \ln L}{\ln L} \forall k \in \{2, \dots, d\} \right\}.$$

In the following we will refer to the condition that

$$\text{for } l' \in \mathbb{S}^{d-1} \text{ one has } P_0 \left[H_{\partial \mathcal{D}_L^{l'}} < H_{\partial_+ \mathcal{D}_L^{l'}} \right] \leq \exp \left\{ -L^{\frac{(1+o(1)) \ln 2}{\ln \ln L}} \right\}, \quad (3.36)$$

as $L \rightarrow \infty$.

Definition 3.32. If (3.36) holds for all l' in a neighborhood of $l \in \mathbb{S}^{d-1}$, then we say that condition $(T)_{\gamma_L} | l$ is fulfilled.

Since γ_L tends to 0 as L tends to infinity, one observes that the condition $(T)_{\gamma_L}$ is weaker than $(T)_\gamma$ for any $\gamma > 0$.

On the other hand, while the condition $(T)_{\gamma_L}$ is a priori stronger than all of the polynomial conditions $(\mathcal{P}^*)_M$, $M > 0$, it can be shown that it is a consequence of $(\mathcal{P}^*)_M$ once M is chosen large enough. This is the content of Proposition 3.33 below.

3.9 Strategy of the proof of Theorem 3.29

Using Theorem 3.26, we observe that in order to prove Theorem 3.29, it is sufficient to establish the effective criterion departing from $(\mathcal{P})_M | l$ with M large enough. On a heuristic level, we will do so via two renormalization schemes:

- (i) The first one starts with assuming condition $(\mathcal{P})_M | l$ for some M large enough and derives the intermediate condition $(T)_{\gamma_L}$ introduced in Definition 3.32.

Proposition 3.33 (Sharpened averaged exit estimates). *Assume **(IID)** and **(UE)** to be fulfilled. Let $M > 15d + 5$, $l \in \mathbb{S}^{d-1}$, and assume that condition $(\mathcal{P})_M | l$ is satisfied. Then $(T)_{\gamma_L} | l$ holds.*

We will not give the technically involved proof of this result and refer to the original source [BDR12] instead.

- (ii) The second renormalization step supplies us with the following large deviations result.

Proposition 3.34 (Weak atypical quenched exit estimates, [BDR12]). *Let $d \geq 2$ and assume **(IID)** and **(UE)** to be fulfilled and let $(T)_{\gamma_L}|l$ hold. Then for $\epsilon(L) := \frac{1}{(\ln \ln L)^2}$, and any function $\beta : (0, \infty) \rightarrow (0, \infty)$, one has that*

$$\mathbb{P} \left[P_{0,\omega} [H_{\partial B} = H_{\partial_+ B}] \leq \frac{1}{2} \exp \{ -c_1 L^{\beta(L)} \} \right] \leq 5^d \frac{e}{\lceil L^{\beta(L) - \epsilon(L)} / 5^d \rceil!}, \quad (3.37)$$

where B is a box specification as in (3.28) with $\tilde{L} = L^3 - 1$, and

$$c_4 := -2d \ln \kappa > 1. \quad (3.38)$$

This result is much less technical to prove, but nevertheless we refer to [BDR12] for its proof in order not to lose the principal thread of these notes.

We do, however, mention that in dimensions $d \geq 4$, Proposition 3.34 can be strengthened significantly as follows:

Theorem 3.35 (Atypical quenched exit estimates, [DR12]). *Let $d \geq 4$, and assume **(IID)**, **(UE)**, and $(T)_{\gamma}|l$ to hold for some $\gamma \in (0, 1)$, $l \in \mathbb{S}^{d-1}$. Fix $c > 0$ and $\beta \in (0, 1)$. Then there exists a constant $C > 0$ such that for all $\alpha \in (0, \beta d)$,*

$$\limsup_{L \rightarrow \infty} L^{-\alpha} \log \mathbb{P} \left[P_{0,\omega} [H_{\partial B} = H_{\partial_+ B}] \leq e^{-cL^\beta} \right] < 0,$$

where B is a box specification as in (3.28) with $\tilde{L} = CL$.

The proof of this result is significantly more involved than that of Proposition 3.34. Note that this theorem is very close to being optimal in the sense that its conclusion will not hold in general for $\alpha > \beta d$. In fact, for plain nestling RWRE, this can be shown by the use of naïve traps introduced above.

For the purpose of proving Theorem 3.29, however, Proposition 3.34 is sufficient.

3.10 Proof of Theorem 3.29 assuming Propositions 3.33 and 3.34

In this section we demonstrate how Propositions 3.33 and 3.34 can be employed in order to establish the effective criterion. We will do so by rewriting $\mathbb{E}[\rho_B^a]$ of (3.27) as a sum of terms typically of the form

$$\mathcal{E}_j := \mathbb{E} \left[\rho_B^a, \frac{1}{2} \exp \{ -c_4 L^{\beta_{j+1}} \} < P_{0,\omega} [H_{\partial B} = H_{\partial_+ B}] \leq \frac{1}{2} \exp \{ -c_4 L^{\beta_j} \} \right] \quad (3.39)$$

with $\beta_{j+1} > \beta_j$.

Generally, the lower bound on $P_{0,\omega} [H_{\partial B} = H_{\partial_+ B}]$ in (3.39) yields a control on the integrand ρ_B^a from above, while the upper bound enforces an atypical behavior which will be exploited using Proposition 3.34. The interplay of the upper bound of the integrand thus obtained with

the estimate from Proposition 3.34 will then determine the asymptotics we obtain for \mathcal{E}_j (cf. also Lemma 3.37 below).

Our proof of Theorem 3.29 goes along establishing the effective criterion. We do so by a subtle decomposition of the expectation occurring in (3.27) into several summands, and in the following we will give some basic lemmas that will prove useful in estimating each of these summands.

For that purpose, we define the quantities

$$\beta_1(L) := \frac{\gamma_L}{2} = \frac{\ln 2}{2 \ln \ln L}, \quad (3.40)$$

$$a := L^{-\gamma_L/3}, \quad (3.41)$$

and write ρ for ρ_B with some arbitrary box specification of (3.28) with $\tilde{L} = L^3 - 1$. We split $\mathbb{E}[\rho^a]$ according to

$$E[\rho^a] = \mathcal{E}_0 + \sum_{j=1}^{n-1} \mathcal{E}_j + \mathcal{E}_n, \quad (3.42)$$

where

$$n := n(L) := \left\lceil \frac{4(1 - \gamma_L/2)}{\gamma_L} \right\rceil + 1,$$

$$\mathcal{E}_0 := \mathbb{E} \left[\rho^a, P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}] > \frac{1}{2} \exp \{ -c_4 L^{\beta_1} \} \right],$$

$$\mathcal{E}_j := \mathbb{E} \left[\rho^a, \frac{1}{2} \exp \{ -c_4 L^{\beta_{j+1}} \} < P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}] \leq \frac{1}{2} \exp \{ -c_4 L^{\beta_j} \} \right]$$

for $j \in \{1, \dots, n-1\}$, and

$$\mathcal{E}_n := \mathbb{E} \left[\rho^a, P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}] \leq \frac{1}{2} \exp \{ -c_4 L^{\beta_n} \} \right],$$

with parameters

$$\beta_j(L) := \beta_1(L) + (j-1) \frac{\gamma_L}{4}, \quad (3.43)$$

for $2 \leq j \leq n(L)$; for the sake of brevity we may sometimes omit the dependence on L of the parameters if that does not cause any confusion. Furthermore, in order to verify equality (3.42), note that due to the uniform ellipticity assumption **(UE)** and the choice of c_4 (cf. (3.38)), one has for \mathbb{P} -a.a. ω that

$$P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}] > e^{-c_4 L},$$

as well as that

$$\beta_n > 1.$$

To bound \mathcal{E}_0 we employ the following lemma.

Lemma 3.36. *Let $(T)_{\gamma_L}$ be fulfilled. Then*

$$\mathcal{E}_0 \leq \exp \{ c_4 L^{\gamma_L/6} - L^{(1+o(1))\gamma_L/2} \},$$

as $L \rightarrow \infty$.

PROOF. Jensen's inequality yields

$$\mathcal{E}_0 \leq 2 \exp \{c_4 L^{\beta_1 - \gamma_L/3}\} P_0[H_{\partial B} \neq H_{\partial_+ B}]^a.$$

Using (3.40) in combination with $(T)_{\gamma_L}$ we obtain the desired result. \square

To deal with the middle summand in the right-hand side of (3.42), we use the following lemma.

Lemma 3.37. *Let (IID) and (UE) be fulfilled and assume $(T)_{\gamma_L}|l$ to hold. Then for all L large enough we have uniformly in $j \in \{1, \dots, n-1\}$ that*

$$\mathcal{E}_j \leq 2 \cdot 5^d \exp \{c_4 L^{\beta_{j+1} - \gamma_L/3}\} \frac{e}{\lceil L^{\beta_j - \epsilon(L)}/5^d \rceil!}.$$

PROOF. Using Markov's inequality, for $j \in \{1, \dots, n-1\}$ we obtain the estimate

$$\mathcal{E}_j \leq 2 \exp \{c_4 L^{\beta_{j+1} - \gamma_L/3}\} \mathbb{P} \left[P_{0,\omega}[H_{\partial B} = H_{\partial_+ B}] \leq \frac{1}{2} \exp \{ -c_4 L^{\beta_j} \} \right]. \quad (3.44)$$

Thus, due to Proposition 3.34, the probability on the right-hand side of (3.44) can be estimated from above by

$$5^d \frac{e}{\lceil L^{\beta_j - \epsilon(L)}/5^d \rceil!}.$$

\square

When it comes to the term \mathcal{E}_n in (3.42) we note that it vanishes because of the choice of c_4 .

Proof of Theorem 3.29. It follows from Lemmas 3.36, 3.37, the choice of parameters in (3.40) to (3.41) and (3.43), and the fact that \mathcal{E}_n vanishes, that for L large enough, (3.42) can be bounded from above by

$$\exp \{c_4 L^{\gamma_L/6} - L^{(1+o(1))\gamma_L/2}\} + 2 \cdot 5^d n(L) \max_{1 \leq j \leq n(L)-1} \left(\exp \{c_4 L^{\beta_{j+1} - \gamma_L/3}\} \frac{e}{\lceil L^{\beta_j - \epsilon(L)}/5^d \rceil!} \right).$$

Thus, we see that for our choice of parameters, (3.42) tends to zero faster than any polynomial in L . Hence, due to (3.27), the effective criterion holds and Theorem 3.26 then yields the desired result. \square

3.11 Relation between directional transience and slab exit estimates

The aim of this subsection is to show how the condition of directional transience relates to slab exit estimates such as condition $(\mathcal{P}^*)_M$.

Lemma 3.38. *Let $l \in \mathbb{S}^{d-1}$, and suppose that $(\mathcal{P}^*)_M|l$ is satisfied.*

(i) There exists a constant C such that

$$P_0[T_{-L}^{-l} \circ \theta_{T_{2L}^l} \leq T_{4L}^l \circ \theta_{T_{2L}^l}] \leq CL^{d-1-M} \quad (3.45)$$

for all $L \in \mathbb{N}$.

(ii) If $M > d$, then P_0 -a.s. the random walk X is transient in direction l , i.e. $P_0[A_l] = 1$.

PROOF. (i) The general idea of this proof is taken from a stretched exponential analog [Sz02, Theorem 2.11]. Note that

$$\{T_L^l < T_L^{-l}\} \subset \left\{ \limsup_{n \rightarrow \infty} X_n \cdot l \geq L \right\}.$$

Due to condition $(\mathcal{P}^*)_M|l$, the probability with respect to P_0 of the left-hand side tends to 1 as $L \rightarrow \infty$, which implies

$$P_0 \left[\limsup_{n \rightarrow \infty} X_n \cdot l = \infty \right] = 1. \quad (3.46)$$

Now choose $l_1, \dots, l_d \in \mathbb{S}^{d-1} \cap V_l$ (with V_l denoting the neighborhood associated to l in the definition of $(\mathcal{P}^*)_M|l$, see Definition 3.15) to be linearly independent. If furthermore l_1, \dots, l_d are chosen sufficiently close to l , setting $l_0 := l$ there exists $\delta > 0$ such that for

$$\Delta_L := \{x \in \mathbb{Z}^d : -\delta L \leq x \cdot l_j \leq L \forall 0 \leq j \leq d\} \quad (3.47)$$

and

$$\partial_+ \Delta_L := \left\{ x \in \partial \Delta_L : \max_{0 \leq j \leq d} x \cdot l_j > L \text{ and } \min_{0 \leq j \leq d} x \cdot l_j \geq -\delta L \right\},$$

we have

$$2\delta L \leq \min\{x \cdot l : x \in \partial_+ \Delta_L\}. \quad (3.48)$$

Now due to (3.46), we infer that T_L^l is finite P_0 -a.s. and hence $\theta_{T_L^l}$ is well-defined for all $L > 0$. Thus, we get using the strong Markov property at time $T_{2\delta L}^l$ (applied to the quenched walk) in combination with the translation invariance of \mathbb{P} and (3.48), that

$$\begin{aligned} & P_0[T_{-\delta L}^{-l} \circ \theta_{T_{2\delta L}^l} \leq T_{4\delta L}^l \circ \theta_{T_{2\delta L}^l}] \\ & \leq P_0[T_{\partial \Delta_L} < T_{2\delta L}^l] + P_0[T_{-\delta L}^{-l} \circ \theta_{T_{2\delta L}^l} \leq T_{4\delta L}^l \circ \theta_{T_{2\delta L}^l}, T_{\partial \Delta_L} \geq T_{2\delta L}^l] \\ & \leq \sum_{j=0}^d P_0[T_{\delta L}^{-l_j} < T_L^{l_j}] + CL^{d-1} P_0[T_{\delta L}^{-l} \leq T_{2\delta L}^l]. \end{aligned} \quad (3.49)$$

To obtain the last line we used the fact that, since l_1, \dots, l_d form a basis, $|\{x \in \Delta_L : x \cdot l \in (2\delta L, 2\delta L + 1]\}| \leq CL^{d-1}$ holds. Since furthermore $(\mathcal{P}^*)_M|l$ is fulfilled we can estimate (3.49) from above by CL^{d-1-M} which proves the first assertion of the lemma.

- (ii) Using this result in combination with the assumption that $M > d$, Borel-Cantelli's lemma yields that P_0 -a.s., for eventually all $L \in \mathbb{N}$,

$$T_{4L}^l \circ \theta_{T_{2L}^l} < T_{-L}^{-l} \circ \theta_{T_{2L}^l}.$$

This implies that $P_0[\lim_{n \rightarrow \infty} X_n \cdot l = \infty] = 1$.

□

We have the following corollary on the relation between transience and the conditions $(\mathcal{P}^*)_M$.

Corollary 3.39. *The implications*

$$(\mathcal{P}^*)_M | l \text{ for some } M > d \implies P_0[A_{l'}] = 1 \forall l' \text{ in a neighborhood } V_l \text{ of } l \implies (\mathcal{P}^*)_0 | l \quad (3.50)$$

hold true.

PROOF. The first implication is a direct consequence of Lemma 3.38. To obtain the second implication note that if $P_0[A_{l'}] = 1$ for all $l' \in V_l$, then we have

$$P_0[H_{bL}^{-l'} < H_L^{l'}] \leq P_0[A_{l'}, H_{bL}^{-l'} < \infty] \rightarrow 0, \quad \text{as } L \rightarrow \infty,$$

where we used that $P_0[\cdot | A_{l'}]$ -a.s. one has $\inf_{n \in \mathbb{N}} X_n \cdot l' \in (-\infty, 0]$.

□

Remark 3.40. *The above corollary immediately leads to two questions:*

- (i) *Which is the minimal M for which the first implication holds?*
- (ii) *Can $(\mathcal{P}^*)_0$ on the right-hand side of the implications be replaced by $(\mathcal{P}^*)_M$ for some $M > 0$, and if so, what is the maximal M ?*

These questions are intimately connected to open question 2.20.

3.12 Ellipticity conditions for ballistic behavior

We have seen in Chapter 2 that there can exist elliptic random walks which are transient in a given direction but which are not ballistic. On the other hand, Proposition 2.17 of this chapter shows that at least some condition on the moments of the jump probabilities of the random environment should be asked if we expect to extend the results of this chapter.

Definition 3.41. Consider a RW in an environment \mathbb{P} . We say that \mathbb{P} satisfies the ellipticity condition $(E)_\beta$ if there exist positive parameters $\{\beta_e : e \in U\}$ such that

$$2 \sum_{e \in U} \beta_e - \sup_{e'} (\beta_{e'} + \beta_{-e'}) > \beta$$

and

$$\mathbb{E} \left[e^{\sum_e \beta_e \log \frac{1}{\omega(0,e)}} \right] < \infty.$$

If in addition there exists a $\bar{\beta}$ such that $\beta_e = \bar{\beta}$ for e such that $e \cdot \hat{v} \geq 0$ (recall that \hat{v} was the asymptotic direction) while $\beta_e \leq \bar{\beta}$ for e such that $e \cdot \hat{v} < 0$, we say that condition $(E)_\beta$ is satisfied directionally. Furthermore, whenever there exists an $\alpha > 0$ such that

$$\sup_e \mathbb{E} \left[\frac{1}{\omega(0,e)^\alpha} \right] < \infty$$

we say that the law \mathbb{P} of the environment satisfies condition $(E')_\alpha$.

We have the following extension of Theorem 3.29 proved in [CR13].

Theorem 3.42. (Campos-Ramírez) *Consider a random walk in an i.i.d. environment which satisfies condition $(E')_\alpha$ for some $\alpha > 0$. Then, if $(\mathcal{P}^*)_M|l$ is satisfied for some $M \geq 15d + 5$, $(T')|l$ is satisfied.*

Furthermore, we have then the following consequence of Theorem 3.42 proved in [CR13].

Theorem 3.43. (Campos-Ramírez) *Consider a random walk in an i.i.d. environment which satisfies condition $(E)_1$ directionally. Then, if $(\mathcal{P}^*)_M|l$ is satisfied for $M \geq 15d + 5$, the walk is ballistic.*

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