

# Limit theorems for one and two-dimensional random walks in random scenery<sup>1</sup>

Fabienne Castell<sup>a</sup>, Nadine Guillotin-Plantard<sup>b</sup> and Françoise Pène<sup>c</sup>

<sup>a</sup>LATP, UMR CNRS 6632, Centre de Mathématiques et Informatique, Université Aix-Marseille I, 39, rue Joliot Curie, 13453 Marseille Cedex 13, France. E-mail: [Fabienne.Castell@cmi.univ-mrs.fr](mailto:Fabienne.Castell@cmi.univ-mrs.fr)

<sup>b</sup>Institut Camille Jordan, CNRS UMR 5208, Université de Lyon, Université Lyon 1, 43, Boulevard du 11 novembre 1918, 69622 Villeurbanne, France. E-mail: [nadine.guillotin@univ-lyon1.fr](mailto:nadine.guillotin@univ-lyon1.fr)

<sup>c</sup>Département de Mathématiques, Université Européenne de Bretagne, Université de Brest, 29238 Brest cedex, France. E-mail: [francoise.pene@univ-brest.fr](mailto:francoise.pene@univ-brest.fr)

Received 12 April 2011; revised 21 October 2011; accepted 17 November 2011

**Abstract.** Random walks in random scenery are processes defined by  $Z_n := \sum_{k=1}^n \xi_{X_1+\dots+X_k}$ , where  $(X_k, k \geq 1)$  and  $(\xi_y, y \in \mathbb{Z}^d)$  are two independent sequences of i.i.d. random variables with values in  $\mathbb{Z}^d$  and  $\mathbb{R}$  respectively. We suppose that the distributions of  $X_1$  and  $\xi_0$  belong to the normal basin of attraction of stable distribution of index  $\alpha \in (0, 2]$  and  $\beta \in (0, 2]$ . When  $d = 1$  and  $\alpha \neq 1$ , a functional limit theorem has been established in (*Z. Wahrsch. Verw. Gebiete* **50** (1979) 5–25) and a local limit theorem in (*Ann. Probab.* To appear). In this paper, we establish the convergence in distribution and a local limit theorem when  $\alpha = d$  (i.e.  $\alpha = d = 1$  or  $\alpha = d = 2$ ) and  $\beta \in (0, 2]$ . Let us mention that functional limit theorems have been established in (*Ann. Probab.* **17** (1989) 108–115) and recently in (*An asymptotic variance of the self-intersections of random walks. Preprint*) in the particular case when  $\beta = 2$  (respectively for  $\alpha = d = 2$  and  $\alpha = d = 1$ ).

**Résumé.** Les promenades aléatoires en paysage aléatoire sont des processus définis par  $Z_n := \sum_{k=1}^n \xi_{X_1+\dots+X_k}$ , où  $(X_k, k \geq 1)$  et  $(\xi_y, y \in \mathbb{Z}^d)$  sont deux suites indépendantes de variables aléatoires i.i.d. à valeurs dans  $\mathbb{Z}^d$  et  $\mathbb{R}$  respectivement. Nous supposons que les lois de  $X_1$  et  $\xi_0$  appartiennent au domaine d'attraction normal de lois stables d'indice  $\alpha \in (0, 2]$  et  $\beta \in (0, 2]$ . Quand  $d = 1$  et  $\alpha \neq 1$ , un théorème limite fonctionnel a été prouvé dans (*Z. Wahrsch. Verw. Gebiete* **50** (1979) 5–25) et un théorème limite local dans (*Ann. Probab.* To appear). Dans ce papier, nous prouvons la convergence en loi et un théorème limite local quand  $\alpha = d$  (i.e.  $\alpha = d = 1$  ou  $\alpha = d = 2$ ) et  $\beta \in (0, 2]$ . Mentionnons que des théorèmes limites fonctionnels ont été établis dans (*Ann. Probab.* **17** (1989) 108–115) et récemment dans (*An asymptotic variance of the self-intersections of random walks. Preprint*) dans le cas particulier où  $\beta = 2$  (respectivement pour  $\alpha = d = 2$  et  $\alpha = d = 1$ ).

MSC: 60F05; 60G52

Keywords: Random walk in random scenery; Local limit theorem; Local time; Stable process

## 1. Introduction

Random walks in random scenery (RWRS) are simple models of processes in disordered media with long-range correlations. They have been used in a wide variety of models in physics to study anomalous dispersion in layered random flows [20], diffusion with random sources, or spin depolarization in random fields (we refer the reader to Le Doussal's review paper [16] for a discussion of these models).

<sup>1</sup>This research was supported by the french ANR project MEMEMO2.

On the mathematical side, motivated by the construction of new self-similar processes with stationary increments, Kesten and Spitzer [15] and Borodin [4,5] introduced RWRS in dimension one and proved functional limit theorems. This study has been completed in many works, in particular in [3] and [10]. These processes are defined as follows. Let  $\xi := (\xi_y, y \in \mathbb{Z}^d)$  and  $X := (X_k, k \geq 1)$  be two independent sequences of independent identically distributed random variables taking values in  $\mathbb{R}$  and  $\mathbb{Z}^d$  respectively. The sequence  $\xi$  is called the *random scenery*. The sequence  $X$  is the sequence of increments of the *random walk*  $(S_n, n \geq 0)$  defined by  $S_0 := 0$  and  $S_n := \sum_{i=1}^n X_i$ , for  $n \geq 1$ . The *random walk in random scenery*  $Z$  is then defined by

$$Z_0 := 0 \quad \text{and} \quad \forall n \geq 1, \quad Z_n := \sum_{k=0}^{n-1} \xi_{S_k}.$$

Denoting by  $N_n(y)$  the local time of the random walk  $S$ :

$$N_n(y) := \#\{k = 0, \dots, n - 1 : S_k = y\},$$

it is straightforward to see that  $Z_n$  can be rewritten as  $Z_n = \sum_y \xi_y N_n(y)$ .

As in [15], the distribution of  $\xi_0$  is assumed to belong to the normal domain of attraction of a strictly stable distribution  $S_\beta$  of index  $\beta \in (0, 2]$ , with characteristic function  $\phi$  given by

$$\phi(u) = e^{-|u|^\beta (A_1 + iA_2 \operatorname{sgn}(u))}, \quad u \in \mathbb{R},$$

where  $0 < A_1 < \infty$  and  $|A_1^{-1} A_2| \leq |\tan(\pi\beta/2)|$ . We will denote by  $\varphi_\xi$  the characteristic function of the  $\xi_x$ 's. When  $\beta > 1$ , this implies that  $\mathbb{E}[\xi_0] = 0$ . When  $\beta = 1$ , we will further assume the symmetry condition

$$\sup_{t>0} |\mathbb{E}[\xi_0 \mathbb{1}_{\{|\xi_0| \leq t\}}]| < +\infty. \tag{1}$$

Under these conditions (for  $\beta \in (0; 2]$ ), there exists  $C_\xi > 0$  such that we have

$$\forall t > 0, \quad \mathbb{P}(|\xi_0| \geq t) \leq C_\xi t^{-\beta}. \tag{2}$$

Concerning the random walk, the distribution of  $X_1$  is assumed to belong to the normal basin of attraction of a stable distribution  $S'_\alpha$  with index  $\alpha \in (0, 2]$ .

Then the following weak convergences hold in the space of càdlàg real-valued functions defined on  $[0, \infty)$  and on  $\mathbb{R}$  respectively, endowed with the Skorohod  $J_1$ -topology (see [2], Chapter 3):

$$(n^{-1/\alpha} S_{[nt]})_{t \geq 0} \xrightarrow[n \rightarrow \infty]{\mathcal{L}} (U(t))_{t \geq 0}$$

and

$$\left( n^{-1/\beta} \sum_{k=0}^{\lfloor nx \rfloor} \xi_{ke_1} \right)_{x \geq 0} \xrightarrow[n \rightarrow \infty]{\mathcal{L}} (Y(x))_{x \geq 0}, \quad \text{with } e_1 = (1, 0, \dots, 0) \in \mathbb{Z}^d,$$

where  $U$  and  $Y$  are two independent Lévy processes such that  $U(0) = 0, Y(0) = 0, U(1)$  has distribution  $S'_\alpha, Y(1)$  has distribution  $S_\beta$ .

### 1.1. Functional limit theorem

Our first result is concerned with a limit theorem for  $(Z_{[nt]})_{t \geq 0}$ . Intuitively speaking,

- when  $\alpha < d$ , the random walk  $S_n$  is transient, its range is of order  $n$ , and  $Z_n$  has the same behaviour as a sum of about  $n$  independent random variables with the same distribution as the variables  $\xi_x$ . It was proved in [5] that for  $\beta = 2, n^{-1/\beta} (Z_{[nt]})_{t \geq 0}$  converges in distribution in the space  $D([0, \infty))$  of càdlàg functions endowed with

the Skorohod  $J_1$ -topology, to a multiple of the process  $(Y_t)$ . The case  $\beta \in (0, 2]$  was also mentioned in [15] (see Remark 3). When  $\beta < 1$  and the scenery is positive, a functional limit theorem in the space  $D([0, \infty))$  endowed with the Skorohod  $M_1$ -topology, is proved in [1] or [13];

- when  $\alpha > d$  (i.e.  $d = 1$  and  $1 < \alpha \leq 2$ ), the random walk  $S_n$  is recurrent, its range is of order  $n^{1/\alpha}$ , its local times are of order  $n^{1-1/\alpha}$ , so that  $Z_n$  is of order  $n^{1-1/\alpha+1/(\alpha\beta)}$ . In this situation, [4] and [15] proved a functional limit theorem for  $n^{-(1-1/\alpha+1/(\alpha\beta))}(Z_{[nt]}, t \geq 0)$  in the space  $\mathbb{C}([0, \infty))$  of continuous functions endowed with the uniform topology, the limiting process being a self-similar process, but not a stable one;
- when  $\alpha = d$  (i.e.  $\alpha = d = 1$  or  $\alpha = d = 2$ ),  $S_n$  is recurrent, its range is of order  $n/\log(n)$ , its local times are of order  $\log(n)$  so that  $Z_n$  is of order  $n^{1/\beta} \log(n)^{(\beta-1)/\beta}$ . In this situation, a functional limit theorem in the space of continuous functions was proved in [3] for  $d = \alpha = \beta = 2$ , and in [10] for  $d = \alpha = 1$  and  $\beta = 2$ .

Our first result gives a limit theorem for  $\alpha = d$  and for any value of  $\beta \in (0; 2)$ . We establish the convergence in the sense of finite distributions, and prove that the convergence in distribution does not hold for the  $J_1$ -topology when  $\beta \neq 2$  but that the convergence in distribution holds for the  $M_1$ -topology when  $\beta \neq 1$  (for technical reasons, our proof does not apply when  $\beta = 1$ ).

**Theorem 1.** *Let  $\beta \in (0; 2)$ . We assume that the random walk is strongly aperiodic and that*

- (a) *either  $d = 2$  and  $X_1$  is centered, square integrable with invertible variance matrix  $\Sigma$  and then we define  $A := 2\sqrt{\det \Sigma}$ ;*
- (b) *or  $d = 1$  and  $(\frac{S_n}{n})_n$  converges in distribution to a random variable with characteristic function given by  $t \mapsto \exp(-a|t|)$  with  $a > 0$  and then we define  $A := a$ .*

*Then, the sequence of random variables*

$$\left( \left( \frac{Z_{[nt]}}{n^{1/\beta} \log(n)^{(\beta-1)/\beta}} \right)_{t \in [0,1]} \right)_{n \geq 2}$$

*converges in the sense of finite distributions to the process*

$$\left( \tilde{Y}_t := \left( \frac{\Gamma(\beta + 1)}{(\pi A)^{\beta-1}} \right)^{1/\beta} Y(t) \right)_{t \in [0,1]}.$$

*For  $\beta < 2$ , the convergence does not hold in  $\mathcal{D}([0, 1])$  endowed with the  $J_1$ -topology, but when  $\beta \neq 1$ , the convergence holds in  $\mathcal{D}([0, 1])$  endowed with the  $M_1$ -topology.*

**Remark 2.** *For  $d > \alpha$  and  $\beta \neq 1$ , the same proof as in Theorem 1 shows that the sequence  $(n^{-1/\beta} Z_{[nt]}, t \in [0, 1])$  converges in  $(\mathcal{D}([0, 1], M_1)$  to the process  $(\mathbb{E}[N_\infty^{\beta-1}]^{1/\beta} Y(t), t \in [0, 1])$ , where  $N_\infty$  is the total number of visits to 0 of a two-sided random walk  $(S_n, n \in \mathbb{Z})$  such that  $S_0 = 0$  and whose increments are distributed according to  $X_1$  (see Remarks 6, 8, 9, 11 below).*

### 1.2. Local limit theorem

Our next results concern a local limit theorem for  $(Z_n)_n$ . The  $d = 1$  case was treated in [7] for  $\alpha \in (0; 2] \setminus \{1\}$  and all values of  $\beta \in (0; 2]$ . Here, we complete this study by proving a local limit theorem for  $\alpha = d = 1$  (and  $\beta \in (0; 2]$ ). By a direct adaptation of the proof of this result, we also establish a local limit theorem for  $\alpha = d = 2$  (we just adapt the definition of “peaks,” see Section 3.5). Let us notice that the same adaptation can be done from [7] (case  $\alpha < 1$ ) to get local limit theorems for  $d \geq 2, \alpha < d$  and  $\beta \in (0; 2]$ .

We give two results corresponding respectively to the case when  $\xi_0$  is lattice and to the case when it is strongly nonlattice. We denote by  $\varphi_\xi$  the characteristic function of  $\xi_0$ .

**Theorem 3.** *Assume that  $\xi_0$  takes its values in  $\mathbb{Z}$ . Let  $d_0 \geq 1$  be the integer such that  $\{u: |\varphi_\xi(u)| = 1\} = \frac{2\pi}{d_0} \mathbb{Z}$ . Let  $b_n := n^{1/\beta} (\log(n))^{(\beta-1)/\beta}$ . Under the previous assumptions on the random walk and on the scenery, for  $\alpha = d \in \{1, 2\}$ , for every  $\beta \in (0, 2]$ , and for every  $x \in \mathbb{R}$ ,*

- if  $\mathbb{P}(n\xi_0 - \lfloor b_n x \rfloor \notin d_0\mathbb{Z}) = 1$ , then  $\mathbb{P}(Z_n = \lfloor b_n x \rfloor) = 0$ ;
- if  $\mathbb{P}(n\xi_0 - \lfloor b_n x \rfloor \in d_0\mathbb{Z}) = 1$ , then

$$\mathbb{P}(Z_n = \lfloor b_n x \rfloor) = d_0 \frac{C(x)}{b_n} + o(b_n^{-1})$$

uniformly in  $x \in \mathbb{R}$ , where  $C(\cdot)$  is the density function of  $\tilde{Y}_1$ .

When  $\xi_0$  is strongly nonlattice, we establish the weak convergence of  $b_n \mathbb{P}_{Z_n}$  to the Lebesgue measure on  $\mathbb{R}$  (in the sense of compact supported function, see Definition 10.2 of [6]). More precisely we state the following result.

**Theorem 4.** Assume now that  $\xi_0$  is strongly nonlattice which means that

$$\limsup_{|u| \rightarrow +\infty} |\varphi_\xi(u)| < 1.$$

We still assume that  $\alpha = d \in \{1, 2\}$  and  $\beta \in (0, 2]$ . Then, for every compactly supported continuous function  $g : \mathbb{R} \rightarrow \mathbb{C}$ , we have

$$\lim_{n \rightarrow +\infty} \sup_{x \in \mathbb{R}} \left| b_n \mathbb{E}[g(Z_n - b_n x)] - C(x) \int_{\mathbb{R}} g(t) dt \right| = 0,$$

with  $b_n := n^{1/\beta} (\log(n))^{(\beta-1)/\beta}$  and where  $C(\cdot)$  is the density function of  $\tilde{Y}_1$ .

## 2. Proof of the functional limit theorem

Before proving the theorem, we prove some technical lemmas. For any real number  $\gamma > 0$ , any integer  $m \geq 1$ , any  $\theta_1, \dots, \theta_m \in \mathbb{R}$ , any  $t_0 = 0 < t_1 < \dots < t_m$ , we consider the sequences of random variables  $(L_n(\gamma))_{n \geq 2}$  and  $(L'_n(\gamma))_{n \geq 2}$  defined by

$$L_n(\gamma) := \frac{1}{n(\log n)^{\gamma-1}} \sum_{x \in \mathbb{Z}^d} \left| \sum_{i=1}^m \theta_i (N_{\lfloor nt_i \rfloor}(x) - N_{\lfloor nt_{i-1} \rfloor}(x)) \right|^\gamma$$

and

$$L'_n(\gamma) := \frac{1}{n(\log n)^{\gamma-1}} \sum_{x \in \mathbb{Z}^d} \left| \sum_{i=1}^m \theta_i (N_{\lfloor nt_i \rfloor}(x) - N_{\lfloor nt_{i-1} \rfloor}(x)) \right|^\gamma \operatorname{sgn} \left( \sum_{i=1}^m \theta_i (N_{\lfloor nt_i \rfloor}(x) - N_{\lfloor nt_{i-1} \rfloor}(x)) \right).$$

**Lemma 5.** For any real number  $\gamma > 0$ , any integer  $m \geq 1$ , any  $\theta_1, \dots, \theta_m \in \mathbb{R}$ , any  $t_0 = 0 < t_1 < \dots < t_m$ , the following convergences hold  $\mathbb{P}$ -almost surely

$$\lim_{n \rightarrow +\infty} L_n(\gamma) = \frac{\Gamma(\gamma + 1)}{(\pi A)^{\gamma-1}} \sum_{i=1}^m |\theta_i|^\gamma (t_i - t_{i-1}) \tag{3}$$

and

$$\lim_{n \rightarrow +\infty} L'_n(\gamma) = \frac{\Gamma(\gamma + 1)}{(\pi A)^{\gamma-1}} \sum_{i=1}^m |\theta_i|^\gamma \operatorname{sgn}(\theta_i) (t_i - t_{i-1}). \tag{4}$$

**Proof.** We fix an integer  $m \geq 1$  and  $2m$  real numbers  $\theta_1, \dots, \theta_m, t_1, \dots, t_m$  such that  $0 < t_1 < \dots < t_m$  and we set  $t_0 := 0$ . To simplify notations, we write  $d_{i,n}(x) := N_{[nt_i]}(x) - N_{[nt_{i-1}]}(x)$ . Following the techniques developed in [8], we first have to prove (3) and (4) for integer  $\gamma$ : for every integer  $k \geq 1$ ,  $\mathbb{P}$ -almost surely, as  $n$  goes to infinity, we have

$$\frac{1}{n(\log n)^{k-1}} \sum_{x \in \mathbb{Z}^d} \left( \sum_{i=1}^m \theta_i d_{i,n}(x) \right)^k \longrightarrow \frac{\Gamma(k+1)}{(\pi A)^{k-1}} \sum_{i=1}^m \theta_i^k (t_i - t_{i-1}). \quad (5)$$

Let us assume (5) for a while, and let us end the proof of (3) and (4) for any positive real  $\gamma$ . Given the random walk  $S := (S_n)_n$ , let  $(U_n)_{n \geq 1}$  be a sequence of random variables with values in  $\mathbb{Z}^d$ , such that for all  $n$ ,  $U_n$  is a point chosen uniformly in the range of the random walk up to time  $[nt_m]$ , that is

$$\mathbb{P}(U_n = x | S) = R_{[nt_m]}^{-1} \mathbf{1}_{[N_{[nt_m]}(x) \geq 1]},$$

with  $R_k := \#\{y: N_k(y) > 0\}$ . Moreover, let  $U'$  be a random variable with values in  $\{1, \dots, m\}$  and distribution

$$\mathbb{P}(U' = i) = (t_i - t_{i-1})/t_m$$

and let  $T$  be a random variable with exponential distribution with parameter one and independent of  $U'$ .

Then, for  $\mathbb{P}$ -almost every realization of the random walk  $S$ , the sequence of random variables

$$\left( W_n := \frac{\pi A}{\log(n)} \sum_{i=1}^m \theta_i d_{i,n}(U_n) \right)_n$$

converges in distribution to the random variable  $W := \theta_{U'} T$ . Indeed, the moment of order  $k$  of  $W_n$  given  $S$  is

$$\mathbb{E}(W_n^k | S) = \frac{(\pi A)^k}{n(\log n)^{k-1}} \sum_{x \in \mathbb{Z}^d} \left( \sum_{i=1}^m \theta_i d_{i,n}(x) \right)^k \frac{n}{\log(n) R([nt_m])}.$$

Using (5) and the fact that  $((\log n)R_n/n)_n$  converges almost surely to  $\pi A$  (see [11,17]), the moments  $\mathbb{E}(W_n^k | S)$  converges a.s. to  $\mathbb{E}(W^k) = \Gamma(k+1) \sum_{i=1}^m \theta_i^k (t_i - t_{i-1})/t_m$ . This proves the convergence of the conditional distribution of  $(W_n)_n$  given  $S$  to  $W$ , since the distribution of  $W$  is identified by its moments (thanks to the Carleman condition). This ensures, in particular, the convergence in distribution of  $(|W_n|^\gamma)_n$  and of  $(|W_n|^\gamma \operatorname{sgn}(W_n))_n$  (given  $S$ ) to  $|W|^\gamma$  and  $|W|^\gamma \operatorname{sgn}(W)$  respectively (for every real number  $\gamma \geq 0$  and for  $\mathbb{P}$ -almost every realization of the random walk  $S$ ). Since, conditional on  $S$ , any moment of  $|W_n|$  can be bounded from above by an integer moment, we deduce that, for any  $\gamma \geq 0$ , we have  $\mathbb{P}$ -almost surely

$$\lim_{n \rightarrow +\infty} \mathbb{E}(|W_n|^\gamma | S) = \mathbb{E}(|W|^\gamma) \quad \text{and} \quad \lim_{n \rightarrow +\infty} \mathbb{E}(|W_n|^\gamma \operatorname{sgn}(W_n) | S) = \mathbb{E}(|W|^\gamma \operatorname{sgn}(W)),$$

which proves Lemma 5.

Let us prove (5). Let  $k \geq 1$ . According to Theorem 1 in [8] (proved for  $\alpha = d = 2$ , but also valid for  $\alpha = d = 1$ ; see [Appendix](#) for additional comments on the proof of this theorem), we have

$$\forall i \in \{1, \dots, m\}, \quad \lim_{n \rightarrow +\infty} \frac{1}{n(\log n)^{k-1}} \sum_{x \in \mathbb{Z}^d} (d_{i,n}(x))^k = \frac{\Gamma(k+1)}{(\pi A)^{k-1}} (t_i - t_{i-1}), \quad \mathbb{P}\text{-a.s.} \quad (6)$$

We define

$$\Sigma_n(\theta_1, \dots, \theta_m) := \sum_{x \in \mathbb{Z}^d} \left( \sum_{i=1}^m \theta_i d_{i,n}(x) \right)^k - \sum_{x \in \mathbb{Z}^d} \sum_{i=1}^m (\theta_i)^k (d_{i,n}(x))^k. \quad (7)$$

According to (6), it is enough to prove that  $\mathbb{P}$ -a.s.,  $\Sigma_n(\theta_1, \dots, \theta_m) = o(n(\log n)^{k-1})$ . We observe that  $\Sigma_n(\theta_1, \dots, \theta_m)$  is the sum of the following terms

$$\sum_{x \in \mathbb{Z}^d} \prod_{j=1}^k (\theta_{i_j} d_{i_j, n}(x)) \quad (8)$$

over all the  $k$ -tuple  $(i_1, \dots, i_k) \in \{1, \dots, m\}^k$ , with at least two distinct indices. We observe that

$$|\Sigma_n(\theta_1, \dots, \theta_m)| \leq \max(|\theta_1|, \dots, |\theta_m|)^k \Sigma_n(1, \dots, 1).$$

But, we have

$$\begin{aligned} \Sigma_n(1, \dots, 1) &= \sum_{x \in \mathbb{Z}^d} (N_{[nt_m]}(x))^k - \sum_{x \in \mathbb{Z}^d} \sum_{i=1}^m (d_{i, n}(x))^k \\ &= \sum_{x \in \mathbb{Z}^d} (N_{[nt_m]}(x))^k - \sum_{i=1}^m \sum_{x \in \mathbb{Z}^d} (d_{i, n}(x))^k = o(n \log(n)^{k-1}), \end{aligned}$$

according to (6). □

**Remark 6.** Case  $d > \alpha$ .

In this case,  $R_n/n$  converges a.s. to  $p = \mathbb{P}[S_k \neq 0 \text{ for any } k \geq 1]$  (cf. [21]), and for all real number  $k \geq 0$ ,  $\frac{1}{n} \sum_{x \in \mathbb{Z}^d} N_n^k(x)$  converges a.s. to  $\mathbb{E}[N_\infty^{k-1}]$  (see Remark 2 for a definition of  $N_\infty$  and the introduction of [15] for a proof of this fact). Setting  $W_n = \sum_{j=1}^m \theta_j d_{j, n}(U_n)$ , it follows that for all integer  $k \geq 1$   $\mathbb{E}[W_n^k | S]$  tends to  $\mathbb{E}_{\mathbb{Q}}[(\theta_U, N_\infty)^k]$ , where  $\mathbb{Q}$  is the probability on the random walk's paths space, whose density w.r.t. the random walk's law  $\mathbb{P}$  is given by  $d\mathbb{Q}/d\mathbb{P} = 1/(pN_\infty)$ . This leads to the following two facts: for any real number  $\gamma > 0$ , any integer  $m \geq 1$ , any  $\theta_1, \dots, \theta_m \in \mathbb{R}$ , any  $t_0 = 0 < t_1 < \dots < t_m$ , the following convergences hold  $\mathbb{P}$ -almost surely

$$\lim_{n \rightarrow +\infty} \frac{1}{n} \sum_{x \in \mathbb{Z}^d} \left| \sum_{i=1}^m \theta_i (N_{[nt_i]} - N_{[nt_{i-1}]}) \right|^\gamma = \mathbb{E}[N_\infty^{\gamma-1}] \sum_{i=1}^m |\theta_i|^\gamma (t_i - t_{i-1}) \quad (9)$$

and

$$\begin{aligned} \lim_{n \rightarrow +\infty} \frac{1}{n} \sum_{x \in \mathbb{Z}^d} \left| \sum_{i=1}^m \theta_i (N_{[nt_i]} - N_{[nt_{i-1}]}) \right|^\gamma \operatorname{sgn} \left( \sum_{i=1}^m \theta_i (N_{[nt_i]} - N_{[nt_{i-1}]}) \right) \\ = \mathbb{E}[N_\infty^{\gamma-1}] \sum_{i=1}^m |\theta_i|^\gamma \operatorname{sgn}(\theta_i) (t_i - t_{i-1}). \end{aligned} \quad (10)$$

**Lemma 7.** For any  $\rho > 0$ ,

$$\sup_{x \in \mathbb{Z}^d} N_n(x) = o(n^\rho) \quad \text{a.s.}$$

**Proof.** See Lemma 2.5 in [3]. □

**Proof of Theorem 1.** Convergence of the finite-dimensional distributions.

Let an integer  $m \geq 1$  and  $2m$  real numbers  $\theta_1, \dots, \theta_m, t_1, \dots, t_m$  such that  $0 < t_1 < \dots < t_m \leq 1$ . We set  $t_0 := 0$ . Again, we use the notation  $d_{i, n}(x) := N_{[nt_i]}(x) - N_{[nt_{i-1}]}(x)$ , and set

$$b_n = n^{1/\beta} (\log(n))^{(\beta-1)/\beta}, \quad \bar{Z}_n := \frac{1}{b_n} \sum_{i=1}^m \theta_i (Z_{[nt_i]} - Z_{[nt_{i-1}]}).$$

We have to prove that

$$\mathbb{E}[e^{i\bar{Z}_n}] \rightarrow \prod_{i=1}^m \phi\left(\theta_i(t_i - t_{i-1})^{1/\beta} \left(\frac{\Gamma(\beta + 1)}{(\pi A)^{\beta-1}}\right)^{1/\beta}\right), \quad (11)$$

as  $n$  goes to infinity. We observe that  $\bar{Z}_n = \frac{1}{b_n} \sum_{x \in \mathbb{Z}^d} \sum_{i=1}^m \theta_i d_{i,n}(x) \xi_x$ . Hence we have

$$\mathbb{E}[e^{i\bar{Z}_n} | S] = \prod_{x \in \mathbb{Z}^d} \varphi_\xi\left(\frac{\sum_{i=1}^m \theta_i d_{i,n}(x)}{b_n}\right).$$

Observe next that

$$|\varphi_\xi(t) - \exp(-|t|^\beta (A_1 + iA_2 \operatorname{sgn}(t)))| \leq |t|^\beta h(|t|) \quad \text{for all } t \in \mathbb{R},$$

with  $h$  a continuous and monotone function on  $[0, +\infty)$  vanishing in 0. According to Lemma 7,  $\mathbb{P}$ -almost surely, for every  $n$  large enough, we have

$$D_n := \sup_x \frac{|\sum_{i=1}^m \theta_i d_{i,n}(x)|}{b_n} \leq m \max(|\theta_i|) \frac{\sup_x N_n(x)}{b_n} \leq \varepsilon_0$$

and so

$$\left| \mathbb{E}[e^{i\bar{Z}_n} | S] - \prod_{x \in \mathbb{Z}^d} e^{-(|\sum_{i=1}^m \theta_i d_{i,n}(x)|^\beta / b_n^\beta)(A_1 + iA_2 \operatorname{sgn}(\sum_{i=1}^m \theta_i d_{i,n}(x)))} \right|$$

is less than  $\sum_{x \in \mathbb{Z}^d} \frac{|\sum_{i=1}^m \theta_i d_{i,n}(x)|^\beta}{b_n^\beta} h(B_n)$ . Hence, according to Lemmas 5 and 7,  $\mathbb{P}$ -almost surely, we have

$$\lim_{n \rightarrow +\infty} \mathbb{E}[e^{i\bar{Z}_n} | S] = e^{-(\Gamma(\beta+1)/(\pi A)^{\beta-1}) \sum_{i=1}^m |\theta_i|^\beta (t_i - t_{i-1})(A_1 + iA_2 \operatorname{sgn}(\theta_i))}$$

which gives (11) thanks to the Lebesgue dominated convergence theorem.

**Remark 8.** Case  $d > \alpha$ .

The proof is exactly the same with  $b_n = n^{1/\beta}$ .

*Study of the tightness.*

When  $\beta = 2$ , the sequence is known to be tight for the  $J_1$  (so also  $M_1$ ) topology (see [3]). For  $\beta < 2$ , we prove that the sequence  $(\frac{Z_{[nt]}}{b_n})_{t \in [0,1]}$  is not tight in  $(\mathcal{D}([0,1]), J_1)$ . To this aim, let  $(Z_n(t), t \in [0,1])$  denote the linear interpolation of  $(Z_{[nt]}, t \in [0,1])$ , i.e.

$$Z_n(t) = Z_{[nt]} + (nt - [nt])\xi_{S_{[nt]}}.$$

Then,  $\forall \epsilon > 0$ ,

$$\begin{aligned} \mathbb{P}\left[\sup_{t \in [0,1]} |Z_n(t) - Z_{[nt]}| \geq \epsilon b_n\right] &= \mathbb{P}\left[\max_{i=0}^{n-1} |\xi_{S_i}| \geq \epsilon b_n\right] \\ &= \mathbb{P}[\exists x \in \{S_0, \dots, S_{n-1}\} \text{ s.t. } |\xi_x| \geq \epsilon b_n] \\ &\leq \mathbb{E}[\#\{S_0, \dots, S_{n-1}\}] \mathbb{P}[|\xi_0| \geq \epsilon b_n] \\ &\leq C \frac{n}{\log(n)} \epsilon^{-\beta} b_n^{-\beta} = C \epsilon^{-\beta} \log(n)^{-\beta}, \end{aligned}$$

where the last inequality comes from (2) and Theorem 6.9 of [17]. Therefore, if  $((\frac{Z_{[nt]}}{b_n})_{t \in [0;1]})_{n \geq 2}$  converges in distribution in  $(\mathcal{D}([0, 1]), J_1)$  to  $(\tilde{Y}_t)_{t \in [0,1]}$ , the same is true for  $((\frac{Z_n(t)}{b_n})_{t \in [0;1]})_{n \geq 2}$  which implies that  $(\frac{Z_n(t)}{b_n})_{t \in [0;1]}$  converges in distribution in  $\mathcal{C}([0, 1])$ , and that the limiting process  $(\tilde{Y}_t)_{t \in [0,1]}$  is therefore continuous, which is false as soon as  $\beta < 2$ .

*M<sub>1</sub>-tightness for  $\beta > 1$ .*

Set  $\tilde{Z}_n(t) = \frac{Z_{[nt]}}{b_n}$ , and let us prove the tightness of the sequence  $(\tilde{Z}_n)_n$  in  $\mathcal{D}([0, 1])$  for the  $M_1$ -topology when  $\beta > 1$ . For any  $y_1, y_2$  and  $y_3$  real, let us denote  $\|y_2 - [y_1, y_3]\| = \inf_{t \in [y_1, y_3]} |y_2 - t|$ . For any function  $z = (z(t))_{t \in [0,1]}$  in  $\mathcal{D}([0, 1])$ , we define

$$\omega(z, \delta) = \sup_{t \in [0,1]} \sup \{ \|z(t_2) - [z(t_1), z(t_3)]\| : (t - \delta) \vee 0 \leq t_1 < t_2 < t_3 \leq (t + \delta) \wedge 1 \}.$$

From Skorohod criteria (see [22] or [23], Chapter 12) it is enough to prove that for every  $\varepsilon > 0$ ,

$$\lim_{k \rightarrow +\infty} \limsup_{n \rightarrow +\infty} \mathbb{P} \left[ \omega \left( \tilde{Z}_n, \frac{1}{k} \right) > \varepsilon \right] = 0. \quad (12)$$

The proof is based on two distinct results: the first one by Louhichi and Rio in [19] where they prove that in the case of a sum of associated random variables, the above  $M_1$ -tightness criteria can be deduced from a maximal inequality for the sum; the second one by Louhichi in [18] where a maximal inequality for the sum of associated random variables without moment conditions (not necessarily stationary) is proved. Let us give the details. Since the sequence  $(\xi_{S_k})_{k \geq 0}$  is stationary, we have for every  $k \geq 3$ ,

$$\mathbb{P} \left[ \omega \left( \tilde{Z}_n, \frac{1}{k} \right) > \varepsilon \right] \leq (k-2) \mathbb{P} \left[ \sup_{0 \leq n_1 < n_2 < n_3 \leq 1 + \lfloor 3n/k \rfloor} \|Z_{n_2} - [Z_{n_1}, Z_{n_3}]\| > \varepsilon b_n \right].$$

Conditionally to the random walk  $S = (S_n)_{n \geq 0}$ , the sequence of random variables  $(\xi_{S_k})_{k \geq 0}$  is associated, therefore by applying inequality (3) in [19], we have

$$\mathbb{P} \left[ \omega \left( \tilde{Z}_n, \frac{1}{k} \right) > \varepsilon \right] \leq (k-2) \mathbb{E} \left[ \mathbb{P} \left( \max_{0 \leq j \leq 1 + \lfloor 3n/k \rfloor} |Z_j| > \frac{\varepsilon b_n}{2} \mid S \right)^2 \right]. \quad (13)$$

Now let us apply Lemma 1 in [18] to the random variables  $X = |\xi_0|$  and  $X_i = \xi_{S_i}$ ,  $i \geq 0$ , conditionally to the random walk. For any sequence of positive reals  $(\tilde{b}_n)_n$ , there exist some constant  $C > 0$  depending on  $\varepsilon$  (the value of  $C$  may change from line to line in the following inequalities) s.t.

$$\begin{aligned} \mathbb{P} \left( \max_{0 \leq j \leq 1 + \lfloor 3n/k \rfloor} Z_j > \frac{\varepsilon b_n}{2} \mid S \right) &\leq C \left\{ \frac{(1 + \lfloor 3n/k \rfloor) \mathbb{E}[\xi_0^2 \mathbf{1}_{\{|\xi_0| \leq \tilde{b}_n\}}]}{b_n^2} + \frac{(1 + \lfloor 3n/k \rfloor) \mathbb{E}[|\xi_0| \mathbf{1}_{\{|\xi_0| > \tilde{b}_n\}}]}{b_n} \right. \\ &\quad \left. + \left( 1 + \left\lfloor \frac{3n}{k} \right\rfloor \right) \left( \frac{\tilde{b}_n}{b_n} \right)^2 \mathbb{P}[|\xi_0| > \tilde{b}_n] + \frac{1}{b_n^2} \sum_{0 \leq i < j \leq 1 + \lfloor 3n/k \rfloor} G_{ij}(\tilde{b}_n) \right\}, \end{aligned}$$

where, in our setting, if we denote for  $v \in \mathbb{R}_+$  by  $g_v$  the function  $(u \wedge v) \vee (-v)$ ,

$$\begin{aligned} G_{ij}(v) &:= \mathbb{E}[g_v(\xi_{S_i}) g_v(\xi_{S_j}) \mid S] - \mathbb{E}[g_v(\xi_{S_i}) \mid S] \mathbb{E}[g_v(\xi_{S_j}) \mid S] \\ &\leq \mathbb{E}[g_v(\xi_{S_i})^2 \mid S] \mathbf{1}_{\{S_i = S_j\}} = \mathbb{E}[g_v(\xi_0)^2 \mid S] \mathbf{1}_{\{S_i = S_j\}}. \end{aligned}$$

The same reasoning holds for the sequence  $(-\xi_{S_i})_{i \geq 0}$ , which is also associated, then since the function  $g_v$  is odd, we deduce, by denoting  $I_n := \sum_{i,j=0}^{n-1} \mathbf{1}_{\{S_i = S_j\}}$ , the following maximal inequality

$$\begin{aligned} \mathbb{P} \left[ \max_{0 \leq j \leq 1 + \lfloor 3n/k \rfloor} |Z_j| > \frac{\varepsilon b_n}{2} \mid S \right] &\leq C \left\{ \frac{(1 + \lfloor 3n/k \rfloor) \mathbb{E}[\xi_0^2 \mathbf{1}_{\{|\xi_0| \leq \tilde{b}_n\}}]}{b_n^2} + \frac{(1 + \lfloor 3n/k \rfloor) \mathbb{E}[|\xi_0| \mathbf{1}_{\{|\xi_0| > \tilde{b}_n\}}]}{b_n} \right. \\ &\quad \left. + \left( 1 + \left\lfloor \frac{3n}{k} \right\rfloor \right) \left( \frac{\tilde{b}_n}{b_n} \right)^2 \mathbb{P}[|\xi_0| > \tilde{b}_n] + \frac{I_{1 + \lfloor 3n/k \rfloor}}{b_n^2} \mathbb{E}[g_{\tilde{b}_n}(\xi_0)^2] \right\}. \end{aligned}$$



Since for every  $x, y \in \mathbb{R}^+$ ,  $(x + y)^2 \leq 2(x^2 + y^2)$ , we get

$$\mathbb{E} \left[ \mathbb{P} \left( \max_{0 \leq j \leq 1 + \lfloor 3n/k \rfloor} |Z_j| > \frac{\varepsilon b_n}{2} \middle| S \right)^2 \right] \leq C \sum_{i=1}^4 \Sigma_i(n, k), \tag{14}$$

where

$$\begin{aligned} \Sigma_1(n, k) &= \frac{(1 + \lfloor 3n/k \rfloor)^2}{b_n^4} \mathbb{E} [\xi_0^2 \mathbf{1}_{\{|\xi_0| \leq \tilde{b}_n\}}]^2, \\ \Sigma_2(n, k) &= \frac{(1 + \lfloor 3n/k \rfloor)^2}{b_n^2} \mathbb{E} [|\xi_0| \mathbf{1}_{\{|\xi_0| > \tilde{b}_n\}}]^2, \\ \Sigma_3(n, k) &= \left( 1 + \left\lfloor \frac{3n}{k} \right\rfloor \right)^2 \left( \frac{\tilde{b}_n}{b_n} \right)^4 \mathbb{P} [|\xi_0| > \tilde{b}_n]^2, \\ \Sigma_4(n, k) &= \frac{\mathbb{E} (I_{1 + \lfloor 3n/k \rfloor}^2)}{b_n^4} \mathbb{E} [g_{\tilde{b}_n}(\xi_0)^2]^2. \end{aligned}$$

Note that  $\mathbb{E} [\xi_0^2 \mathbf{1}_{\{|\xi_0| \leq \tilde{b}_n\}}] \asymp \tilde{b}_n^{2-\beta}$ ,  $\mathbb{E} [|\xi_0| \mathbf{1}_{\{|\xi_0| > \tilde{b}_n\}}] \asymp \tilde{b}_n^{1-\beta}$  for  $\beta < 1$ , and  $\mathbb{E} [g_{\tilde{b}_n}(\xi_0)^2] \asymp \tilde{b}_n^{2-\beta}$ . Therefore, by choosing  $\tilde{b}_n = (\frac{n}{\log(n)})^{1/\beta}$ , we deduce that for  $i = 1, 3$ ,

$$\limsup_{n \rightarrow +\infty} \Sigma_i(n, k) = 0, \tag{15}$$

and (recall that  $\mathbb{E} (I_n^2) = \mathcal{O}((n \log(n))^2)$ ) for  $i = 2, 4$ , there exist two constants  $C_i > 0$  s.t.

$$\limsup_{n \rightarrow +\infty} \Sigma_i(n, k) \leq \frac{C_i}{k^2}. \tag{16}$$

Therefore, by combining (13), (14), (15) and (16), there exists some constant  $C > 0$  s.t.

$$\limsup_{n \rightarrow +\infty} \mathbb{P} \left[ \omega \left( \tilde{Z}_n, \frac{1}{k} \right) > \varepsilon \right] \leq \frac{C}{k}$$

then (12) follows.

**Remark 9.** Case  $d > \alpha$  and  $\beta > 1$ .

It is easy to see that for  $d > \alpha$ ,  $\mathbb{E} (I_n^2) = \mathcal{O}(n^2)$ . Taking  $\tilde{b}_n = b_n = n^{1/\beta}$ , the same proof leads to  $\limsup_{n \rightarrow \infty} \Sigma_i(n, k) \leq C_i/k^2$  for every  $i \in \{1, \dots, 4\}$ , and to the tightness in  $M_1$ -topology.

*$M_1$ -tightness for  $\beta < 1$ .*

For  $\beta < 1$ , to get a control of the oscillation, we write  $\xi_x = \xi_x^+ - \xi_x^-$  to obtain the decomposition  $\tilde{Z}_n = \tilde{Z}_n^+ - \tilde{Z}_n^-$ , where  $\tilde{Z}_n^+(t) := \frac{1}{b_n} Z_{[nt]}^+$ , and  $Z_n^+$  is the random walk in the random scenery  $(\xi_x^+, x \in \mathbb{Z}^d)$ :

$$Z_n^+ = \sum_{k=0}^{n-1} \xi_{S_k}^+ = \sum_{x \in \mathbb{Z}^d} \xi_x^+ N_n(x).$$

$\tilde{Z}_n^-$  is defined in the same way as  $\tilde{Z}_n^+$  using the negative part of the scenery. Since the processes  $\tilde{Z}_n^-, \tilde{Z}_n^+$  are increasing, for any  $\delta > 0$ ,  $\omega(\tilde{Z}_n^-, \delta) = \omega(\tilde{Z}_n^+, \delta) = 0$ . Assume for a while that  $\tilde{Z}_n^-(1)$  and  $\tilde{Z}_n^+(1)$  both converge in distribution (this is false for  $\beta \geq 1$  due to centering term). It follows that the processes  $\tilde{Z}_n^-$  and  $\tilde{Z}_n^+$  are tight in  $M_1$ -topology. To get the tightness of their difference  $\tilde{Z}_n$ , we have then to prove that the limiting processes of  $\tilde{Z}_n^-$  and  $\tilde{Z}_n^+$  do not have common discontinuity points (see Corollary 12.7.1 in [23]). This is the case if these two processes are independent. Therefore, all that remains to prove is the following lemma.  $\square$

**Lemma 10.** *Let an integer  $m \geq 1$  and  $3m$  real numbers  $\theta_1, \dots, \theta_m, \gamma_1, \dots, \gamma_m, t_1, \dots, t_m$  such that  $0 < t_1 < \dots < t_m \leq 1$ . We set  $t_0 := 0$ . Then*

$$\begin{aligned} & \lim_{n \rightarrow \infty} \mathbb{E} \left[ \exp \left( i \sum_{j=1}^m (\theta_j (\tilde{Z}_n^+(t_j) - \tilde{Z}_n^+(t_{j-1})) + \gamma_j (\tilde{Z}_n^-(t_j) - \tilde{Z}_n^-(t_{j-1}))) \right) \right] \\ &= \prod_{j=1}^m \phi_1(\theta_j(t_j - t_{j-1})^{1/\beta}) \phi_2(\gamma_j(t_j - t_{j-1})^{1/\beta}), \end{aligned}$$

where  $\phi_1$  and  $\phi_2$  are characteristic functions of positive  $\beta$ -stable laws.

**Proof.** We use the notation

$$d_{i,n}(x) := N_{[nt_i]}(x) - N_{[nt_{i-1}]}(x), \quad d_n(x) := (d_{1,n}(x), \dots, d_{m,n}(x)).$$

Observe that

$$\sum_{j=1}^m (\theta_j (\tilde{Z}_n^+(t_j) - \tilde{Z}_n^+(t_{j-1})) + \gamma_j (\tilde{Z}_n^-(t_j) - \tilde{Z}_n^-(t_{j-1}))) = \frac{1}{b_n} \sum_{x \in \mathbb{Z}^d} \xi_x^+(\theta; d_n(x)) + \xi_x^-(\gamma; d_n(x)).$$

Therefore

$$\begin{aligned} & \mathbb{E} \left[ \exp \left( i \sum_{j=1}^m (\theta_j (\tilde{Z}_n^+(t_j) - \tilde{Z}_n^+(t_{j-1})) + \gamma_j (\tilde{Z}_n^-(t_j) - \tilde{Z}_n^-(t_{j-1}))) \right) \middle| S \right] \\ &= \prod_{x \in \mathbb{Z}^d} \mathbb{E} \left[ \exp \left( i \left( \xi_x^+ \frac{\langle \theta; d_n(x) \rangle}{b_n} + \xi_x^- \frac{\langle \gamma; d_n(x) \rangle}{b_n} \right) \right) \middle| S \right]. \end{aligned}$$

Note that for any real  $s, t$ ,  $\mathbb{E}[\exp(i(t\xi_0^+ + s\xi_0^-))] = \varphi_{\xi^+}(t) + \varphi_{\xi^-}(s) - 1$ . Since  $\xi$  is in the domain of attraction of  $\mathcal{S}_\beta$ , the tails of the variables  $\xi^+$  and  $\xi^-$  satisfy  $\mathbb{P}[\xi^+ \geq t] \asymp p\mathbb{P}[|\xi| \geq t]$ ,  $\mathbb{P}[\xi^- \geq t] \asymp (1-p)\mathbb{P}[|\xi| \geq t]$  for some  $p \in [0, 1]$ . Thus,  $\xi^+$  and  $\xi^-$  belong to the domain of attraction of positive stable laws with index  $\beta$  whose characteristic functions are denoted by  $\phi_+$  and  $\phi_-$ . Since  $\beta < 1$ , it follows (see Theorem 2, p. 448 in [12]) that  $\frac{1}{n^\beta} \sum_{j=1}^n \xi_j^+$  converges to a  $\beta$  stable random variable with characteristic function  $\phi_+$ . Therefore, we get  $|\varphi_{\xi^+}(t) - \phi_+(t)| \leq |t|^\beta h_+(|t|)$  for some increasing continuous function  $h_+$  such that  $h_+(0) = 0$ . The analogous statement is true for  $\varphi_{\xi^-}$ . Hence, for any real numbers  $s, t$

$$\begin{aligned} & |\varphi_{\xi^+}(t) + \varphi_{\xi^-}(s) - 1 - \phi_+(t)\phi_-(s)| \\ & \leq |\varphi_{\xi^+}(t) - \phi_+(t)| + |\varphi_{\xi^-}(s) - \phi_-(s)| + |(\phi_+(t) - 1)(\phi_-(s) - 1)| \\ & \leq |t|^\beta h_+(|t|) + |s|^\beta h_-(|s|) + C|s|^\beta |t|^\beta. \end{aligned}$$

Note also that  $|\langle \theta; d_n(x) \rangle| \leq m \max(|\theta_i|) N_n(x)$ . It follows that

$$\begin{aligned} & \left| \mathbb{E} \left[ e^{i(\sum_{j=1}^m (\theta_j (\tilde{Z}_n^+(t_j) - \tilde{Z}_n^+(t_{j-1})) + \gamma_j (\tilde{Z}_n^-(t_j) - \tilde{Z}_n^-(t_{j-1})))} \right) \middle| S \right] - \prod_x \phi_+ \left( \frac{\langle \theta; d_n(x) \rangle}{b_n} \right) \phi_- \left( \frac{\langle \gamma; d_n(x) \rangle}{b_n} \right) \right| \\ & \leq \sum_x \left| \varphi_{\xi^+} \left( \frac{\langle \theta; d_n(x) \rangle}{b_n} \right) + \varphi_{\xi^-} \left( \frac{\langle \gamma; d_n(x) \rangle}{b_n} \right) - 1 - \phi_+ \left( \frac{\langle \theta; d_n(x) \rangle}{b_n} \right) \phi_- \left( \frac{\langle \gamma; d_n(x) \rangle}{b_n} \right) \right| \\ & \leq C_{\beta, \gamma, \theta} \frac{\sum_x N_n^\beta(x)}{b_n^\beta} \left[ h_+ \left( \frac{m \|\theta\| N_n^*}{b_n} \right) + h_- \left( \frac{m \|\gamma\| N_n^*}{b_n} \right) + \left( \frac{N_n^*}{b_n} \right)^\beta \right], \end{aligned}$$

where  $N_n^* = \sup_x N_n(x)$  and  $\|\theta\| = \max(|\theta_i|)$ . Using Lemmas 5 and 7, the above quantity tends to 0 almost surely. Now,  $\phi_+$  and  $\phi_-$  get the same form as  $\phi$  (with  $A_2/A_1 = -\tan(\pi\beta/2)$ ). And as in the proof of the convergence of the finite-dimensional distributions, we get that almost surely

$$\lim_{n \rightarrow +\infty} \prod_x \phi_+ \left( \frac{\langle \theta; d_n(x) \rangle}{b_n} \right) = \prod_{j=1}^m \phi_+ \left( \frac{\theta_j (t_j - t_{j-1})^{1/\beta} \Gamma(\beta + 1)^{1/\beta}}{(\pi A)^{(\beta-1)/\beta}} \right).$$

The same is true for  $\prod_x \phi_- \left( \frac{\langle \gamma; d_n(x) \rangle}{b_n} \right)$ . □

**Remark 11.** Case  $d > \alpha$  and  $\beta < 1$ .

The proof is exactly the same using (9) and (10).

### 3. Proof of the local limit theorem in the lattice case

#### 3.1. The event $\Omega_n$

Set

$$N_n^* := \sup_y N_n(y) \quad \text{and} \quad R_n := \#\{y: N_n(y) > 0\}.$$

We also define, for every  $n \geq 1$ ,

$$V_n := \sum_{i,j=0}^{n-1} N_n(x)^\beta.$$

**Lemma 12.** For every  $n \geq 1$  and  $1 > \gamma > 0$ , set

$$\Omega_n = \Omega_n(\gamma) := \left\{ R_n \leq \frac{n}{(\log \log(n))^{1/4}} \text{ and } N_n^* \leq n^\gamma \right\}.$$

Then,  $\mathbb{P}(\Omega_n) = 1 - o(b_n^{-1})$ . Moreover, the following also holds on  $\Omega_n$ :

$$(\log \log(n))^{1/4} \leq N_n^* \quad \text{and} \quad V_n \geq n^{1-\gamma(1-\beta)_+}. \quad (17)$$

**Proof.** We first prove that

$$\mathbb{P}(R_n \geq n(\log \log(n))^{-1/4}) = o(b_n^{-1}). \quad (18)$$

Let us recall that for every  $a, b \in \mathbb{N}$ , we have

$$\mathbb{P}(R_n \geq a + b) \leq \mathbb{P}(R_n \geq a) \mathbb{P}(R_n \geq b). \quad (19)$$

The proof is given for instance in [9]. We will moreover use the fact that  $\mathbb{E}[R_n] \sim cn(\log(n))^{-1}$  and  $\text{Var}(R_n) = O(n^2 \log^{-4}(n))$  (see [17]). Hence, for  $n$  large enough, there exists  $C > 0$  such that we have

$$\begin{aligned} \mathbb{P}\left(R_n \geq \frac{n}{(\log \log(n))^{1/4}}\right) &\leq \mathbb{P}\left(R_n \geq \left\lfloor \frac{n(\log \log(n))^{1/4}}{\log(n)} \right\rfloor\right)^{\lfloor \log(n)(\log \log(n))^{-1/2} \rfloor} \\ &\leq \mathbb{P}\left(|R_n - \mathbb{E}[R_n]| \geq \frac{1}{2} \left\lfloor \frac{n(\log \log(n))^{1/4}}{\log(n)} \right\rfloor\right)^{\lfloor \log(n)(\log \log(n))^{-1/2} \rfloor} \end{aligned}$$

$$\begin{aligned}
 &\leq \left( \frac{5 \operatorname{Var}(R_n) \log^2(n)}{n^2 (\log \log(n))^{1/2}} \right)^{\lfloor \log(n) (\log \log(n))^{-1/2} \rfloor} \\
 &\leq \left( \frac{C n^2 \log^2(n) / \log^4(n)}{n^2 \sqrt{\log \log(n)}} \right)^{\lfloor \log(n) (\log \log(n))^{-1/2} \rfloor} \\
 &\leq \left( \frac{C}{(\log(n))^2} \right)^{\lfloor \log(n) (\log \log(n))^{-1/2} \rfloor} \\
 &= \exp \left( -\log(n) \sqrt{\log \log(n)} \left( 1 - \frac{\log(C)}{2 \log \log(n)} \right) \right).
 \end{aligned}$$

This ends the proof of (18).

Let us now prove that

$$\mathbb{P}[N_n^* \geq n^\gamma] = o(b_n^{-1}). \quad (20)$$

We have

$$\begin{aligned}
 \mathbb{P}(N_n^* \geq n^\gamma) &\leq \sum_x \mathbb{P}(N_n(x) \geq n^\gamma) \\
 &= \sum_x \mathbb{P}(T_x \leq n; N_n(x) \geq n^\gamma), \quad \text{where } T_x := \inf\{n > 1, \text{ s.t. } S_n = x\}, \\
 &\leq \sum_x \mathbb{P}(T_x \leq n) \mathbb{P}(N_n(0) \geq n^\gamma) \\
 &\leq \mathbb{E}[R_n] \mathbb{P}(T_0 \leq n)^{n^\gamma}.
 \end{aligned}$$

Hence, (20) follows now from  $\mathbb{E}[R_n] \sim cn(\log(n))^{-1}$ , and from  $\mathbb{P}(T_0 > n) \sim C/\log(n)$ .

Since  $n = \sum_y N_n(y) \leq R_n N_n^*$ , we get that  $N_n^* \geq \frac{n}{R_n} \geq (\log \log(n))^{1/4}$  on  $\Omega_n$ .

To prove the lower bound for  $V_n$ , note that, for  $\beta \geq 1$ ,  $V_n = \sum_y N_n(y)^\beta \geq \sum_y N_n(y) = n$ . For  $\beta < 1$ , on  $\Omega_n$ , we have

$$n = \sum_y N_n(y) = \sum_y N_n(y)^\beta N_n(y)^{1-\beta} \leq V_n (N_n^*)^{1-\beta} \leq V_n n^{\gamma(1-\beta)}. \quad \square$$

### 3.2. Scheme of the proof

It is easy to see (cf. the proof of Lemma 5 in [7]) that  $\mathbb{P}(Z_n = \lfloor b_n x \rfloor) = 0$  if  $\mathbb{P}(n\xi_0 - \lfloor b_n x \rfloor \notin d_0\mathbb{Z}) = 1$ , and that if  $\mathbb{P}(n\xi_0 - \lfloor b_n x \rfloor \in d_0\mathbb{Z}) = 1$ ,

$$\mathbb{P}(Z_n = \lfloor b_n x \rfloor) = \frac{d_0}{2\pi} \int_{-\pi/d_0}^{\pi/d_0} e^{-it \lfloor b_n x \rfloor} \mathbb{E} \left[ \prod_y \varphi_\xi(t N_n(y)) \right] dt.$$

In view of Lemma 12, we have to estimate

$$\frac{d_0}{2\pi} \int_{-\pi/d_0}^{\pi/d_0} e^{-it \lfloor b_n x \rfloor} \mathbb{E} \left[ \prod_y \varphi_\xi(t N_n(y)) \mathbf{1}_{\Omega_n} \right] dt.$$

This is done in several steps presented in the following propositions.

**Proposition 13.** Let  $\gamma \in (0, 1/(\beta + 1))$  and  $\delta \in (0, 1/(2\beta))$  s.t.  $\gamma \frac{(1-\beta)_+}{\beta} < \delta < 1/\beta - \gamma$ . Then, we have

$$\frac{d_0}{2\pi} \int_{\{|t| \leq n^\delta/b_n\}} e^{-it \lfloor b_n x \rfloor} \mathbb{E} \left[ \prod_y \varphi_\xi(t N_n(y)) \mathbf{1}_{\Omega_n} \right] dt = d_0 \frac{C(x)}{b_n} + o(b_n^{-1}),$$

uniformly in  $x \in \mathbb{R}$ .

Recall next that the characteristic function  $\phi$  of the limit distribution of  $(n^{-1/\beta} \sum_{k=1}^n \xi_{ke_1})_n$  has the following form:

$$\phi(u) = e^{-|u|^\beta (A_1 + iA_2 \operatorname{sgn}(u))},$$

with  $0 < A_1 < \infty$  and  $|A_1^{-1} A_2| \leq |\tan(\pi\beta/2)|$ . It follows that the characteristic function  $\varphi_\xi$  of  $\xi_0$  satisfies:

$$1 - \varphi_\xi(u) \sim |u|^\beta (A_1 + iA_2 \operatorname{sgn}(u)) \quad \text{when } u \rightarrow 0. \quad (21)$$

Therefore there exist constants  $\varepsilon_0 > 0$  and  $\sigma > 0$  such that

$$\max(|\phi(u)|, |\varphi_\xi(u)|) \leq \exp(-\sigma |u|^\beta) \quad \text{for all } u \in [-\varepsilon_0, \varepsilon_0]. \quad (22)$$

Since  $\overline{\varphi_\xi(t)} = \varphi_\xi(-t)$  for every  $t \geq 0$ , the following propositions achieve the proof of Theorem 3:

**Proposition 14.** Let  $\delta$  and  $\gamma$  be as in Proposition 13. Then there exists  $c > 0$  such that

$$\int_{n^\delta/b_n}^{\varepsilon_0 n^{-\gamma}} \mathbb{E} \left[ \prod_y |\varphi_\xi(t N_n(y))| \mathbf{1}_{\Omega_n} \right] dt = o(e^{-n^c}).$$

**Proposition 15.** There exists  $c > 0$  such that

$$\int_{\varepsilon_0 n^{-\gamma}}^{\pi/d_0} \mathbb{E} \left[ \prod_y |\varphi_\xi(t N_n(y))| \mathbf{1}_{\Omega_n} \right] dt = o(e^{-n^c}).$$

### 3.3. Proof of Proposition 13

Remember that  $V_n = \sum_{z \in \mathbb{Z}^d} N_n^\beta(z)$ . We start by a preliminary lemma.

**Lemma 16.** (1) If  $\beta > 1$ ,  $\sup_n \mathbb{E}[(\frac{n \log(n)^{\beta-1}}{V_n})^{1/(\beta-1)}] < +\infty$ .

(2) If  $\beta \leq 1$ ,  $\forall p \in \mathbb{N}$ ,  $\sup_n \mathbb{E}[(\frac{n \log(n)^{\beta-1}}{V_n})^p] < +\infty$ .

**Proof.** For  $\beta > 1$ , using Hölder's inequality with  $p = \beta$ , we get

$$n = \sum_x N_n(x) \leq V_n^{1/\beta} R_n^{(\beta-1)/\beta}$$

which means that

$$\left( \frac{n \log(n)^{\beta-1}}{V_n} \right)^{1/(\beta-1)} \leq \frac{\log(n) R_n}{n}.$$

But it is proved in [17], Eq. (7.a), that  $\mathbb{E}[R_n] = \mathcal{O}(n/\log(n))$ . The result follows.

The result is obvious for  $\beta = 1$ . For  $\beta < 1$ , Hölder's inequality with  $p = 2 - \beta$  yields

$$n = \sum_x N_n^{\beta/(2-\beta)}(x) N_n^{2(1-\beta)/(2-\beta)}(x) \leq V_n^{1/(2-\beta)} \left( \sum_x N_n^2(x) \right)^{(1-\beta)/(2-\beta)}$$

and so

$$\frac{n \log(n)^{\beta-1}}{V_n} \leq \left( \frac{\sum_x N_n^2(x)}{n \log(n)} \right)^{1-\beta}.$$

It is therefore enough to prove that there exists  $c > 0$  such that

$$\sup_n \mathbb{E} \left[ \exp \left( c \frac{\sum_x N_n^2(x)}{n \log(n)} \right) \right] < \infty. \tag{23}$$

Note that  $\sum_x N_n^2(x) = \sum_{k=0}^{n-1} N_n(S_k)$ . By Jensen's inequality, we get thus

$$\mathbb{E} \left[ \exp \left( c \frac{\sum_x N_n^2(x)}{n \log(n)} \right) \right] \leq \frac{1}{n} \sum_{k=0}^{n-1} \mathbb{E} \left[ \exp \left( c \frac{N_n(S_k)}{\log(n)} \right) \right].$$

Observe now that  $N_n(S_k) = \sum_{j=0}^k \mathbf{1}_{\{S_k - S_j = 0\}} + \sum_{j=k+1}^{n-1} \mathbf{1}_{\{S_j - S_k = 0\}} \stackrel{(d)}{=} N_{k+1}(0) + N'_{n-k}(0) - 1$ , where  $(N'_n(x), n \in \mathbb{N}, x \in \mathbb{Z}^d)$  is an independent copy of  $(N_n(x), n \in \mathbb{N}, x \in \mathbb{Z}^d)$ . Hence,

$$\mathbb{E} \left[ \exp \left( c \frac{\sum_x N_n^2(x)}{n \log(n)} \right) \right] \leq \mathbb{E} \left[ \exp \left( c \frac{N_n(0)}{\log(n)} \right) \right]^2.$$

But,  $\forall t > 0$ ,

$$\mathbb{P}(N_n(0) \geq t \log(n)) \leq \mathbb{P}(T_0 \leq n)^{\lceil t \log(n) \rceil}$$

and

$$\mathbb{E} \left[ \exp \left( c \frac{N_n(0)}{\log(n)} \right) \right] \leq 1 + \int_0^\infty c \exp(ct) \exp(-\lceil t \log(n) \rceil \mathbb{P}(T_0 > n)) dt.$$

Now (23) follows then from the fact that  $\exists C > 0$  such that  $\mathbb{P}(T_0 > n) \sim C / \log(n)$  for any integer  $n \geq 1$ . □

The next step is

**Lemma 17.** *Under the hypotheses of Proposition 13, we have*

$$\int_{\{|t| \leq n^\delta / b_n\}} e^{-it \lfloor b_n x \rfloor} \mathbb{E} \left[ \left\{ \prod_y \varphi_\xi(t N_n(y)) - e^{-|t|^\beta (A_1 + i A_2 \operatorname{sgn}(t)) V_n} \right\} \mathbf{1}_{\Omega_n} \right] dt = o(b_n^{-1}),$$

uniformly in  $x \in \mathbb{R}$ .

**Proof.** Let

$$E_n(t) := \prod_y \varphi_\xi(t N_n(y)) - \prod_y \exp(-|t|^\beta N_n^\beta(y) (A_1 + i A_2 \operatorname{sgn}(t))).$$

Since  $\gamma + \delta < \beta^{-1}$ , we get, on  $\Omega_n$  and if  $|t| \leq n^\delta b_n^{-1}$

$$|E_n(t)| \leq \sum_y |\varphi_\xi(t N_n(y)) - \exp(-|t|^\beta N_n^\beta(y) (A_1 + i A_2 \operatorname{sgn}(t)))| \exp \left( -\sigma |t|^\beta \sum_{z \neq y} N_n^\beta(z) \right)$$

for  $n$  large enough. Observe next that (21) implies

$$|\varphi_\xi(u) - \exp(-|u|^\beta (A_1 + i A_2 \operatorname{sgn}(u)))| \leq |u|^\beta h(|u|) \quad \text{for all } u \in \mathbb{R},$$

with  $h$  a continuous and monotone function on  $[0, +\infty)$  vanishing at 0. Therefore we get

$$|E_n(t)| \leq |t|^\beta h(n^{\gamma+\delta} b_n^{-1}) \sum_y N_n^\beta(y) \exp\left(-\sigma |t|^\beta \sum_{z \neq y} N_n^\beta(z)\right).$$

Now, according to (17) and since  $\gamma < \frac{1}{\beta+1} \leq \frac{1}{\beta+(1-\beta)_+}$ , if  $n$  is large enough, we have on  $\Omega_n$

$$\sum_{z \neq y} N_n^\beta(z) \geq V_n/2 \quad \text{for all } y \in \mathbb{Z}.$$

By using this and the change of variables  $v = t V_n^{1/\beta}$ , we get

$$\int_{\{|t| \leq n^\delta b_n^{-1}\}} \mathbb{E}[|E_n(t)| \mathbf{1}_{\Omega_n}] dt \leq h(n^{\gamma+\delta} b_n^{-1}) \mathbb{E}[V_n^{-1/\beta}] \int_{\mathbb{R}} |v|^\beta \exp(-\sigma |v|^\beta/2) dv = o(\mathbb{E}[V_n^{-1/\beta}]),$$

which proves the result according to Lemma 16. □

Finally Proposition 13 follows from the

**Lemma 18.** *Under the hypotheses of Proposition 13, we have*

$$\frac{d_0}{2\pi} \int_{\{|t| \leq n^\delta b_n^{-1}\}} e^{-it \lfloor b_n x \rfloor} \mathbb{E}[e^{-|t|^\beta V_n(A_1 + iA_2 \operatorname{sgn}(t))} \mathbf{1}_{\Omega_n}] dt = d_0 \frac{C(x)}{b_n} + o(b_n^{-1}),$$

uniformly in  $x \in \mathbb{R}$ .

**Proof.** Set

$$I_{n,x} := \int_{\{|t| \leq n^\delta b_n^{-1}\}} e^{-it \lfloor b_n x \rfloor} e^{-|t|^\beta V_n(A_1 + iA_2 \operatorname{sgn}(t))} dt = \int_{\{|t| \leq n^\delta b_n^{-1}\}} e^{-it \lfloor b_n x \rfloor} \phi(t V_n^{1/\beta}) dt.$$

Since  $|\lfloor b_n x \rfloor - b_n x| \leq 1$  and  $\delta < (2\beta)^{-1}$ , we have

$$I_{n,x} = \int_{\{|t| \leq n^\delta b_n^{-1}\}} e^{-it b_n x} \phi(t V_n^{1/\beta}) dt + o(b_n^{-1}).$$

Next, with the change of variable  $v = t b_n$ , we get:

$$\int_{\{|t| \leq n^\delta b_n^{-1}\}} e^{-it b_n x} \phi(t V_n^{1/\beta}) dt = b_n^{-1} \{V_n^{-1/\beta} b_n f(x V_n^{-1/\beta} b_n) - J_{n,x}\}, \tag{24}$$

where  $f$  is the density function of the distribution with characteristic function  $\phi$  and where

$$J_{n,x} := \int_{\{|v| \geq n^\delta\}} e^{-ivx} \phi(v b_n^{-1} V_n^{1/\beta}) dv.$$

By Lemma 5 (applied with  $m = 1$ ,  $t_1 = \theta_1 = 1$ ,  $\gamma = \beta$ ),  $(W_n := b_n V_n^{-1/\beta})_n$  converges almost surely, as  $n \rightarrow \infty$ , to the constant  $\Gamma(\beta + 1)^{-1/\beta} (\pi A)^{1-1/\beta}$ . Moreover, Lemma 16 ensures that the sequence  $(W_n, n \geq 1)$  is uniformly integrable, so actually the convergence holds in  $\mathbb{L}^1$ . From which we conclude that

$$\mathbb{E}[W_n f(x W_n)] = \mathbb{E}[W f(x W)] + o(1) = C(x) + o(1),$$

uniformly in  $x$ .

In view of (24), it only remains to prove that  $\mathbb{E}[J_{n,x} \mathbf{1}_{\Omega_n}] = o(1)$  uniformly in  $x$ . But this follows from the basic inequality

$$\mathbb{E}[|J_{n,x} \mathbf{1}_{\Omega_n}|] \leq \int_{\{|v| \geq n^\delta\}} \mathbb{E}[e^{-A_1|v|^\beta V_n/b_n^\beta} \mathbf{1}_{\Omega_n}] dv,$$

and from the lower bound for  $V_n$  given in (17) and from the choice  $\delta > \gamma(1 - \beta)_+/\beta$ .  $\square$

### 3.4. Proof of Proposition 14

Recall that on  $\Omega_n$ ,  $N_n(y) \leq n^\gamma$ , for all  $y \in \mathbb{Z}^d$ . Hence by (22),

$$K_n := \int_{n^\delta/b_n}^{\varepsilon_0 n^{-\gamma}} \mathbb{E}\left[\prod_y |\varphi_\xi(tN_n(y))| \mathbf{1}_{\Omega_n}\right] dt \leq \int_{n^\delta/b_n}^{\varepsilon_0 n^{-\gamma}} \mathbb{E}[\exp(-\sigma t^\beta V_n) \mathbf{1}_{\Omega_n}] dt.$$

With the change of variable  $s = tV_n^{1/\beta}$ , we get

$$\begin{aligned} K_n &\leq \mathbb{E}\left[V_n^{-1/\beta} \int_{n^\delta V_n^{1/\beta} b_n^{-1}}^{\varepsilon_0 n^{-\gamma} V_n^{1/\beta}} \exp(-\sigma s^\beta) ds \mathbf{1}_{\Omega_n}\right] \\ &\leq \frac{1}{n^{1/\beta - \gamma(1-\beta)_+/\beta}} \int_{n^{\delta - \gamma(1-\beta)_+/\beta} \log(n)^{(1-\beta)_+/\beta}}^{+\infty} \exp(-\sigma s^\beta) ds, \end{aligned}$$

which proves the proposition since  $\delta > \gamma(1 - \beta)_+/\beta$ .

### 3.5. Proof of Proposition 15

We adapt the proof of [7], Proposition 10. We will see that the argument of ‘‘peaks’’ still works here. We endow  $\mathbb{Z}^d$  with the ordered structure given by the relation  $<$  defined by

$$(\alpha_1, \dots, \alpha_d) < (\beta_1, \dots, \beta_d) \Leftrightarrow \exists i \in \{1, \dots, d\}, \alpha_i < \beta_i, \forall j < i, \alpha_j = \beta_j.$$

We consider  $\mathcal{C}^+ = (x_1, \dots, x_T) \in (\mathbb{Z}^d \setminus \{0\})^T$  for some positive integer  $T$  such that:

- $x_1 + \dots + x_T = 0$ ;
- for every  $i = 1, \dots, T$ ,  $\mathbb{P}(X_1 = x_i) > 0$ ;
- there exists  $I_1 \in \{1, \dots, T\}$  such that
  - for every  $i = 1, \dots, I_1$ ,  $x_i > 0$ ,
  - for every  $i = I_1 + 1, \dots, T$ ,  $x_i < 0$ .

Let us write  $\mathcal{C}^- := (x_{T-i+1})_{i=1, \dots, T}$ . We define  $B := \sum_{i=1}^{I_1} x_i$ . We observe that

$$p := \mathbb{P}((X_1, \dots, X_T) = \mathcal{C}^+) = \mathbb{P}((X_1, \dots, X_T) = \mathcal{C}^-) > 0.$$

We notice that  $(X_1, \dots, X_T) = \mathcal{C}^+$  corresponds to a trajectory visiting  $B$  only once before going back to the origin at time  $T$  (and without visiting  $-B$ ). Analogously,  $(X_1, \dots, X_T) = \mathcal{C}^-$  corresponds to a trajectory that goes down to  $-B$  and comes back up to 0 (and without visiting  $B$ ), and staying at a distance smaller than  $\tilde{d}/2$  of the origin with  $\tilde{d} := \sum_{i=1}^T |x_i|$  (where  $|\cdot|$  is the absolute value if  $d = 1$  and  $|(a, b)| = \max(|a|, |b|)$  if  $d = 2$ ). We introduce now the event

$$\mathcal{D}_n := \left\{ C_n > \frac{np}{2T} \right\},$$

where

$$C_n := \#\left\{ k = 0, \dots, \left\lfloor \frac{n}{T} \right\rfloor - 1 : (X_{kT+1}, \dots, X_{(k+1)T}) = \mathcal{C}^\pm \right\}.$$



Since the sequences  $(X_{kT+1}, \dots, X_{(k+1)T})$ , for  $k \geq 0$ , are independent of each other, Chernoff's inequality implies that there exists  $c > 0$  such that

$$\mathbb{P}(\mathcal{D}_n) = 1 - o(e^{-cn}).$$

We introduce now the notion of "loop." We say that there is a loop based on  $y$  at time  $n$  if  $S_n = y$  and  $(X_{n+1}, \dots, X_{n+T}) = \mathcal{C}^\pm$ . We will see (in Lemma 19 below) that, on  $\Omega_n \cap \mathcal{D}_n$ , there is a large number of  $y \in \mathbb{Z}^d$  on which are based a large number of loops. For any  $y \in \mathbb{Z}^d$ , let

$$C_n(y) := \#\left\{k = 0, \dots, \left\lfloor \frac{n}{T} \right\rfloor - 1 : S_{kT} = y \text{ and } (X_{kT+1}, \dots, X_{(k+1)T}) = \mathcal{C}^\pm\right\},$$

be the number of loops based on  $y$  before time  $n$  (and at times which are multiple of  $T$ ), and let

$$p_n := \#\left\{y \in \mathbb{Z}^d : C_n(y) \geq \frac{\log \log(n)^{1/4} p}{4T}\right\},$$

be the number of sites  $y \in \mathbb{Z}^d$  on which at least  $a_n := \lfloor \frac{\log \log(n)^{1/4} p}{4T} \rfloor$  loops are based.

**Lemma 19.** *On  $\Omega_n \cap \mathcal{D}_n$ , we have,  $p_n \geq c' n^{1-\gamma}$  with  $c' = p/(4T)$ .*

**Proof.** Note that  $C_n(y) \leq N_n^*$  for all  $y \in \mathbb{Z}^d$ . Thus on  $\Omega_n \cap \mathcal{D}_n$ , we have

$$\begin{aligned} \frac{np}{2T} &\leq \sum_{y \in \mathbb{Z}^d : C_n(y) < a_n} C_n(y) + \sum_{y \in \mathbb{Z}^d : C_n(y) \geq a_n} C_n(y) \\ &\leq R_n a_n + N_n^* p_n \leq \frac{np}{4T} + p_n n^\gamma, \end{aligned}$$

according to Lemma 12. This proves the lemma. □

We have proved that, if  $n$  is large enough, the event  $\Omega_n \cap \mathcal{D}_n$  is contained in the event

$$\mathcal{E}_n := \{p_n \geq c' n^{1-\gamma}\}.$$

Now, on  $\mathcal{E}_n$ , we consider  $(Y_i)_{i=1, \dots, \lfloor c'' n^{1-\gamma} \rfloor}$  (with  $c'' := c'/(2\tilde{d})$  if  $d = 1$  and with  $c'' := c'/2\tilde{d}^2$  if  $d = 2$ ) such that

- on each  $Y_i$ , at least  $a_n$  loops are based;
- for every  $i, j$  such that  $i \neq j$ , we have  $|Y_i - Y_j| > \tilde{d}/2$ .

For every  $i = 1, \dots, \lfloor c'' n^{1-\gamma} \rfloor$ , let  $t_i^{(1)}, \dots, t_i^{(a_n)}$  be the  $a_n$  first times (which are multiples of  $T$ ) when a loop is based on the site  $Y_i$ . We also define  $N_n^0(Y_i + B)$  as the number of visits of  $S$  before time  $n$  to  $Y_i + B$ , which do not occur during the time intervals  $[t_i^{(j)}, t_i^{(j)} + T]$ , for  $j \leq a_n$ .

Since our construction is basically the same as in [7], Section 2.8, the proof of the following lemma is exactly the same as the proof of [7], Lemma 16, and we do not prove it again.

**Lemma 20.** *Conditionally to the event  $\mathcal{E}_n$ ,  $(N_n(Y_i + B) - N_n^0(Y_i + B))_{i \geq 1}$  is a sequence of independent identically distributed random variables with binomial distribution  $\mathcal{B}(a_n; \frac{1}{2})$ . Moreover this sequence is independent of  $(N_n^0(Y_i + B))_{i \geq 1}$ .*

Let  $\eta$  be a real number such that  $\gamma < \eta < (1 - \gamma)/\beta$  (this is possible since  $\gamma < 1/(\beta + 1)$ ). We define

$$\forall n \geq 1, \quad d_n := n^{-\eta}.$$

Let now  $\rho := \sup\{|\varphi_\xi(u)| : d(u, \frac{2\pi}{d_0}\mathbb{Z}) \geq \varepsilon_0\}$ . According to Formula (22) and since  $\lim_{n \rightarrow \infty} d_n = 0$ , for  $n$  large enough, we have

$$\begin{aligned} |\varphi_\xi(u)| &\leq \rho \mathbf{1}_{\{d(u, (2\pi/d_0)\mathbb{Z}) \geq \varepsilon_0\}} + \exp\left(-\sigma d\left(u, \frac{2\pi}{d_0}\mathbb{Z}\right)^\beta\right) \mathbf{1}_{\{d(u, (2\pi/d_0)\mathbb{Z}) < \varepsilon_0\}} \\ &\leq \exp(-\sigma d_n^\beta), \end{aligned}$$

as soon as  $d(u, \frac{2\pi}{d_0}\mathbb{Z}) \geq d_n$ . Therefore, for  $n$  large enough,

$$\prod_z |\varphi_\xi(tN_n(z))| \leq \exp\left(-\sigma d_n^\beta \#\left\{z : d\left(tN_n(z), \frac{2\pi}{d_0}\mathbb{Z}\right) \geq d_n\right\}\right). \tag{25}$$

Then notice that

$$d\left(tN_n(z), \frac{2\pi}{d_0}\mathbb{Z}\right) \geq d_n \iff N_n(z) \in \mathcal{I} := \bigcup_{k \in \mathbb{Z}} I_k, \tag{26}$$

where for all  $k \in \mathbb{Z}$ ,

$$I_k := \left[ \frac{2k\pi}{d_0 t} + \frac{d_n}{t}, \frac{2(k+1)\pi}{d_0 t} - \frac{d_n}{t} \right].$$

In particular  $\mathbb{R} \setminus \mathcal{I} = \bigcup_{k \in \mathbb{Z}} J_k$ , where for all  $k \in \mathbb{Z}$ ,

$$J_k := \left( \frac{2k\pi}{d_0 t} - \frac{d_n}{t}, \frac{2k\pi}{d_0 t} + \frac{d_n}{t} \right).$$

**Lemma 21.** *Under the hypotheses of Proposition 15, for every  $i \leq \lfloor c''n^{1-\gamma} \rfloor$ ,  $t \in (\varepsilon_0 n^{-\gamma}, \pi/d_0)$  and  $n$  large enough,*

$$\mathbb{P}(N_n(Y_i + B) \in \mathcal{I} | \mathcal{E}_n, N_n^0(Y_i + B)) \geq \frac{1}{3} \quad \text{almost surely.}$$

Assume for a moment that this lemma holds true and let us finish the proof of Proposition 15. Lemmas 20 and 21 ensure that conditionally to  $\mathcal{E}_n$  and  $((N_n^0(Y_i + B), i \geq 1))$ , the events  $\{N_n(Y_i + B) \in \mathcal{I}\}$ ,  $i \geq 1$ , are independent of each other, and all happen with probability at least  $1/3$ . Therefore, since  $\Omega_n \cap \mathcal{D}_n \subseteq \mathcal{E}_n$ , there exists  $c > 0$ , such that

$$\mathbb{P}\left(\Omega_n \cap \mathcal{D}_n, \#\{i : N_n(Y_i + B) \in \mathcal{I}\} \leq \frac{c''n^{1-\gamma}}{4}\right) \leq \mathbb{P}\left(B_n \leq \frac{c''n^{1-\gamma}}{4}\right) = o(\exp(-cn^{1-\gamma})),$$

where for all  $n \geq 1$ ,  $B_n$  has binomial distribution  $\mathcal{B}(\lfloor c''n^{1-\gamma} \rfloor; \frac{1}{3})$ .

But if  $\#\{z : N_n(z) \in \mathcal{I}\} \geq \frac{c''n^{1-\gamma}}{4}$ , then by (25) and (26), there exists a constant  $c > 0$ , such that

$$\prod_z |\varphi_\xi(tN_n(z))| \leq \exp(-cn^{1-\gamma} d_n^\beta),$$

which proves Proposition 15 since  $1 - \gamma - \beta\eta > 0$ .

**Proof of Lemma 21.** First notice that by Lemma 20, for any  $H \geq 0$ ,

$$\mathbb{P}(N_n(Y_i + B) \in \mathcal{I} | \mathcal{E}_n, N_n^0(Y_i + B) = H) = \mathbb{P}(H + \beta_n \in \mathcal{I}), \tag{27}$$

where  $\beta_n$  is a random variable with binomial distribution  $\mathcal{B}(a_n; \frac{1}{2})$ . We will use the following result whose proof is postponed.

**Lemma 22.** *Under the hypotheses of Proposition 15, for every  $t \in (\varepsilon_0 n^{-\gamma}, \pi/d_0)$  and for  $n$  large enough, the following holds:*

(i) *For any integer  $k$  such that all the elements of  $I_k - H$  are smaller than  $\frac{a_n}{2}$ ,*

$$\mathbb{P}(\beta_n \in (I_k - H)) \geq \mathbb{P}(\beta_n \in (J_k - H)).$$

(ii) *For any integer  $k$  such that all the elements of  $I_k - H$  are larger than  $\frac{a_n}{2}$ ,*

$$\mathbb{P}(\beta_n \in (I_k - H)) \geq \mathbb{P}(\beta_n \in (J_{k+1} - H)).$$

Now call  $k_0$  the largest integer satisfying the condition appearing in (i) and  $k_1$  the smallest integer satisfying the condition appearing in (ii). We have  $k_1 = k_0 + 1$  or  $k_1 = k_0 + 2$ . According to Lemma 22, we have

$$\begin{aligned} \mathbb{P}(H + \beta_n \in \mathcal{I}) &\geq \sum_{k \leq k_0} \mathbb{P}(H + \beta_n \in I_k) + \sum_{k \geq k_1} \mathbb{P}(H + \beta_n \in I_k) \\ &\geq \sum_{k \leq k_0} \mathbb{P}(H + \beta_n \in J_k) + \sum_{k \geq k_1} \mathbb{P}(H + \beta_n \in J_{k+1}) \\ &= \mathbb{P}(H + \beta_n \notin \mathcal{I}) - \mathbb{P}(H + \beta_n \in J_{k_0+1} \cup J_{k_1}). \end{aligned}$$

Hence,

$$\mathbb{P}(H + \beta_n \in \mathcal{I}) \geq \frac{1}{2} [1 - \mathbb{P}(H + \beta_n \in J_{k_0+1} \cup J_{k_1})].$$

The interval  $J_{k_1}$  being of length  $2d_n/t$ , according to the uniform version of the local limit theorem for  $\beta_n$ , for every  $t \geq \varepsilon_0 n^{-\gamma}$ , we have

$$\mathbb{P}(H + \beta_n \in J_{k_1}) \leq \left( \frac{2d_n}{\varepsilon_0 n^{-\gamma}} + 1 \right) a_n^{-1/2}.$$

We conclude that  $\mathbb{P}(H + \beta_n \in J_{k_1}) = o(1)$ . The same holds for  $\mathbb{P}(H + \beta_n \in J_{k_0+1})$ , so that for  $n$  large enough,

$$\mathbb{P}(H + \beta_n \in \mathcal{I}) \geq \frac{1}{2} [1 - o(1)] \geq \frac{1}{3}.$$

Together with (27), this concludes the proof of Lemma 21. □

**Proof of Lemma 22.** We only prove (i), since (ii) is similar. So let  $k$  be an integer such that all the elements of  $I_k - H$  are smaller than  $\frac{a_n}{2}$ . Assume that  $(J_k - H) \cap \mathbb{Z}$  contains at least one nonnegative integer (otherwise  $\mathbb{P}(\beta_n \in (J_k - H)) = 0$  and there is nothing to prove). Let  $z_k$  denote the greatest integer in  $J_k - H$ , so that by our assumption  $\mathbb{P}(\beta_n = z_k) > 0$  (remind that  $0 \leq z_k < \frac{a_n}{2}$ ). By monotonicity of the function  $z \mapsto \mathbb{P}(\beta_n = z)$ , for  $z \leq \frac{a_n}{2}$ , we get

$$\mathbb{P}(\beta_n \in J_k - H) \leq \mathbb{P}(\beta_n = z_k) \# ((J_k - H) \cap \mathbb{Z}) \leq \mathbb{P}(\beta_n = z_k) \left\lceil \frac{2d_n}{t} \right\rceil.$$

In the same way,

$$\mathbb{P}(\beta_n \in I_k - H) \geq \mathbb{P}(\beta_n = z_k) \# ((I_k - H) \cap \mathbb{Z}) \geq \mathbb{P}(\beta_n = z_k) \left[ \frac{2\pi}{d_0 t} - \frac{2d_n}{t} \right].$$

Hence

$$\mathbb{P}(\beta_n \in I_k - H) \geq \frac{\lfloor 2\pi/(d_0 t) - 2d_n/t \rfloor}{\lceil 2d_n/t \rceil} \mathbb{P}(\beta_n \in J_k - H).$$

But  $\pi/(d_0t) \geq 1$  and  $\lim_{n \rightarrow +\infty} d_n = 0$  by hypothesis. It follows immediately that for  $n$  large enough, we have  $2d_n < \pi/(2d_0)$ , and so

$$\left\lfloor \frac{2\pi}{d_0t} - \frac{2d_n}{t} \right\rfloor \geq \left\lfloor \frac{3\pi}{2d_0t} \right\rfloor \geq 1 + \left\lfloor \frac{\pi}{2d_0t} \right\rfloor \geq \left\lfloor \frac{\pi}{2d_0t} \right\rfloor \geq \left\lfloor \frac{2d_n}{t} \right\rfloor.$$

This concludes the proof of the lemma. □

#### 4. Proof of the local limit theorem in the strongly nonlattice case

As in [7], the proof in the strongly nonlattice case is closely related to the proof in the lattice case. We assume here that  $\xi$  is strongly nonlattice. In that case, there exist  $\varepsilon_0 > 0$ ,  $\sigma > 0$  and  $\rho < 1$  such that  $|\varphi_\xi(u)| \leq \rho$  if  $|u| \geq \varepsilon_0$  and  $|\varphi_\xi(u)| \leq \exp(-\sigma|u|^\beta)$  if  $|u| < \varepsilon_0$ .

We use here the notations of Section 3 with the hypotheses on  $\gamma$ , and  $\delta$  of Proposition 13. According to Lemma IV-5 of [14], it is enough to prove that

$$\lim_{n \rightarrow \infty} \sup_{x \in \mathbb{R}} |b_n \mathbb{E}[h(Z_n - b_n x)] - C(x)\hat{h}(0)| = 0 \tag{28}$$

for any positive, Lebesgue-integrable and continuous real function  $h$  with continuously differentiable and compactly supported Fourier transform (let us notice that such functions exist, take for example  $h_0(u) := \int_{u-\pi/2}^{u+\pi/2} (\frac{\sin t}{t})^4 dt$ ). Let  $h$  be such a function. By Fourier inverse transform, we have

$$b_n \mathbb{E}[h(Z_n - b_n x)] = \frac{b_n}{2\pi} \int_{\mathbb{R}} e^{-iub_n x} \mathbb{E} \left[ \prod_{x \in \mathbb{Z}^d} \varphi_\xi(u N_n(x)) \right] \hat{h}(u) du.$$

Since  $\hat{h}$  is  $L^1$ , we can restrict our study to the event  $\Omega_n$  of Lemma 12. The part of the integral corresponding to  $|u| \leq n^\delta b_n^{-1}$  is treated exactly as in Proposition 13. The only change is that we have to check that

$$\lim_{n \rightarrow \infty} b_n \int_{\{|u| \leq n^\delta b_n^{-1}\}} \mathbb{E} [e^{-A_1|u|^\beta V_n} \mathbf{1}_{\Omega_n}] \sup_{|u| \leq n^\delta b_n^{-1}} |\hat{h}(u) - \hat{h}(0)| du = 0,$$

which is obviously true since  $\hat{h}$  is a Lipschitz function.

Now, since  $\hat{h}$  is bounded, the part corresponding to  $n^\delta b_n^{-1} \leq |u| \leq \varepsilon_0 n^{-\gamma}$  is treated as in the proof of Proposition 14 (since it only uses the behavior of  $\varphi_\xi$  around 0, which is the same).

Finally, it remains to prove that

$$\lim_{n \rightarrow \infty} b_n \int_{\{|u| \geq \varepsilon_0 n^{-\gamma}\}} \left| \mathbb{E} \left[ \prod_x \varphi_\xi(u N_n(x)) \mathbf{1}_{\Omega_n} \right] \right| |\hat{h}(u)| du = 0. \tag{29}$$

We note that, if  $|u| \geq \varepsilon_0 n^{-\gamma}$  and  $x \in \mathbb{Z}^d$ , we have

$$\begin{aligned} |\varphi_\xi(u N_n(x))| &\leq \exp(-\sigma|u|^\beta N_n^\beta(x)) \mathbf{1}_{\{|u N_n(x)| \leq \varepsilon_0\}} + \rho \mathbf{1}_{\{|u N_n(x)| \geq \varepsilon_0\}} \\ &\leq \exp(-\sigma \varepsilon_0^\beta n^{-\gamma\beta} N_n^\beta(x)) \mathbf{1}_{\{|u N_n(x)| \leq \varepsilon_0\}} + \rho \mathbf{1}_{\{|u N_n(x)| \geq \varepsilon_0\}}. \end{aligned}$$

For  $n$  large enough,  $\rho \leq \exp(-\sigma \varepsilon_0^\beta n^{-\gamma\beta})$ . Therefore, if  $n$  is large enough, then for all  $x$  and  $u$  such that  $N_n(x) \geq 1$  and  $|u| \geq \varepsilon_0 n^{-\gamma}$ , we have

$$|\varphi_\xi(u N_n(x))| \leq \exp(-\sigma \varepsilon_0^\beta n^{-\gamma\beta}).$$

Hence,

$$\left| \mathbb{E} \left[ \prod_x \varphi_\xi(u N_n(x)) \mathbf{1}_{\Omega_n} \right] \right| \leq \mathbb{E} [\exp(-\sigma \varepsilon_0^\beta n^{-\gamma\beta} R_n) \mathbf{1}_{\Omega_n}] \leq \exp(-\sigma \varepsilon_0^\beta n^{1-\gamma(1+\beta)}).$$

Therefore, since  $\gamma(1 + \beta) < 1$  and  $\hat{h}$  is compactly supported, we have

$$\lim_{n \rightarrow \infty} b_n \int_{\{|u| \geq \varepsilon_0 n^{-\gamma}\}} \left| \mathbb{E} \left[ \prod_x \varphi_\xi(u N_n(x)) \mathbf{1}_{\mathcal{O}_n} \right] \right| |\hat{h}(u)| \, du = 0.$$

This concludes the proof of Theorem 4.

### Appendix: Complement to Cerny's paper

There is a missing argument in the proof of (6) in [8]. It concerns the control of the term

$$A_n := \sum_{(m_0, \dots, m_{2k-1}) \in M_n} (\mathbb{P}(S_{m_u+m_v} = 0) - \mathbb{P}(S_{m_u+\dots+m_v} = 0)) \prod_{i \in \{1, \dots, 2k-1\} \setminus \{u, v\}} \mathbb{P}(S_{m_i} = 0),$$

where  $k \geq 2$ ,  $1 \leq u \leq k-1$  and  $v = u+k$  are fixed integers and  $M_n := \{(m_0, \dots, m_{2k-1}) \in \mathbb{N}^{2k} : m_0 + \dots + m_{2k-1} \leq n; \forall i \notin \{u, v\}, m_i \geq 1\}$ . In order to obtain (6), it is necessary to prove that

$$A_n = O(n^2 (\ln n)^{2k-4}).$$

In [8], this estimate is proved using Karamata's Tauberian theorem. However, it is not clear that the sequence  $A_n$  is monotone.

To be complete, let us explain how this can be solved thanks to the argument used in [10] by Deligiannidis and Utev to prove their Theorem 2.2.

Summing over  $m_0, \dots, m_{u-1}, m_{v+1}, \dots, m_{2k-1}$ , and using the fact that  $\mathbb{P}(S_n = 0) = O(n^{-1})$ , we have

$$|A_n| \leq O(n (\ln n)^{k-2}) B_n$$

with

$$B_n := \sum_{(m_u, \dots, m_v) \in M'_n} |\mathbb{P}(S_{m_u+m_v} = 0) - \mathbb{P}(S_{m_u+\dots+m_v} = 0)| \prod_{i=u+1}^{v-1} \mathbb{P}(S_{m_i} = 0),$$

and  $M'_n := \{(m_u, \dots, m_v) \in \mathbb{N}^{k+1} : m_u + \dots + m_v \leq n; \forall i = u+1, \dots, v-1, m_i \geq 1\}$ . Summing over  $m_u, m_v$ , we get

$$B_n = \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \sum_{N=0}^{n - \sum_{i=1}^{k-1} m_i} (N+1) |\mathbb{P}(S_N = 0) - \mathbb{P}(S_{N + \sum_{i=1}^{k-1} m_i} = 0)| \prod_{i=1}^{k-1} \mathbb{P}(S_{m_i} = 0)$$

with  $\tilde{M}_{k-1, n} := \{(m_1, \dots, m_{k-1}) \in (\mathbb{N} \setminus \{0\})^{k-1} : m_1 + \dots + m_{k-1} \leq n\}$ . Now from the assumptions on the random walk, there exists  $\sigma > 0$  such that, for every  $t \in [-\pi, \pi]^d$  ( $d = 1, 2$ ) and every  $j \in \mathbb{N}$ , we have  $|\varphi_{X_1}(t)| \leq e^{-\sigma|t|^d}$  and  $|1 - (\varphi_{X_1}(t))^j| \leq (2 + \sigma) \min(j|t|^d, 1)$ . Therefore, we have

$$\begin{aligned} B_n &\leq O(1) \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \left( \prod_{i=1}^{k-1} \frac{1}{m_i} \right)^{n - \sum_{i=1}^{k-1} m_i} \sum_{N=0}^{n - \sum_{i=1}^{k-1} m_i} (N+1) \int_{[-\pi, \pi]^d} |\varphi_{X_1}(t)|^N |1 - (\varphi_{X_1}(t))^{\sum_{i=1}^{k-1} m_i}| \, dt \\ &\leq O(1) \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \left( \prod_{i=1, \dots, k-1} \frac{1}{m_i} \right)^{n - \sum_{i=1}^{k-1} m_i} \sum_{N=0}^{n - \sum_{i=1}^{k-1} m_i} (N+1) J_N \left( \sum_{i=1}^{k-1} m_i \right) \end{aligned}$$

with

$$J_N(x) := \int_0^{\pi\sqrt{d}} e^{-N\sigma t} \min(tx, 1) dt.$$

We observe that  $J_0 \leq \pi\sqrt{d}$  and that, for every  $N \geq 1$  and every  $x \in \mathbb{N}$ , we have

$$J_N(x) \leq \frac{x}{(N\sigma)^2} (1 - e^{-N\sigma/x}) + \frac{e^{-N\sigma/x}}{N\sigma} = \frac{1}{N\sigma} f\left(\frac{N\sigma}{x}\right), \quad (30)$$

where  $f(y) = \frac{1}{y}(1 - e^{-y}) + e^{-y}$ . Since  $f(y) \asymp 1$  for  $y \ll 1$ , and  $f(y) \asymp \frac{1}{y}$  for  $y \gg 1$ , there exists a constant  $C$  such that  $f(y) \leq Cg(y)$ , where  $g(y) := \mathbb{1}_{[0,1]}(y) + \frac{1}{y}\mathbb{1}_{[1,+\infty[}(y)$ . Hence, we have for  $1 \leq x \leq n$ ,

$$\begin{aligned} \sum_{N=0}^{n-x} (N+1)J_N(x) &\leq O(1) \left( 1 + \sum_{N=1}^{n-x} g\left(\frac{N\sigma}{x}\right) \right) \\ &\leq O(1) \left( 1 + \frac{x}{\sigma} \int_0^{n\sigma/x} g(y) dy \right) \\ &\leq O(1) \left( x + x \log\left(\frac{n}{x}\right) \right). \end{aligned}$$

Hence,

$$\begin{aligned} B_n &\leq O(1) \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \left( \prod_{i=1}^{k-1} \frac{1}{m_i} \right) \left( \sum_{i=1}^{k-1} m_i \right) \left[ 1 + \ln\left(\frac{n}{\sum_{i=1}^{k-1} m_i}\right) \right] \\ &= O(1) \sum_{i=1}^{k-1} \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \left( \prod_{j=1, j \neq i}^{k-1} \frac{1}{m_j} \right) \left[ 1 + \ln\left(\frac{n}{\sum_{i=1}^{k-1} m_i}\right) \right] \\ &\leq O(1) I_n, \end{aligned}$$

with

$$\begin{aligned} I_n &:= \sum_{(m_1, \dots, m_{k-1}) \in \tilde{M}_{k-1, n}} \left( \prod_{i=1}^{k-2} \frac{1}{m_i} \right) \left[ 1 + \ln\left(\frac{n}{\sum_{i=1}^{k-1} m_i}\right) \right] \\ &= \sum_{(m_1, \dots, m_{k-2}) \in \tilde{M}_{k-2, n}} \left( \prod_{i=1}^{k-2} \frac{1}{m_i} \right) \sum_{l=\sum_{i=1}^{k-2} m_i + 1}^n \left[ 1 + \ln\left(\frac{n}{l}\right) \right] \\ &\leq \sum_{(m_1, \dots, m_{k-2}) \in \tilde{M}_{k-2, n}} \left( \prod_{i=1}^{k-2} \frac{1}{m_i} \right) n \int_0^1 (-\ln x + 1) dx = O(n(\ln n)^{k-2}). \end{aligned}$$

## Acknowledgments

The authors thank the referees for insightful comments that helped us to improve the paper significantly. They are also deeply grateful to Bruno Schapira and Loïc Hervé for helpful and stimulating discussions.

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