

## THE QUENCHED CENTRAL LIMIT THEOREM FOR A MODEL OF RANDOM WALK IN RANDOM ENVIRONMENT

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**ABSTRACT.** In the present paper we provide a proof of the quenched central limit theorem for the random walk in random environment model introduced by Boldrighini, Minlos, and Pellegrinotti in [3].

У цій статті дано доведення квенч-центральної граничної теореми для випадкових блукань у моделі з випадковим середовищем, запропонованій Болдрігіні, Мілосом і Пеллегрінотті [3].

### 1. INTRODUCTION

In this article we prove the quenched Central Limit Theorem (CLT) for a model of random walk in random environment, as it has been introduced by Boldrighini, Minlos, and Pellegrinotti, see in particular [3, 4, 6]. At each site the transition probability kernel is affected by the current state of the environment at this site. A more detailed description can be found in Section 2. For a nice overview of the literature on the subject, we refer to [5], and a survey on the recent progress in the related models can be found in [10] or [2]. In [7] the anomalous behavior of the corrections to the CLT in low dimensions has been studied. Moreover, related models are considered in [1] and [9].

We underline that the novelty of this work is that the proofs are essentially based on the multidimensional martingale CLT by K uchler and S orensen, see [8].

The paper is organized as follows: in Section 2 we describe the model and give the statement; while in Section 3 we provide all the proofs and some further comments.

### 2. MODEL, CONDITIONS AND RESULTS

Consider a particle moving in a  $n$ -dimensional infinite lattice and denote by  $X_t$  its position at time  $t$ . On the lattice, a dynamical random environment is considered. It is described by the random field

$$\xi = \{\xi_t(x) : x \in \mathbb{Z}^n, t \in \mathbb{Z}^+\}$$

Note that the time is discrete. We assume that  $\xi$  is the result of independent copies of the same random variable taking values in some finite space  $\mathbb{S}$ . The space of configurations is given by  $\tilde{\Omega} = \mathbb{S}^{\mathbb{Z}^n \times \mathbb{Z}^+}$ . Thus,  $\{\xi_t(x)\}_{(x,t) \in \mathbb{Z}^n \times \mathbb{Z}^+}$  is a collection of i.i.d random variables, distributed according to a given probability measure on  $\mathbb{S}$  denoted by  $\pi$ . We denote by  $\Pi$  the distribution of  $\xi$  in  $\tilde{\Omega}$ .

The one step transition probability from position  $x$  at time  $t$  to position  $y$  at the subsequent time step  $t + 1$  is given by

$$\mathbb{P}\{X_{t+1} = y | X_t = x, \xi\} = P(y - x, \xi_t(x)) = P_0(y - x) + c(y - x, \xi_t(x))$$

where  $P_0$  is the transition probability of a free random walk and  $c$  is the function which provides the influence of the environment on the particle's dynamic. We note that in the original work [3] there was a small factor  $\varepsilon$  before the function  $c$ .

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*Keywords.* Random walk in random environment; quenched central limit theorem; multidimensional martingale central limit theorem.

In order for the probability  $P$  to be well-defined, the following conditions must be fulfilled:

- $0 \leq P(u, s) = P_0(u) + c(u, s) \leq 1 \quad \forall s \in \mathbb{S} \quad \forall u \in \mathbb{Z}^n$ ;
- $\sum_{u \in \mathbb{Z}^n} c(u, s) = 0 \quad \forall s \in \mathbb{S}$ .

Without loss of generality we assume that the random environment has the following property:

$$\sum_{s \in \mathbb{S}} c(u, s) \pi(s) = 0 \text{ for any } u \in \mathbb{Z}^n, \quad (2.1)$$

which means that  $P_0$  is the mean transition probability. Indeed, if (2.1) is not satisfied, then we can replace  $P_0(u)$  with  $\tilde{P}_0(u) := P_0(u) + \sum_{s \in \mathbb{S}} c(u, s) \pi(s)$  and  $c$  with  $\tilde{c}(u, s) = c(u, s) - \tilde{P}_0(u) + P_0(u)$ . This replacement wouldn't change the law of the random walk and (2.1) would hold.

Additionally, let  $P_0$  and  $c$  be of bounded range. We denote by  $\mathbb{P}_\xi$  the conditional probability with respect to the environment  $\xi$ .

We will also assume that  $\sum_{u \in \mathbb{Z}^n} uc(u, s_1) = \sum_{u \in \mathbb{Z}^n} uc(u, s_2)$  for  $s_1, s_2 \in \mathbb{S}$ . It then follows from (2.1) that in fact

$$\sum_{u \in \mathbb{Z}^n} uc(u, s) = 0 \quad \forall s \in \mathbb{S}. \quad (2.2)$$

Let  $Y = \{Y_t\}_{t \in \mathbb{Z}^+}$  be the stochastic processes defined by  $Y_t = X_t - tb$ , where  $b = \sum_{u \in \mathbb{Z}^n} uP_0(u)$ . Note that

$$\sum_{u \in \mathbb{Z}^n} (u - b)P_0(u) = 0. \quad (2.3)$$

For a vector  $u \in \mathbb{R}^n$ ,  $u_i$  denotes its  $i$ -th coordinate.

**Theorem 1.** *For almost every realization  $\xi$  of the random environment we have*

$$\frac{1}{\sqrt{t}}Y_t \Rightarrow \eta^2 U, \quad (2.4)$$

$\mathbb{P}_\xi$  -a.s., where  $U$  is a standard normal vector and  $\eta^2$  is the positive semidefinite matrix with entries

$$(\eta^2)_{ij} = \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j)P_0(u). \quad (2.5)$$

### 3. PROOFS

**Lemma 1.** *For every  $\xi \in \tilde{\Omega}$ , the process  $Y$  is a martingale under  $\mathbb{P}_\xi$ .*

**Proof.** This is a consequence of the definition of  $Y$  along with the condition (2.2). Indeed, by (2.3), we have

$$\begin{aligned} \mathbb{E}[Y_{t+1}|Y_t, Y_{t-1}, \dots, Y_0] &= \mathbb{E}[Y_t + (Y_{t+1} - Y_t)|Y_t] \\ &= Y_t + \sum_{u \in \mathbb{Z}^n} (u - b) [P_0(u) + c(u, \xi_t)] = Y_t. \end{aligned}$$

□

Let us define the following  $n \times n$  matrices:  $H_t = \mathbb{E}(Y_t Y_t')$ , where  $Y_t'$  the transposed matrix, the matrix  $[Y]_t = ([Y^i, Y^j]_t)_{1 \leq i, j \leq n}$ , and  $H_t^\xi = \mathbb{E}_\xi(Y_t Y_t')$  (here and below we treat  $Y_t$  as a column-vector). Let also  $K_t = \frac{1}{\sqrt{t}}I_n$ , where  $I_n$  is the  $n \times n$  identity matrix. Then, the following result holds true.

**Lemma 2.** *We have*

$$\begin{aligned} & \mathbb{E}_\xi \left[ (Y_{r+1}^i - Y_r^i)(Y_{r+1}^j - Y_r^j) \right] \\ &= \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_r = y \mid \xi\} \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, \xi_r(y))] . \end{aligned} \tag{3.6}$$

The above sum by  $y$  is taken over the countable set

$$\mathcal{Y} := \{z_1 + z_2 b \mid z_1, z_2 \in \mathbb{Z}\} .$$

(Note that  $\mathbb{P}\{Y_t \in \mathcal{Y} \text{ for all } t \in \mathbb{N}\} = 1$ ).

**Proof.** By definition of  $Y$  and  $\mathbb{P}_\xi$ , we have

$$\begin{aligned} & \mathbb{E}_\xi \left[ (Y_{r+1}^i - Y_r^i)(Y_{r+1}^j - Y_r^j) \right] \\ &= \mathbb{E} \left[ (Y_{r+1}^i - Y_r^i)(Y_{r+1}^j - Y_r^j) \mid \xi \right] = \mathbb{E} \left[ \mathbb{E} \left\{ (Y_{r+1}^i - Y_r^i)(Y_{r+1}^j - Y_r^j) \mid \xi, Y_r \right\} \mid \xi \right] \\ &= \mathbb{E} \left[ \sum_u (u_i - b_i - Y_r^i)(u_j - b_j - Y_r^j) [P_0(u - Y_r) + c(u - Y_r, \xi(Y_r))] \mid \xi \right] \\ &= \mathbb{E} \left[ \sum_u (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, \xi(Y_r))] \mid \xi \right] \\ &= \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_r = y \mid \xi\} \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, \xi_r(y))] . \end{aligned}$$

□

**Lemma 3.** *For  $\Pi$ -a.a.  $\xi$ , we have  $\mathbb{P}_\xi$ -a.s.*

$$\frac{\#\{r : r \leq t, \xi_r(Y_r) = s\}}{t} \rightarrow \pi(s), \quad t \rightarrow \infty. \tag{3.7}$$

**Proof.** Recall that  $\Pi$  is defined on Page 311. The events  $\{Y_r = y\}$  and  $\{\xi_r(y) = s\}$  are independent, so (3.7) holds  $\mathbb{P}$ -a.s. by the law of large numbers. Hence (3.7) also holds  $\mathbb{P}_\xi$ -a.s for  $\Pi$ -a.a.  $\xi$ , otherwise, denoting the event in (3.7) by  $A$ , we would have

$$\mathbb{P}(A) = \int \mathbb{P}_\xi(A) \Pi(d\xi) < 1 .$$

**Lemma 4.** *We have*

$$\frac{1}{t} [Y]_t \rightarrow \eta^2, \tag{3.8}$$

$\mathbb{P}_\xi$ -a.s. for  $\Pi$ -a.a.  $\xi$ .

**Proof.** Note that for  $1 \leq i, j \leq n$ ,

$$([Y]_t)_{ij} = \sum_{0 \leq r < t} \Delta_{r,ij},$$

where

$$\Delta_{r,ij} = [Y_{r+1}^i - Y_r^i] [Y_{r+1}^j - Y_r^j] .$$

Under  $\mathbb{P}_\xi$  conditionally on  $\{Y_t = y\}$  the distribution of  $Y_{t+1} - Y_t$  is  $P_0(u) + c(u, \xi_t(y))$ . Since under  $\mathbb{P}_\xi$  the random vectors  $Y_{t+1} - Y_t$  are independent of each other for different  $t$ , the statement of the lemma follows from the law of large numbers. Indeed, by (3.7) and the law of large numbers  $\mathbb{P}_\xi$ -a.s.

$$\sum_{\{r:r \leq t, \xi_r(Y_r)=s\}} \frac{\Delta_{r,ij}}{\#\{r : r \leq t, \xi_r(Y_r) = s\}} \rightarrow \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) P(u, s),$$

and hence by (3.7) we obtain

$$\sum_{\{r:r \leq t, \xi_r(Y_r)=s\}} \frac{\Delta_{r,ij}}{\pi(s)t} \rightarrow \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j)P(u, s). \quad (3.9)$$

Therefore,  $\mathbb{P}_\xi$ -a.s.

$$\begin{aligned} \lim_{t \rightarrow \infty} \sum_{\{r:r \leq t\}} \frac{\Delta_{r,ij}}{t} &= \lim_{t \rightarrow \infty} \sum_{s \in \mathbb{S}} \sum_{\{r:r \leq t, \xi_r(Y_r)=s\}} \frac{\Delta_{r,ij}}{t} = \lim_{t \rightarrow \infty} \sum_{s \in \mathbb{S}} \pi(s) \sum_{\{r:r \leq t, \xi_r(Y_r)=s\}} \frac{\Delta_{r,ij}}{\pi(s)t} \\ &= \sum_{s \in \mathbb{S}} \pi(s) \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j)P(u, s) = \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j)P_0(u). \end{aligned}$$

□

**Corollary 1.** *The convergence in Lemma 4 also holds  $\mathbb{P}$ -a.s.*

**Lemma 5.**

(i) *We have*

$$(H_{r+1})_{ij} - (H_r)_{ij} = \sum_{s \in \mathbb{S}} \pi(s) \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, s)]. \quad (3.10)$$

(ii) *We also have*

$$\begin{aligned} &(H_{r+1}^\xi)_{ij} - (H_r^\xi)_{ij} \\ &= \sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, \xi_r(y))]. \end{aligned} \quad (3.11)$$

**Proof.** (i) We start by noting that for  $i, j \in \{1, \dots, n\}$ ,

$$\mathbb{E} \left( (Y_{t+1}^i - Y_t^i) Y_t^j \right) = 0. \quad (3.12)$$

Indeed,

$$\begin{aligned} &\mathbb{E} \left( (Y_{t+1}^i - Y_t^i) Y_t^j \right) = \mathbb{E} \mathbb{E} \left[ (Y_{t+1}^i - Y_t^i) Y_t^j \middle| Y_t \right] \\ &= \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} \sum_{u \in \mathbb{Z}^n} (y_i + u_i - b_i - y_i) y_j P_0(u) = \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} y_j \sum_{u \in \mathbb{Z}^n} (u_i - b_i) P_0(u) = 0 \end{aligned}$$

by (2.3). Moreover, by (3.12), we have

$$\begin{aligned} (H_{r+1})_{ij} - (H_r)_{ij} &= \mathbb{E} \left( Y_{t+1}^i Y_{t+1}^j - Y_t^i Y_t^j \right) \\ &= \mathbb{E} \left( (Y_{t+1}^i - Y_t^i) (Y_{t+1}^j - Y_t^j) \right) + \mathbb{E} \left( (Y_{t+1}^i - Y_t^i) Y_t^j \right) + \mathbb{E} \left( Y_t^i (Y_{t+1}^j - Y_t^j) \right) \\ &= \mathbb{E} \left( (Y_{t+1}^i - Y_t^i) (Y_{t+1}^j - Y_t^j) \right). \end{aligned}$$

Conditioning on  $Y_t$ , we get

$$\begin{aligned} (H_{r+1})_{ij} - (H_r)_{ij} &= \sum_y P\{Y_t = y\} \sum_u (u_i - b_i)(u_j - b_j) [P_0(u) + \mathbb{E}[c(u, \xi_r(y)) | Y_t = y]] \\ &= \sum_u (u_i - b_i)(u_j - b_j) P_0(u). \end{aligned}$$

(ii) (3.12) holds for  $\mathbb{E}_\xi$  too, since

$$\begin{aligned} &\mathbb{E} \mathbb{E} \left[ (Y_{t+1}^i - Y_t^i) Y_t^j \middle| Y_t, \xi \right] \\ &= \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} \sum_{u \in \mathbb{Z}^n} (y_i + u_i - b_i - y_i) y_j [P_0(u) + c(u, \xi_t(y))] \end{aligned}$$

$$\begin{aligned}
 &= \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} y_j \sum_{u \in \mathbb{Z}^n} (u_i - b_i) P_0(u) + \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} y_j \sum_{u \in \mathbb{Z}^n} u_i c(u, \xi_t(y)) \\
 &\quad - b_i \sum_{y \in \mathcal{Y}} \mathbb{P}\{Y_t = y\} y_j \sum_{u \in \mathbb{Z}^n} c(u, \xi_t(y)) = 0
 \end{aligned}$$

and the proof continues as in (i). □

**Lemma 6.** *We have*

$$\frac{1}{t} H_t \rightarrow \eta^2, \quad \frac{1}{t} H_t^\xi \rightarrow \eta^2, \tag{3.13}$$

where  $\eta^2$  is as in (2.5),  $\mathbb{P}_\xi$ -a.s for  $\Pi$ -a.a.  $\xi$ .

**Proof.** Let us only prove the second convergence in (3.13). By Lemma 5,

$$\begin{aligned}
 (H_t^\xi)_{ij} &= \sum_{r=0}^{t-1} \sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) [P_0(u) + c(u, \xi_r(y))] \\
 &= t \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) P_0(u) \\
 &\quad + \sum_{u \in \mathbb{Z}^n} (u_i - b_i)(u_j - b_j) \sum_{r=0}^{t-1} \sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} c(u, \xi_r(y)).
 \end{aligned} \tag{3.14}$$

Since  $\sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} c(u, \xi_r(y)) \stackrel{(d)}{=} c(u, \xi_r(\mathbf{0}))$  under  $\mathbb{P}$  (the symbol  $\stackrel{(d)}{=}$  means here ‘equal in distribution’), where  $\mathbf{0}$  is the origin, and

$$\sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} c(u, \xi_r(y)), \quad r \in \mathbb{N},$$

constitute a sequence of independent random variables indexed by  $r \in \mathbb{N}$ , by the law of large numbers for  $\Pi$ -a.a.  $\xi$

$$\frac{1}{t} \sum_{r=0}^{t-1} \sum_{y \in \mathcal{Y}} \mathbb{P}_\xi\{Y_r = y\} c(u, \xi_r(y)) \rightarrow \sum_{s \in \mathcal{S}} \pi(s) c(u, s). \tag{3.15}$$

Combining (3.14) and (3.15) and recalling the definition of  $\eta^2$ , we get the desired result. □

Recall that we defined  $K_t = \frac{1}{\sqrt{t}} I_n$ .

**Proof of Theorem 1.** Theorem 2.1 in [8] and Lemmas 4 and 6 imply that  $P_\xi$ -a.s.

$$\frac{1}{\sqrt{t}} Y_t \Rightarrow \eta^2 U, \tag{3.16}$$

where  $U$  is a standard  $n$ -dimensional Gaussian vector. The theorem in [8] is formulated for continuous time processes, so to apply it we define  $Y_t, H_t$ , etc. for  $t \in (1, \infty)$  by  $Y_t = Y_{\lfloor t \rfloor}, H_t = H_{\lfloor t \rfloor}$ , etc. □

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