

# Local times on curves and uniform invariance principles

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**Summary.** Sufficient conditions are given for a family of local times  $\{L_t^\mu\}$  of  $d$ -dimensional Brownian motion to be jointly continuous as a function of  $t$  and  $\mu$ . Then invariance principles are given for the weak convergence of local times of lattice valued random walks to the local times of Brownian motion, uniformly over a large family of measures. Applications include some new results for intersection local times for Brownian motions on  $\mathbb{R}^2$  and  $\mathbb{R}^3$ .

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## 1 Introduction

For local times of one-dimensional Brownian motion, there is a huge body of literature for the modulus of joint continuity and for invariance principles (see, e.g., [Bo2]). However, when one turns to  $d$ -dimensional Brownian motion, much less is known. Local times at points do not exist, and the appropriate analogue to study is additive functionals  $L_t^\mu$  corresponding to certain measures  $\mu$ . For continuity, there are a few results concerning joint continuity in  $t$  and  $\mu$ , such as [B] and [Y]. There are some results on the convergence of functionals of random walks to a single additive functional (see [Dy]), but nothing, as far as we know, on uniform convergence to a family of additive functionals.

The purpose of this paper is to study continuity properties and invariance principles which are uniform over large families  $\mathfrak{M}$  of measures  $\mu$ . We use the term “local times on curves” instead of “additive functionals” because (1) most of the examples we look at have  $\mu$ 's supported on curves and (2) the term “additive functional” is strongly associated with probabilistic potential theory; we make no use of this deep subject, but instead rely on stochastic calculus methods.

Our first set of results concerns the continuity of  $L_t^\mu$  as a function of  $t$  and  $\mu$ . If  $\mathfrak{M}$  is a family of measures  $\mu$ , each of which satisfies a very mild regularity condition,

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we show that  $L_t^\mu$  is jointly continuous in  $t$  and  $\mu$ , even when  $\mathfrak{M}$  is a very large family. Largeness, here, is measured by the metric entropy of  $\mathfrak{M}$  with respect to a certain metric for the space of measures on  $\mathbb{R}^d$  with the topology of weak convergence.

The majority of the paper is concerned with invariance principles. We suppose that  $X_1, X_2, \dots$  is a sequence of mean 0, lattice valued i.i.d. random variables with finite variance, and possibly satisfying additional moment conditions. We let  $S_n$  denote the partial sums. We suppose that for each  $\mu \in \mathfrak{M}$ , there is a sequence of measures  $\mu_n$  converging weakly to  $\mu$ . Since the  $X_i$  are lattice valued, we suppose the  $\mu_n$  are supported on  $n^{-1/2}\mathbb{Z}^d$ . Then, if the  $\mu_n$  satisfy the same mild regularity condition as we imposed on the  $\mu$  and the metric entropy of the  $\mu_n$  is suitably bounded, then the local time for  $S_j/\sqrt{n}$  corresponding to  $\mu_n$  converges weakly to  $L_t^\mu$ , uniformly over  $\mu \in \mathfrak{M}$ . The size of the family  $\mathfrak{M}$  that is allowed is determined by the number of moments of the  $X_i$ .

Although our theorems are quite general, they also seem to be quite powerful, as a number of examples show. For example, in the case of classical additive functionals, where the  $\mu$ 's have densities with respect to Lebesgue measure, we get continuity results and invariance principles over a large class of functions, with minimal smoothness assumptions. If  $\mu$  is a measure supported on a curve and we approximate  $\mu$  by curves containing the support of  $S_j/\sqrt{n}$ , we get an invariance principle that is uniform over a large family of curves.

One of the most interesting examples is that of intersection local times. If  $\alpha(x, s, t)$  is the intersection local time of two independent Brownian motions, then  $\alpha$  measures the amount of time that the two Brownian motions differ by  $x$ ,  $x \in \mathbb{R}^2$ . LeGall [LG] and Rosen [Ro] have shown that the number of intersections of two independent random walks converges to the intersection local time of two independent Brownian motions at a single level  $x$  when the random walk has two moments. This result can also be obtained as a corollary of our methods. In addition, if the random walk has  $2 + \rho$  moments for some  $\rho > 0$ , we get the new result that weak convergence holds uniformly at all levels  $x$ . LeGall and Rosen also have results for invariance principles for  $k$ -multiple points. Again, with  $2 + \rho$  moments, we can get the corresponding uniform invariance principle.

To get some idea of the relative sharpness of our theorems, we look at the case of local times of one-dimensional Brownian motion. A problem that has been studied by a number of people is the question of an invariance principle that is uniform over all the levels  $x$ ; see [Bo2] and the references therein. As an immediate corollary of our theorems, we get an invariance principle, uniform over all levels  $x$ , provided the  $X_i$  have  $2 + \rho$  moments for some  $\rho > 0$ . The reader should compare this with the results of [Bo1]; there, using techniques highly specific to one-dimensional Brownian motion, the uniform invariance principle is obtained under the assumption of finite variance. See also [BK2] for further extensions.

Our results on the joint continuity of local times of curves with respect to  $t$  and the measure  $\mu$  are given in Sect. 2. We also remark there that many of the results have analogues for symmetric stable processes.

In Sect. 3 we prove a local central limit theorem. The theorem is that of Spitzer [S]; we derive an estimate of the error term that may be of independent interest.

In Sects. 4 and 5, we derive exponential estimates for the tails of the difference of two local times for the random walk. Some of these ideas seem likely to have applications elsewhere: the theme is that if one wants weak convergence or exponential estimates for additive functionals, one only has to compute first moments.

In Sect. 6, we give our invariance principles, with different versions depending on how well-behaved the tails of the  $X_i$  are. The fewer moments, the smaller the family  $\mathfrak{M}$  that is allowed. If one has only finite variance, one can still get convergence of the finite dimensional distributions if  $d \leq 3$ , but not (by our techniques) uniform results.

Finally, we give our examples, already discussed above, in Sect. 7. For the reader primarily interested in the applications, we suggest reading Sects. 6 and 7 first.

### 2 Construction and joint continuity

Let  $Z_t$  be Brownian motion on  $d$ -dimensional Euclidean space  $\mathbb{R}^d$ . Let  $g$  be the Green function of  $Z_t$  if  $d \geq 3$ . If  $d = 1$  or  $2$ ,  $g$  shall denote the 1-potential density of  $Z_t$ . So

$$g(x, y) = \begin{cases} \int_0^\infty p_s(x, y) ds & \text{if } d \geq 3 \\ 0 & \\ \int_0^\infty e^{-s} p_s(x, y) ds & \text{if } d = 1 \text{ or } 2 \end{cases}$$

where  $p_s(x, y)$  is the transition function of  $Z$ . We define the potential of a measure  $\mu$  by

$$g\mu(x) = \int g(x, y) \mu(dy) .$$

Then it is well-known [BG] that if the map  $x \mapsto g\mu(x)$  is bounded and continuous, then there is a continuous additive functional  $\{L_t^\mu\}$  so that

$$M_t^\mu = g\mu(Z_t) - g\mu(Z_0) + L_t^\mu \tag{2.1}$$

is a mean zero martingale.  $L_t^\mu$  is called a local time of  $Z$  on the support of  $\mu$ .

If  $\mathfrak{M}$  is a family of positive measures on  $\mathbb{R}^d$ , define

$$d_G(\mu, \nu) = \sup_{x \in \mathbb{R}^d} |g\mu(x) - g\nu(x)|, \quad \mu, \nu \in \mathfrak{M} .$$

Define  $H_G(\varepsilon) = H_G^{\mathfrak{M}}(\varepsilon)$  to be the metric entropy of  $\mathfrak{M}$  with respect to the norm  $d_G$ . In other words,  $H_G(\varepsilon) = \log N_G(\varepsilon)$ , where  $N_G(\varepsilon)$  is the minimum number of  $d_G$ -balls of radius  $\varepsilon$  required to cover  $\mathfrak{M}$ . If

$$H_G(x) \leq c_{2.1} x^{-r}, \quad x < 1 , \tag{2.2}$$

for some  $r$ , we say that the exponent of metric entropy of  $H_G$  is  $\leq r$ .

We then have

**Proposition 2.1** *If  $g\mu$  is bounded and continuous for each  $\mu \in \mathfrak{M}$  and if  $d_G(\mu, \nu) \leq 1$ , then*

$$\mathbb{P}^y \left( \sup_{t \leq 1} |L_t^\mu - L_t^\nu| \geq \lambda \right) \leq c_{2.2} \exp(-\lambda/c_{2.3} \sqrt{d_G(\mu, \nu)}), \quad \mu, \nu \in \mathfrak{M}, y \in \mathbb{R}^d ,$$

where  $c_{2.3}$  depends only on  $\sup_{\mu \in \mathfrak{M}} \|g\mu\|_\infty$ .

*Proof.* Let  $U_t^\mu = g\mu(Z_t) - g\mu(Z_0)$  and similarly for  $U_t^\nu$ . Note  $|U_t^\mu - U_t^\nu| \leq 2d_G(\mu, \nu)$ . Write  $N_t$  for  $M_t^\mu - M_t^\nu$ . Applying Itô's formula,

$$(U_t^\mu - U_t^\nu)^2 = 2 \int_0^t (U_s^\mu - U_s^\nu) dN_s - 2 \int_0^t (U_s^\mu - U_s^\nu) d(L_s^\mu - L_s^\nu) + [U^\mu - U^\nu, U^\mu - U^\nu]_t.$$

Since  $[U^\mu - U^\nu, U^\mu - U^\nu]_t$  is  $[N, N]_t$ , we take expectations to get

$$\begin{aligned} \mathbb{E}^\nu N_\tau^2 &= \mathbb{E}^\nu [N, N]_\tau \leq 4(d_G(\mu, \nu))^2 + 2d_G(\mu, \nu) \mathbb{E}^\nu (L_\tau^\mu + L_\tau^\nu) \\ &\leq 4(d_G(\mu, \nu))^2 + 4 \sup_{\mu \in \mathfrak{M}} \|g\mu\|_\infty d_G(\mu, \nu) \\ &\leq c_{2.4} d_G(\mu, \nu) \end{aligned} \tag{2.3}$$

for bounded stopping times  $\tau$ .

Consider arbitrary bounded stopping times  $T \geq S$ . Then denoting the shift operator by  $\theta_t$ ,

$$\begin{aligned} \mathbb{E}^\nu \{|N_T - N_S| | \mathcal{F}_S\} &\leq [\mathbb{E}^\nu \{|N_T - N_S|^2 | \mathcal{F}_S\}]^{1/2} \\ &= [\mathbb{E}^\nu \{[N, N]_T - [N, N]_S | \mathcal{F}_S\}]^{1/2} \\ &\leq [\mathbb{E}^\nu \{[N, N]_\infty \circ \theta_S | \mathcal{F}_S\}]^{1/2} \leq \left[ \sup_x \mathbb{E}^x [N, N]_\infty \right]^{1/2} \\ &\leq c_{2.4}^{1/2} (d_G(\mu, \nu))^{1/2} \quad (\text{by (2.3).}) \end{aligned}$$

Using (2.1), we get

$$\mathbb{E}^\nu \{|(L_T^\mu - L_T^\nu) - (L_S^\mu - L_S^\nu)| | \mathcal{F}_S\} \leq (4 + c_{2.4}^{1/2}) (d_G(\mu, \nu))^{1/2}.$$

Since  $L_t^\mu$  and  $L_t^\nu$  are continuous, they are predictable, and therefore an application of [DM] p. 193, completes the proof.  $\square$

**Theorem 2.2** *Let  $\mathfrak{M}$  be a family of positive measures on  $\mathbb{R}^d$ . Suppose*

- (i)  $\sup_{x \in \mathbb{R}^d} \sup_{\mu \in \mathfrak{M}} g\mu(x) < \infty$ , and for all  $\mu \in \mathfrak{M}$ ,  $x \mapsto g\mu(x)$  is continuous;
- (ii)  $H_G$  has exponent of metric entropy  $< r < 1/2$ .

*Then there exist versions of  $L_t^\mu$  such that  $(t, \mu) \mapsto L_t^\mu$  is almost surely jointly continuous. Moreover,*

$$\limsup_{\delta \rightarrow 0} \sup_{0 \leq t \leq 1} \sup_{\substack{\mu, \nu \in \mathfrak{M} \\ d_G(\mu, \nu) \leq \delta}} \frac{|L_t^\mu - L_t^\nu|}{\delta^{1/2-r}} < \infty, \quad \text{a.s.}$$

*Remark 2.3* One could give an integral condition that  $H_G$  needs to satisfy and also a more precise modulus of continuity, but even in the case of one-dimensional local times our result here is not sharp. This reflects the fact that Proposition 2.1 yields exponential tails for  $L_t^\mu - L_t^\nu$  and not Gaussian ones.

*Proof of Theorem 2.2* By the estimates of Proposition 2.1 and a standard metric entropy argument (cf. [Du]), we get the desired modulus of continuity for  $(t, \mu)$

restricted to  $[0, 1] \times D$ , where  $D$  is a countable dense subset of  $\mathfrak{M}$ . Define

$$\hat{L}_t^\mu = \lim_{\substack{\mu_n \rightarrow \mu \\ \mu_n \in D}} L_t^{\mu_n}.$$

In view of Proposition 2.1,  $\hat{L}_t^\mu = L_t^\mu, t \in [0, 1]$ , a.s. In particular, the potential of  $\hat{L}_t^\mu$  is the same as that of  $L_t^\mu$ , hence  $\hat{L}_t^\mu$  also corresponds to the measure  $\mu$ .

The  $\hat{L}_t^\mu$ , then, are versions of  $L_t^\mu$  satisfying the desired modulus of continuity.  $\square$

Define another metric on our family of measures  $\mathfrak{M}$  by

$$d_L(\mu, \nu) = \sup_{\psi \in \mathfrak{Q}} \left| \int \psi d\mu - \int \psi d\nu \right|, \quad \mu, \nu \in \mathfrak{M},$$

where  $\mathfrak{Q}$  is the collection of all functions  $\psi: \mathbb{R}^d \mapsto \mathbb{R}^+$  such that  $\|\psi\|_\infty \vee \|\nabla\psi\|_\infty \leq 1$ . It is not hard to show that the  $d_L$  metric metrizes weak convergence of probability measures. (This metric is equal to what is sometimes known as the *bounded Lipschitz metric*.)

*Example 2.4* Suppose  $d = 1$ . Consider point masses,  $\delta_x$  and  $\delta_y$ , on  $x$  and  $y$ , respectively. Then  $d_L(\delta_x, \delta_y) = \sup_{\psi \in \mathfrak{Q}} |\psi(x) - \psi(y)| \leq |x - y| \wedge 2$ . It is easy to see that we actually have equality here.

*Example 2.5* Fix two maps  $F_i: [0, 1] \rightarrow \mathbb{R}^d, i = 1, 2$ . Define for all Borel sets  $A \subseteq \mathbb{R}^d$ , the measures

$$\mu_i(A) = |\{0 \leq t \leq 1: F_i(t) \in A\}|, \quad i = 1, 2,$$

where  $|\cdot|$  denotes Lebesgue measure.

Choose  $\psi \in \mathfrak{Q}$ . Then

$$\left| \int \psi d\mu_1 - \int \psi d\mu_2 \right| = \left| \int_0^1 \psi(F_1(t)) dt - \int_0^1 \psi(F_2(t)) dt \right|,$$

and so

$$d_L(\mu_1, \mu_2) \leq \int_0^1 (|F_1(t) - F_2(t)| \wedge 2) dt,$$

much as in Example 2.4. The right hand side is equivalent to the  $L_0$ -metric corresponding to convergence in measure.

**Definition 2.6** Let  $\mathfrak{M}$  be a family of positive finite measures on  $\mathbb{R}^d$  such that for some  $\gamma \in \mathbb{R}^+$  and constant  $c_{2.5} = c_{2.5}(\gamma)$ ,

$$\sup_{\mu \in \mathfrak{M}} \sup_{x \in \mathbb{R}^d} \mu(B(x, r)) \leq c_{2.5} r^{d-2+\gamma}, \quad r \leq 1.$$

We call the largest such  $\gamma$  the *index* of  $\mathfrak{M}$ . If  $\mathfrak{M} = \{\mu_0\}$ , then we say that  $\gamma$  is the index of  $\mu_0$ .

**Proposition 2.7** *If index  $(\mathfrak{M}) > 0$  and  $\sup_{\mu \in \mathfrak{M}} \mu(\mathbb{R}^d) < \infty$ , then  $\|g\mu\|_\infty < \infty$  and  $g\mu(\cdot)$  is Hölder continuous for each  $\mu \in \mathfrak{M}$ .*

*Proof.* Consider the  $d \geq 3$  case first. Then  $g(x, y) = c_d|x - y|^{2-d}$  for some  $c_d$ . So

$$\begin{aligned} g\mu(x) &= \int g(x, y)\mu(dy) = \int_{B(x, 1)} g(x, y)\mu(dy) + \int_{B(x, 1)^c} g(x, y)\mu(dy) \\ &\leq c_d\mu(\mathbb{R}^d) + \int_{B(x, 1)} g(x, y)\mu(dy), \end{aligned}$$

where  $B(x, r)$  is the ball of radius  $r$  centered about  $x$ . But if  $0 < \gamma < \text{index}(\mathfrak{M})$ ,

$$\begin{aligned} \int_{B(x, 1)} g(x, y)\mu(dy) &= \sum_{j=0}^{\infty} \int_{2^{-(j+1)} \leq |x-y| < 2^{-j}} g(x, y)\mu(dy) \\ &\leq c_d \sum_{j=0}^{\infty} |2^{-(j+1)}|^{2-d} \mu(B(x, 2^{-j})), \\ &\leq c_{2.6} \sum_{j=0}^{\infty} 2^{-j\gamma} < \infty. \end{aligned}$$

The  $d = 2$  case is similar, since  $g(x, y) \leq -c_{2.7} \log|x - y|$  for  $|x - y| < 1$ . The  $d = 1$  case is also easy and is done in a similar fashion.

To show Hölder continuity, consider the  $d \geq 3$  case again. Then for  $\varepsilon > 2|x - y|$ ,

$$\begin{aligned} |g\mu(x) - g\mu(y)| &\leq \left| \int_{B(x, \varepsilon)} g(x, y)\mu(dy) - \int_{B(x, \varepsilon)} g(y, z)\mu(dz) \right| \\ &\quad + \left| \int_{B(x, \varepsilon)^c} (g(x, z) - g(y, z))\mu(dz) \right| = I_{2.4} + II_{2.4}. \end{aligned} \tag{2.4}$$

The second term is estimated as follows.

$$\begin{aligned} II_{2.4} &\leq \int_{B(x, \varepsilon)^c} |g(x, z) - g(y, z)| \\ &\leq c_{2.8}|x - y| \int_{B(x, \varepsilon)^c} (|x - z| \vee |y - z|)^{1-d} \mu(dz) \leq c_{2.9}\varepsilon^{1-d}|x - y|. \end{aligned} \tag{2.5}$$

For the first term of (2.4),

$$\begin{aligned} I_{2.4} &\leq \int_{B(x, \varepsilon)} (g(x, y) + g(y, z))\mu(dz) \\ &= \sum_{j=0}^{\infty} \int_{2^{-(j+1)\varepsilon} \leq |x-z| < 2^{-j\varepsilon}} (g(x, z) + g(y, z))\mu(dz) \\ &\leq 2 \sup_{\alpha} \sum_{j=-1}^{\infty} \int_{2^{-(j+1)\varepsilon} \leq |\alpha-z| < 2^{-j\varepsilon}} g(\alpha, z)\mu(dz) \\ &\leq c_{2.10} \sup_{\alpha} \sum_{j=-1}^{\infty} (2^{-(j+1)\varepsilon})^{2-d} \mu(B(\alpha, 2^{-j\varepsilon})) \\ &\leq c_{2.11} \sum_{j \geq -1} 2^{-j\gamma} \varepsilon^{\gamma} \leq c_{2.12} \varepsilon^{\gamma}. \end{aligned} \tag{2.6}$$

Putting (2.4), (2.5), and (2.6) together gives the existence of a constant  $c_{2.13}$  such that

$$\sup_{|x-y| \leq \delta} |g\mu(x) - g\mu(y)| \leq c_{2.13} |\delta \varepsilon^{1-d} + \varepsilon^\gamma|, \quad \varepsilon > 2\delta.$$

Therefore letting  $\varepsilon = c_{2.14} \delta^{1/(\gamma+d-1)}$ , we get,

$$\begin{aligned} \sup_{|x-y| \leq \delta} |g\mu(x) - g\mu(y)| &\leq c_{2.15} \delta^{\gamma/(d+\gamma-1)} \\ &\rightarrow 0 \quad \text{as } \delta \rightarrow 0. \end{aligned}$$

This proves the proposition for  $d \geq 3$ . The cases when  $d \leq 2$  are quite similar.  $\square$

The following relates the two metrics,  $d_L$  and  $d_G$ :

**Proposition 2.8** *If  $\mu$  and  $\nu$  are two positive finite measures on  $\mathbb{R}^d$  so that  $\text{index}(\{\mu, \nu\}) > \gamma > 0$ , then for some constant  $c_{2.16}$  depending only on  $\gamma$ ,*

$$d_G(\mu, \nu) \leq c_{2.16} [d_L(\mu, \nu)]^l$$

where  $l = \gamma/(d + \gamma - 1)$ .

*Proof.* Take  $d \geq 3$ :

$$\begin{aligned} d_G(\mu, \nu) &= \sup_{x \in \mathbb{R}^d} |g\mu(x) - g\nu(x)| \\ &\leq \sup_{x \in \mathbb{R}^d} \left| \int_{B(x, \varepsilon)} g(x, y)(\mu - \nu)(dy) \right| + \sup_{x \in \mathbb{R}^d} \left| \int_{B(x, \varepsilon)^c} g(x, y)(\mu - \nu)(dy) \right| \\ &= I_{2.7} + II_{2.7}. \end{aligned} \tag{2.7}$$

We proceed to estimate each term on the right hand side of (2.7) separately. Consider the second term first.

$$II_{2.7} = \sup_{x \in \mathbb{R}^d} c_d \left| \int_{B(x, \varepsilon)^c} |x - y|^{2-d} (\mu - \nu)(dy) \right|.$$

But  $\psi_0(y) \equiv |x - y|^{2-d} \wedge \varepsilon^{2-d}$  satisfies  $\|\psi_0\|_\infty \leq \varepsilon^{2-d}$  and  $\|\nabla \psi_0\|_\infty \leq c_{2.17} \varepsilon^{1-d}$  for a constant  $c_{2.17}$ . So

$$\begin{aligned} II_{2.7} &\leq c_{2.18} \sup_{x \in \mathbb{R}^d} \left\{ \left| \int \psi d(\mu - \nu) \right| : \|\psi\|_\infty \leq \varepsilon^{2-d}, \|\nabla \psi\|_\infty \leq c_{2.17} \varepsilon^{1-d} \right\} \\ &\leq c_{2.19} \varepsilon^{1-d} d_L(\mu, \nu), \quad \varepsilon \leq 1. \end{aligned} \tag{2.8}$$

We estimate the first term that appears in (2.7) exactly as in (2.6) to get

$$I_{2.7} \leq c_{2.20} \varepsilon^\gamma. \tag{2.9}$$

Putting (2.9), (2.8), and (2.7) together, and letting  $\varepsilon = d_L(\mu, \nu)^{1/(d+\gamma-1)}$ ,

$$\begin{aligned} d_G(\mu, \nu) &\leq c_{2.21} [\varepsilon^\gamma + d_L(\mu, \nu) \varepsilon^{1-d}] \\ &= c_{2.22} [d_L(\mu, \nu)]^l. \end{aligned}$$

The cases when  $d \leq 2$  are much the same.  $\square$

Now let  $H_L(\varepsilon)$  be the metric entropy with respect to metric  $d_L$ . Then Proposition 2.8 and Theorem 2.2 together yield the following

**Theorem 2.9** *Let  $\mathfrak{M}$  be a family of positive finite measures on  $\mathbb{R}^d$ . Assume that  $\text{index}(\mathfrak{M}) > \gamma$ , and let  $l = \gamma/(\gamma + d - 1)$ . If the exponent of metric entropy of  $H_L$  is  $< r < l/2$ , then there exist versions of  $L_t^\mu$  such that almost surely,*

$$\limsup_{\delta \rightarrow 0} \sup_{0 \leq t \leq 1} \sup_{\substack{\mu, \nu \in \mathfrak{M} \\ d_L(\mu, \nu) \leq \delta}} \frac{|L_t^\mu - L_t^\nu|}{\delta^{l/2 - r}} < \infty .$$

*Remark 2.10* Theorem 2.2 holds for many other Markov processes as well as for Brownian motion. For example, if  $Z_t$  is a symmetric stable process of order  $\alpha$  the statements and proofs of Proposition 2.1 and Theorem 2.2 go through with only minor changes.

In the stable case,  $g(x) = c_{2.23}|x|^{\alpha-d}$ . Just as above,  $g\mu$  will be continuous and bounded if  $\mu(B(x, r)) \leq c_{2.24}r^{d-\alpha+\gamma}$  uniformly for  $x \in \mathbb{R}^d, r \leq 1$ . Proposition 2.8 still holds provided we here define  $l$  by  $l = \gamma/(\gamma + d - \alpha + 1)$ . Similarly, with this change in the definition of  $l$ , Theorem 2.9 holds as well.

### 3 A local central limit theorem

In this section, we derive a local central limit theorem, which is that of Spitzer [S] pp. 76–78, but we use the additional moments to get better estimates of the error terms. We apply this to the problem of estimating the potential kernel of a random walk (cf. [NS] and [BR, Sect. 22]).

Let  $X_1, X_2, \dots$  be i.i.d.  $\mathbb{R}^d$ -valued random vectors. Here  $X_j = (X_j^1, \dots, X_j^d)$ . We consider the case  $d \geq 3$  for a random walk,  $S_n = \sum_{j=1}^n X_j$ . Assume the  $X_i$ 's take values in  $\mathbb{Z}^d$ , are mean 0, have the identity for covariance matrix, are strongly aperiodic, and have finite third moments. Let  $\phi(u) = \mathbb{E} \exp(iu \cdot X_1)$ , where  $a \cdot b$  is the usual inner product. Also let  $p_n(x, y) = \mathbb{P}^x\{S_n = y\}$ . Then we have the following local central limit theorem:

**Proposition 3.1** *There is a constant  $c_{3.1}$  such that for all  $n$ ,*

$$\sup_x |p_n(x, 0) - (2\pi n)^{-d/2} e^{-|x|^2/2n}| \leq c_{3.1} (1 + \mathbb{E}|X_1|^3) n^{-(d+1)/2} (\log^+ n)^{(d+3)/2} .$$

(Here  $\log^+(n) = \log(n) \vee 1$ ).

*Proof.* We follow the proof of P9 given in [S] pp. 76–78 closely. Let

$$E(n, x) = |p_n(0, x) - (2\pi n)^{-d/2} e^{-|x|^2/2n}| .$$

Then

$$\sup_x (2\pi n)^{d/2} E(n, x) \leq (2\pi)^{-d/2} \sum_{j=1}^4 I_j^{(n)}$$



where

$$\begin{aligned}
 I_1^{(n)} &= \sup_x \left| \int_{|\alpha| \leq A_n} (\phi^n(\alpha n^{-1/2}) - e^{-|\alpha|^2/2}) e^{-ix \cdot \alpha / \sqrt{n}} d\alpha \right|, \\
 I_2^{(n)} &= \sup_x \left| \int_{|\alpha| \geq A_n} e^{-|\alpha|^2/2 - ix \cdot \alpha n^{-1/2}} d\alpha \right|, \\
 I_3^{(n)} &= \sup_x \left| \int_{A_n \leq |\alpha| \leq r\sqrt{n}} \phi^n(\alpha n^{-1/2}) e^{-ix \cdot \alpha / \sqrt{n}} d\alpha \right|, \text{ and} \\
 I_4^{(n)} &= \sup_x \left| \int_{\substack{|\alpha| > r\sqrt{n} \\ \alpha \in \sqrt{n}\mathcal{C}}} \phi^n(\alpha n^{-1/2}) e^{-ix \cdot \alpha / \sqrt{n}} d\alpha \right|.
 \end{aligned}$$

Here  $\mathcal{C} = \{x \in \mathbb{R}^d : \max_{i \leq d} |x^i| \leq \pi\}$  is the unit cube of side  $\pi$ . Furthermore,  $A_n = \sqrt{2\beta \log n}$  for some large  $\beta$  and  $r > 0$  is a constant that is small. We proceed to estimate each term separately. Take  $n > 1$ .

$$I_1^{(n)} \leq n \int_{|\alpha| \leq A_n} |\phi(\alpha n^{-1/2}) - e^{-|\alpha|^2/2n}| d\alpha$$

since for all  $a, b \in \mathbb{R}^d$ ,

$$||a|^n - |b|^n| \leq n|a - b|(|a| \vee |b|)^{n-1}.$$

By definition,  $A_n/\sqrt{n} \rightarrow 0$ . So a Taylor expansion implies

$$\begin{aligned}
 \phi(\alpha n^{-1/2}) &= 1 - \frac{|\alpha|^2}{n} + E_1(\alpha, n) \\
 e^{-|\alpha|^2/2n} &= 1 - \frac{|\alpha|^2}{n} + E_2(\alpha, n)
 \end{aligned}$$

and for all  $|\alpha| \leq A_n$ ,  $E_i(\alpha, n) \leq c_{3.2}(1 + \mathbb{E}|X_1|^3)(|\alpha|^3/n^{3/2})$ ,  $i = 1, 2$ .

Therefore there exists a constant  $c_{3.3}$ , independent of  $\alpha \in \{x \in \mathbb{R}^d : |x| \leq A_n\}$ , so that

$$\begin{aligned}
 \sup_{|\alpha| \leq A_n} |\phi(\alpha n^{-1/2}) - e^{-|\alpha|^2/2n}| &\leq c_{3.3}(1 + \mathbb{E}|X_1|^3)A_n^3 n^{-3/2} \\
 &\leq c_{3.3}\sqrt{8\beta^3} (1 + \mathbb{E}|X_1|^3)n^{-3/2}(\log n)^{3/2}.
 \end{aligned}$$

Therefore,

$$\begin{aligned}
 I_1^{(n)} &\leq c_{3.4}(1 + \mathbb{E}|X_1|^3)nA_n^d n^{-3/2}(\log n)^{3/2}|B(0, 1)| \\
 &= c_{3.5}(1 + \mathbb{E}|X_1|^3)n^{-1/2}(\log n)^{(d+3)/2}.
 \end{aligned} \tag{3.1}$$

Next,

$$I_2^{(n)} \leq \int_{|\alpha| \geq A_n} e^{-|\alpha|^2/2} d\alpha \leq c_{3.6}n^{-\beta}. \tag{3.2}$$

The upper bound for  $I_3^{(n)}$  and  $I_4^{(n)}$  is done exactly as in [S]: for  $r$  small enough

$$I_3^{(n)} \leq 2n^{-\beta}. \tag{3.3}$$

Also, for  $r$  small enough, there exists  $\delta \in (0, 1)$  so that

$$\begin{aligned} I_4^{(n)} &\leq (1 - \delta)^n |\{\alpha : |\alpha| > r\sqrt{n}; \alpha \in \sqrt{n}\mathcal{C}\}| \\ &\leq c_{3.7} n^{-\beta}, \end{aligned} \tag{3.4}$$

where  $|B|$  is the Lebesgue measure of the Borel set  $B$ . Putting (3.1)–(3.4) together, the proposition is proved.  $\square$

Recall that  $X_1$  is subgaussian if there exists  $r > 0$  such that for all  $t > 0$ ,

$$\mathbb{E}e^{t|X_1|} \leq 2e^{t^2 r}. \tag{3.5}$$

Define the potential kernel for the random walk:

$$G(x, y) = \sum_n p_n(x, y).$$

Recall  $d \geq 3$  and hence  $G$  is well-defined and finite. Also recall that for some constant  $c_d$ ,  $g(x, y) = c_d|x - y|^{2-d}$ . Then the above proposition implies:

**Proposition 3.2** *Assume the  $X$ 's are subgaussian. Then there is a constant  $c_{3.8}$  so that for every  $x, y \in \mathbb{Z}^d$ ,*

$$|G(x, y) - g(x, y)| \leq c_{3.8}|x - y|^{1-d}(\log^+ |x - y|)^{1+d}.$$

*Proof.* By translation, it is enough to do this for  $y = 0$ . Clearly we shall only need to consider the case  $|x| \geq 1$ . By Chebyshev's inequality and (3.5), for all  $x \in \mathbb{R}^d$ ,

$$\mathbb{P}\{|S_n| \geq |x|\} \leq 2\exp(-|x|^2/4nr). \tag{3.6}$$

Let  $f(x) = \left\lceil \frac{|x|^2}{k \log |x|} \right\rceil$ ,  $k$  large. Then

$$\sum_{n=1}^{f(x)} E(n, x) \leq \sum_{n=1}^{f(x)} p_n(0, x) + \sum_{n=1}^{f(x)} (2\pi n)^{-d/2} e^{-|x|^2/2n}. \tag{3.7}$$

We bound each term on the right hand side of (3.7) separately:

$$\begin{aligned} \sum_{n=1}^{f(x)} p_n(0, x) &\leq \sum_{n=1}^{f(x)} \mathbb{P}\{|S_n| \geq |x|\} \\ &\leq \sum_{n=1}^{f(x)} 2e^{-|x|^2/4nr} \quad (\text{by (3.6)}) \\ &\leq 2f(x)\exp\left(-\frac{|x|^2}{4rf(x)}\right) \leq c_{3.9} \frac{|x|^2}{k \log |x|} \cdot |x|^{-k/4r} \\ &\leq |x|^{1-d} \quad \text{if } k \text{ is large enough.} \end{aligned} \tag{3.8}$$

Similarly,

$$\sum_{n=1}^{f(x)} (2\pi n)^{-d/2} e^{-|x|^2/2n} \leq |x|^{1-d} \quad \text{if } k \text{ is large enough, } |x| \geq 1. \tag{3.9}$$

Then (3.7), (3.8), and (3.9) imply that if  $k$  is large enough,

$$\sum_{n=1}^{f(x)} E(n, x) \leq 2|x|^{1-d}. \tag{3.10}$$

Now we estimate  $\sum_{n=f(x)}^{\infty} E(n, x)$  as follows:

$$\begin{aligned} \sum_{n \geq f(x)} E(n, x) &\leq c_{3.10} \sum_{n \geq f(x)} n^{-(d+1)/2} (\log^+ n)^{(d+3)/2} \quad (\text{Proposition 3.1}) \\ &\leq c_{3.11} |x|^{1-d} (\log^+ |x|)^{d+1}. \end{aligned} \tag{3.11}$$

Putting (3.11) and (3.10) together implies

$$\sum_{n \geq 1} E(n, x) \leq c_{3.12} |x|^{1-d} (\log^+ |x|)^{d+1}.$$

This in turn implies

$$\left| G(0, x) - \sum_{n \geq 1} (2\pi n)^{-d/2} e^{-|x|^2/2n} \right| \leq 1 + c_{3.12} |x|^{1-d} (\log^+ |x|)^{d+1}.$$

However, it is easy to show that

$$\left| \sum_{n \geq 1} (2\pi n)^{-d/2} e^{-|x|^2/2n} - g(x, 0) \right| \leq c_{3.13} |x|^{1-d} (\log^+ |x|)^{d+1}.$$

This proves the proposition.  $\square$

**Corollary 3.3** *Suppose  $d \geq 3$  and the  $X_i$ 's are subgaussian. Then*

- (a)  $G(0, x) \leq c_{3.14} (1 \wedge |x|^{2-d})$ ;
- (b) For each  $\beta \in (0, 1)$ , there exists  $c_{3.15} = c_{3.15}(\beta)$  such that for all  $x, y \in \mathbb{Z}^d - \{0\}$ ,

$$|G(0, x) - G(0, y)| \leq c_{3.15} \frac{|x - y|}{(|x| \wedge |y|)^{d-1}} + c_{3.15} \frac{|x - y|^{1-\beta}}{(|x| \wedge |y|)^{d-1-\beta}}. \tag{3.12}$$

*Proof.*

$$G(0, 0) = p_0(0, 0) + \sum_{n=1}^{\infty} p_n(0, 0) \leq 1 + \sum_{n=1}^{\infty} c_{3.16} n^{-d/2} \leq c_{3.17},$$

by Proposition 3.1. So part (a) follows by this equation if  $x = 0$  and by Proposition 3.2 if  $|x| \geq 1$ .

Note that part (b) is trivial if  $x = y$ . So let us exclude this case. By Proposition 3.2, if  $\beta \in (0, 1)$ ,

$$|G(0, x) - g(0, x)| \leq c_{3.18} |x|^{-(d-1-\beta)},$$

and similarly for  $|G(0, y) - g(0, y)|$ . But

$$|g(0, x) - g(0, y)| \leq c_{3.19} |y - x| / (|x| \wedge |y|)^{1-d}.$$

Since  $|x - y| \geq 1$ , part (b) follows by the triangle inequality.  $\square$

*Remark 3.4* Note that if  $\text{Cov}(X_1) = Q$  for  $Q$  any positive definite matrix, Corollary 3.3 still holds. To see this, one merely needs to replace the proofs of Propositions

3.1 and 3.2 with ones where the identity matrix  $I$  is replaced by  $Q$  and the density of a  $\mathfrak{R}(0, nI)$  r.v. is replaced by that of a  $\mathfrak{R}(0, nQ)$  r.v. (cf. [S]).

*Remark 3.5* The assumption that the random walk be strongly aperiodic may be removed by the method of Spitzer.

### 4 Moment bounds

In this section we consider the analogues of some of the results of Sect. 2 with random walk in place of Brownian motion. Assume that  $d \geq 3$  and that the  $X$ 's are such that conclusions (a) and (b) of Corollary 3.3 hold. Fix  $n$ . Let  $\mu_n$  be a finite measure supported on  $n^{-1/2}\mathbb{Z}^d$ . Let  $\mathfrak{M}_n$  be a family of such measures.

Let us define  $index_n(\mathfrak{M}_n)$  to be the largest  $\gamma$  such that there exists  $c_{4.1}$  with

$$\mu_n(B(x, s)) \leq c_{4.1} s^{d-2+\gamma}, \quad x \in \mathbb{R}^d, s \in [1/2\sqrt{n}, 1], \mu_n \in \mathfrak{M}_n. \tag{4.1}$$

Note, taking  $x \in n^{-1/2}\mathbb{Z}^d$  and  $s = 1/2\sqrt{n}$ , then in particular

$$\mu_n(\{x\}) \leq c_{4.1} n^{1-(d+\gamma)/2} \leq c_{4.1} n^{1-d/2}. \tag{4.2}$$

Define

$$L_k^{n, \mu_n} = n^{d/2-1} \sum_{j=0}^{k-1} \mu_n(\{S_j/\sqrt{n}\}). \tag{4.3}$$

**Proposition 4.1** *If  $index_n(\{\mu_n\}) > \gamma$ , then  $\sup_x \mathbb{E}^x L_\infty^{n, \mu_n} \leq c_{4.2}$ , where  $c_{4.2}$  depends only on  $\mu_n(\mathbb{R}^d)$ ,  $\gamma$ , and the constant  $c_{4.1}$  of (4.1).*

*Proof.* By translation invariance, it suffices to suppose  $x = 0$ .

$$\begin{aligned} \mathbb{E}^0 L_\infty^{n, \mu_n} &= n^{d/2-1} \sum_{j=0}^\infty \sum_{y \in \mathbb{Z}^d} \mu_n(\{y/\sqrt{n}\}) p_j(0, y) \\ &= n^{d/2-1} \sum_{y \in \mathbb{Z}^d} G(0, y) \mu_n(\{y/\sqrt{n}\}) \\ &\leq c_{4.3} n^{d/2-1} \sum_{k=0}^\infty \sum_{2^k \leq |y| < 2^{k+1}} |y|^{2-d} \mu_n(\{y/\sqrt{n}\}) \\ &\quad + n^{d/2-1} G(0, 0) \mu_n(\{0\}) \\ &\leq c_{4.4} n^{d/2-1} \sum_{k=0}^\infty 2^{k(2-d)} \mu_n(B(0, 2^{k+1}/\sqrt{n})) + n^{d/2-1} G(0, 0) \mu_n(\{0\}) \\ &= I_{4.5} + II_{4.5} \end{aligned} \tag{4.5}$$

$II_{4.5}$  is bounded using (4.2) and Corollary 3.3(a). For  $I_{4.5}$ , note

$$\begin{aligned} \sum_{k=0}^\infty 2^{k(2-d)} \mu_n(B(0, 2^{k+1}/\sqrt{n})) &= \sum_{2^k \leq \sqrt{n}} + \sum_{2^k > \sqrt{n}} \\ &\leq c_{4.1} \sum_{2^k \leq \sqrt{n}} 2^{k(2-d)} (2^{k+1}/\sqrt{n})^{d-2+\gamma} \\ &\quad + c_{4.5} \sum_{2^k > \sqrt{n}} 2^{k(2-d)} \leq c_{4.6} n^{1-d/2}. \end{aligned} \tag{4.6} \quad \square$$

**Corollary 4.2** *Let  $y \in n^{-1/2} \mathbb{Z}^d$ . If  $\mu_n^r$  is  $\mu_n$  restricted to  $B(y, r) - \{y\}$ , then  $\mathbb{E}^y L_\infty^{n, \mu_n^r} \leq c_{4.7} r^\gamma$ .*

*Proof.* The proof is very similar to that of Proposition 4.1, except that we may omit  $I_{4.5}$  and in estimating  $I_{4.5}$  in (4.6), we need only look at  $\sum_{2^k \leq r\sqrt{n}}$ .  $\square$

Recalling the definition of  $d_L$  from Sect. 2, notice that

$$d_L(\mu_n, \nu_n) = \sup \left\{ \sum_{y \in \mathbb{Z}^d} \psi(y/\sqrt{n})(\mu_n - \nu_n)(\{y/\sqrt{n}\}) : \psi \in \mathcal{Q} \right\}.$$

Taking  $\psi = c_{4.8}/\sqrt{n}$  at  $x \in n^{-1/2} \mathbb{Z}^d$  and 0 on  $B(x, 1/2\sqrt{n})^c$ , we see

$$|\mu_n(\{x\}) - \nu_n(\{x\})| \leq c_{4.9} \sqrt{n} d_L(\mu_n, \nu_n). \tag{4.7}$$

**Lemma 4.3** *Suppose  $\|\psi\|_\infty \leq 1$ ,  $\mu(\mathbb{R}^d), \nu(\mathbb{R}^d) \leq c_{4.10}$ , and  $|\psi(x) - \psi(y)| \leq |x - y|^\alpha$ . Then  $|\int \psi(y)(\mu - \nu)(dy)| \leq c_{4.11}(d_L(\mu, \nu))^\alpha$ .*

*Proof.* Let  $\varphi$  be a smooth, nonnegative, radially symmetric function with compact support and  $\int_{\mathbb{R}^d} \varphi(x) dx = 1$ . Let  $\varphi_\varepsilon(x) = \varepsilon^{-d} \varphi(x/\varepsilon)$ ,  $\psi_\varepsilon = \psi * \varphi_\varepsilon$  for  $\varepsilon > 0$ . First,

$$\begin{aligned} |\psi_\varepsilon(x) - \psi(x)| &= \left| \int [\psi(x - y) - \psi(x)] \varphi_\varepsilon(y) dy \right| \\ &\leq \int |y|^\alpha \varphi_\varepsilon(y) dy = \varepsilon^\alpha \int |y|^\alpha \varphi(y) dy \leq c_{4.12} \varepsilon^\alpha. \end{aligned}$$

Next, let  $u$  be a unit vector,  $\nabla_u f = \nabla f \cdot u$ . Since  $\int \varphi_\varepsilon(x - y) dy$  is constant, then

$$\int \nabla_u \varphi_\varepsilon(y) dy = 0.$$

So

$$\begin{aligned} |\nabla_u \psi_\varepsilon(x)| &= \left| \int \psi(x - y) \nabla_u \varphi_\varepsilon(y) dy \right| = \left| \int [\psi(x - y) - \psi(x)] \nabla_u \varphi_\varepsilon(y) dy \right| \\ &\leq \int |y|^\alpha \varepsilon^{-(d+1)} \varphi(y/\varepsilon) dy \leq c_{4.13} \varepsilon^{\alpha-1}. \end{aligned}$$

Hence,

$$\begin{aligned} \left| \int \psi d(\mu - \nu) \right| &\leq \int |\psi - \psi_\varepsilon| d(\mu + \nu) + \left| \int \psi_\varepsilon d(\mu - \nu) \right| \\ &\leq c_{4.12} \varepsilon^\alpha (\mu(\mathbb{R}^d) + \nu(\mathbb{R}^d)) + c_{4.13} \varepsilon^{\alpha-1} d_L(\mu, \nu). \end{aligned}$$

Now take  $\varepsilon = d_L(\mu, \nu)$ .  $\square$

Let  $\mathfrak{M}_n$  be a family of measures supported on  $n^{-1/2} \mathbb{Z}^d$  with  $\text{index}_n(\mathfrak{M}_n) > \gamma$ . We now obtain:

**Proposition 4.4** *For each  $\beta > 0$ ,*

$$\sup_{\mu_n, \nu_n \in \mathfrak{M}_n} \sup_x |\mathbb{E}^x(L_\infty^{n, \mu_n} - L_\infty^{n, \nu_n})| \leq c_{4.14} (d_L(\mu_n, \nu_n))^{l_\beta},$$

where

$$l_\beta = \gamma(1 - \beta)/(d + \gamma - 1 - \beta) \tag{4.8}$$

and  $c_{4.14}$  depends on  $c_{4.1}$ ,  $\gamma$ ,  $\beta$ , and  $\sup_{\mathfrak{M}_n} \mu_n(\mathbb{R}^d)$ .

*Proof.* By translation invariance we may suppose  $x = 0$ . Write  $\delta$  for  $d_L(\mu_n, \nu_n)$ ,  $l$  for  $l_0$ . Let  $G_K(y) = G(0, y) \wedge K$ . As in (4.5),

$$\begin{aligned} \mathbb{E}^0 L_\infty^{n, \mu_n} - \mathbb{E}^0 L_\infty^{n, \nu_n} &= n^{d/2-1} \sum_{y \neq 0} [G(0, y) - G_K(y)](\mu_n - \nu_n)(\{y/\sqrt{n}\}) \\ &\quad + n^{d/2-1} \sum_y G_K(y)(\mu_n - \nu_n)(y/\sqrt{n}) \\ &\quad + n^{d/2-1} [G(0, 0) - G_K(0)](\mu_n - \nu_n)(\{0\}) \\ &= I_{4.9} + II_{4.9} + III_{4.9}. \end{aligned} \tag{4.9}$$

Note  $G(0, y) - G_K(y) = 0$  if  $|y| > c_{4.15} K^{1/(2-d)}$ . So, writing  $\zeta = K^{1/(2-d)}$ ,

$$\begin{aligned} I_{4.9} &\leq n^{d/2-1} \sum_{0 < |y| < c_{4.15}\zeta} G(0, y)(\mu_n + \nu_n)(\{y/\sqrt{n}\}) \\ &= 2(\mathbb{E}^0 L_\infty^{n, \mu_n^r} + \mathbb{E}^0 L_\infty^{n, \nu_n^r}) \leq 4c_{4.16} r^\gamma, \end{aligned}$$

where  $r = c_{4.15}\zeta/\sqrt{n}$ .

From Corollary 3.3(a), (b) it follows that if  $x \in n^{-1/2}\mathbb{Z}^d$ , then

$$G(0, x\sqrt{n}) \leq c_{3.14} n^{1-d/2} |x|^{2-d},$$

and hence

$$|G_K(x\sqrt{n})| \leq c_{4.18} n^{1-d/2} \zeta^{2-d}.$$

Similarly, if  $x, y \in n^{-1/2}\mathbb{Z}^d$ , then

$$|G_K(x\sqrt{n}) - G_K(y\sqrt{n})| \leq c_{4.17} n^{1-d/2} \left( \frac{|x-y|^{1-\beta}}{\zeta^{d-1-\beta}} + \frac{|x-y|}{\zeta^{d-1}} \right).$$

Define  $\psi(x) = n^{d/2-1} G_K(x\sqrt{n})$  for  $x \in n^{-1/2}\mathbb{Z}^d$  and define  $\psi(x)$  by some suitable interpolation procedure if  $x \notin n^{-1/2}\mathbb{Z}^d$ . Looking at the cases  $\zeta > |x-y|$  and  $\zeta \leq |x-y|$  separately, we see

$$|\psi(x) - \psi(y)| \leq c_{4.19} \left( \frac{|x-y|^{1-\beta}}{\zeta^{d-1-\beta}} \wedge \zeta^{2-d} \right).$$

Applying Lemma 4.3 to  $c_{4.19}^{-1} \zeta^{d-1-\beta} \psi$ , we get

$$II_{4.9} \leq c_{4.20} \frac{\delta^{1-\beta}}{\zeta^{d-1-\beta}}.$$

Finally, by (4.7) and (4.2)

$$III_{4.9} \leq c_{4.21} (n^{d/2-1/2} \delta \wedge n^{-\gamma/2}). \tag{4.10}$$

Looking at the cases when  $n^{(d-1)/2}$  is greater than and less than  $n^{-\gamma/2}$  separately,

$$III_{4.9} \leq c_{4.22} \delta^l.$$

Choose  $K$  so that  $\zeta = (\delta^{1-\beta} n^{\gamma/2})^{1/(\gamma+d-1-\beta)}$ . So

$$\begin{aligned} I_{4.9} + II_{4.9} + III_{4.9} &\leq c_{4.23} n^{(\gamma/2)[\gamma/(d-1+\gamma-\beta)-1]} \delta^{\gamma(1-\beta)/(d-1+\gamma-\beta)} + c_{4.22} \delta^l \\ &\leq c_{4.24} \delta^{l_\beta}, \end{aligned}$$

since  $n \geq 1$  and  $\gamma/(d-1+\gamma-\beta) - 1 < 0$ .  $\square$

### 5 Martingale calculus estimates

As in Sects. 2–4, we restrict attention to the case  $d \geq 3$ . Assume again that the conclusions (a) and (b) of Corollary 3.3 hold. Let  $\mathfrak{M}_n$  be as in Sect. 4. Fix  $\mu_n, \nu_n \in \mathfrak{M}_n$  and let

$$A_k^n = L_k^{n, \mu_n} - L_k^{n, \nu_n}, \quad U^n(x) = \mathbb{E}^x A_\infty^n .$$

There exists a mean 0 martingale  $M_k^n$  so that

$$U_k^n \equiv U^n(S_k) - U^n(S_0) = M_k^n - A_k^n .$$

Define  $B_m^n = \max_{k \leq m} |A_k^n|$ . We proceed to estimate  $\mathbb{E}^y |B_\infty^n|^2$ .

**Proposition 5.1** *There is a constant  $c_{5.1} = c_{5.1}(\mathfrak{M}_n)$  so that*

$$\sup_{n \geq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_L(\mu_n, \nu_n) \leq \delta}} \mathbb{E}^y |B_\infty^n|^2 \leq c_{5.1} \delta^{l_\beta}$$

where  $l_\beta$  is defined in (4.8).

*Proof.* Fix  $n$ . We shall temporarily drop the  $n$  superscripts. Notice that

$$\begin{aligned} |A_k|^2 &\leq 2|U_k|^2 + 2|M_k|^2 \\ &\leq 2c_{4.14}^2 \delta^{2l_\beta} + 2|M_k|^2 \quad (\text{Proposition 4.4}), \end{aligned}$$

where  $\delta = d_L(\mu, \nu)$ . Hence

$$\begin{aligned} \mathbb{E}^y |B_\infty|^2 &\leq 2c_{4.14}^2 \delta^{2l_\beta} + 2\mathbb{E}^y \sup_k |M_k|^2 \\ &\leq 2c_{4.14}^2 \delta^{2l_\beta} + 8\mathbb{E}^y |M_\infty|^2 \quad (\text{Doob's inequality}). \end{aligned} \tag{5.1}$$

But

$$\begin{aligned} |M_\infty|^2 &\leq 2|U_\infty|^2 + 2|A_\infty|^2 \\ &\leq 2c_{4.14}^2 \delta^{2l_\beta} + 2|A_\infty|^2 \quad (\text{Proposition 4.4}). \end{aligned}$$

Therefore (5.1) yields

$$\mathbb{E}^y |B_\infty|^2 \leq 18c_{4.14}^2 \delta^{2l_\beta} + 16\mathbb{E}^y |A_\infty|^2 . \tag{5.2}$$

Letting  $\Delta A_k \equiv A_{k+1} - A_k$ , note that

$$\begin{aligned} A_\infty^2 &= \sum_{k=0}^\infty (A_{k+1}^2 - A_k^2) = \sum_k \Delta A_k (A_{k+1} + A_k) \\ &= \sum_k \Delta A_k (2A_{k+1} - \Delta A_k) \\ &= 2 \sum_k A_{k+1} \Delta A_k - \sum_k (\Delta A_k)^2 . \end{aligned}$$

So

$$A_\infty^2 = 2 \sum_k (A_\infty - A_{k+1}) \Delta A_k + \sum_k (\Delta A_k)^2. \tag{5.3}$$

$$= I_{5.3} + II_{5.3}.$$

$$\begin{aligned} \mathbb{E}^y I_{(5.3)} &= \sum_k \mathbb{E}^y \{ \mathbb{E}(A_\infty - A_{k+1} | \widehat{\mathcal{F}}_{k+1}) \Delta A_k \} \quad (\widehat{\mathcal{F}}_i = \sigma\{X_1, \dots, X_i\}) \\ &= \sum_k \mathbb{E}^y \{ \mathbb{E}^{S_{k+1}} [A_\infty] \Delta A_k \} \\ &\leq 2c_{4.14} \delta^{l_\beta} \sum_k \mathbb{E}^y |\Delta A_k| \quad (\text{Proposition 4.4}) \\ &\leq 2c_{4.14} \delta^{l_\beta} \mathbb{E}^y (L_\infty^{n,\mu} + L_\infty^{n,\nu}) \\ &\leq 4c_{4.14} c_{4.1} \delta^{l_\beta}. \quad (\text{Proposition 4.1}). \end{aligned} \tag{5.4}$$

Next, using (4.2) and (4.7),

$$\sup_x |(\mu_n - \nu_n)(\{x\})| \leq c_{5.2} (\sqrt{n} \delta \wedge n^{1-(d+\gamma)/2}).$$

So

$$\begin{aligned} \mathbb{E}^y II_{5.3} &= n^{d-2} \mathbb{E}^y \sum_j [(\mu_n - \nu_n)(\{S_{j+1}/\sqrt{n}\})]^2 \\ &\leq n^{d/2-1} \sup_x |(\mu_n - \nu_n)(\{x\})| \mathbb{E}^y n^{d/2-1} \sum_j (\mu_n + \nu_n)(\{S_{j+1}/\sqrt{n}\}) \\ &\leq c_{5.2} [n^{(d-1)/2} \delta \wedge n^{-\gamma/2}] [\mathbb{E}^y L_\infty^{n,\mu_n} + \mathbb{E}^y L_\infty^{n,\nu_n}]. \end{aligned}$$

By the argument following (4.10) and Proposition 4.1,

$$\mathbb{E}^y II_{5.3} \leq c_{5.3} \delta^l. \tag{5.5}$$

Adding (5.4) to (5.5), we get Proposition 5.1.  $\square$

Using this, we prove the following exponential estimate:

**Proposition 5.2** For all  $x \in (0, \infty)$ , all  $\beta \in (0, 1)$ , and all  $\delta \leq 1$ ,

$$\sup_{n \geq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_L(\mu_n, \nu_n) \leq \delta}} \mathbb{P}^y \left\{ \sup_k |L_k^{n,\mu_n} - L_k^{n,\nu_n}| \geq x \right\} \leq 2 \exp \left\{ - \frac{x}{\sqrt{c_{5.4} \delta^{l_\beta}}} \right\}.$$

*Proof.* Define  $A^n(t) = A^n_{\lceil t \rceil}$  and  $B^n_t = \sup_{s < t} |A^n(s)|$ . Then  $t \mapsto B^n_t$  is predictable and increasing. Since  $t \mapsto B^n_t$  is also a sub-additive functional, Proposition 5.1 and the Cauchy-Schwarz inequality show that

$$\mathbb{E}^y \{ B_\infty^n - B_T^n | \mathcal{F}_T \} \leq \sqrt{c_{5.1} \delta^{l_\beta}}.$$

Therefore, by [DM] p. 193, for every  $\delta \leq 1$ , all  $x > 0$ , and  $\lambda \in (0, (c_{5.1} \delta^{l_\beta})^{-1/2}/8)$ ,

$$\sup_{n \geq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_L(\mu_n, \nu_n) \leq \delta}} \mathbb{E}^y e^{\lambda |B_\infty^n|} \leq (1 - \lambda \sqrt{c_{5.1} \delta^{l_\beta}})^{-1}.$$



Hence

$$\sup_{n \geq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_D(\mu_n, \nu_n) \leq \delta}} \mathbb{P}^{\nu} \{ |B_{\infty}^n| \geq x \} \leq e^{-\lambda x} (1 - \lambda \sqrt{c_{5.1} \delta^{l_{\beta}}})^{-1}.$$

Letting  $\lambda = 1/16 \sqrt{c_{5.1} \delta^{l_{\beta}}}$ , we get the result.  $\square$

### 6 Invariance principles

Throughout this section assume that  $X_1, X_2, \dots$  are mean zero random vectors taking values in  $\mathbb{Z}^d$  and that  $\text{Cov}(X_1) = I$ , the identity matrix. Further moment conditions will be imposed later. Let  $S_n = \sum_{j=1}^n X_j$ . Let  $\mathfrak{M}$  be a family of positive measures on  $\mathbb{R}^d$ . Suppose for each  $\mu \in \mathfrak{M}$  there exists a sequence of positive measures,  $\mu_n = \mu(n)$  converging weakly to  $\mu$ , and for each  $n$ ,  $\mu_n$  is supported on  $n^{-1/2} \mathbb{Z}^d$ . Let

$$\mathfrak{M}_n = \{ \mu(n) : \mu \in \mathfrak{M} \}.$$

*Hypothesis 6.1*

- (a) *There exists  $c_{6.1}$ , independent of  $n$ , such that  $\mu_n(\mathbb{R}^d) \leq c_{6.1}$ ,  $\mu_n \in \mathfrak{M}_n$ ;*
- (b) *for some  $\gamma > 0$ , there exists  $c_{6.2} \in (0, \infty)$ , independent of  $n$  and  $\mu$ , such that*

$$\sup_x \mu_n(B(x, s)) \leq c_{6.2} s^{d-2+\gamma} \quad \text{if } 1/2 \sqrt{n} \leq s \leq 1, n \geq 1, \mu_n \in \mathfrak{M}_n;$$

- (c $_{\beta}$ ) *there exists  $c_{6.3}$  and  $\varepsilon > 0$ , independent of  $n$ , such that if  $H_L^n$  is the metric entropy of  $\mathfrak{M}_n$  with respect to  $d_L$ , then*

$$H_L^n(x) \leq c_{6.3} x^{-(l_{\beta}/2 - \varepsilon)}, \quad H_L(x) \leq c_{6.3} x^{-(l_{\beta}/2 - \varepsilon)}, \quad x \in (0, 1).$$

In what follows we will formulate a number of invariance principles. See [Bi] for the appropriate definitions concerning weak convergence on metric spaces. But perhaps the simplest way to describe what converging weakly uniformly over a family means is to say: one can find a probability space supporting a Brownian motion  $Z_t$  and a random walk with the same distribution as the  $S_n$ 's such that  $S_{[nt]}/\sqrt{n}$  converges uniformly to  $Z_t$ ,  $t \in [0, 1]$ , a.s., and  $L_{[nt]}^{n, \mu_n}$  converges uniformly to  $L_t^{\mu}$ ,  $t \in [0, 1]$ ,  $\mu \in \mathfrak{M}$ , a.s.

*A. Subgaussian case.* In this subsection, assume  $d \geq 3$  and assume that the  $X_i$ 's are subgaussian. The following proposition follows from Proposition 5.2 just as Theorem 2.2 followed from Proposition 2.1, by standard metric entropy arguments.

**Proposition 6.2** *If Hypothesis 6.1 holds for some  $\beta \in (0, 1)$ , then for each  $\eta > 0$*

$$\limsup_{\delta \rightarrow 0} \sup_{n \geq 1} \mathbb{P} \left\{ \sup_{k \geq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_L(\mu_n, \nu_n) \leq \delta}} |L_k^{n, \mu_n} - L_k^{n, \nu_n}| \geq \eta \right\} = 0.$$

**Proposition 6.3** *If Hypothesis 6.1 holds for some  $\beta \in (0, 1)$ , then the process*

$$\{ (n^{-1/2} S_{[nt]}, L_{[nt]}^{n, \mu_n}) : 0 \leq t \leq 1, \mu \in \mathfrak{M} \}$$

*converges weakly to the process  $\{ (X_t, L_t^{\mu}) : 0 \leq t \leq 1, \mu \in \mathfrak{M} \}$ .*

*Proof.* In order to keep things as simple as possible, we will prove that  $L_{[nt]}^{n, \mu_n} \Rightarrow L_t^\mu$ . A standard modification to our argument will show the joint convergence of the local time together with the random walk.

We start by showing the convergence of the finite dimensional distributions. We give the proof for the one dimensional marginals, the general case being entirely analogous.

Define  $\varphi$  and  $\varphi_\varepsilon$  as in the proof of Lemma 4.3. Recall  $\varphi_\varepsilon * \mu_n(x) = \int \varphi_\varepsilon(x - y)\mu_n(dy)$ . Define  $\mu_n^\varepsilon$  to be the measure on  $n^{-1/2}\mathbb{Z}^d$  that puts mass  $n^{-d/2} \varphi_\varepsilon * \mu_n(\{z/\sqrt{n}\})$  on the point  $z/\sqrt{n}$ ,  $z \in \mathbb{Z}^d$ .

First, we show  $d_L(\mu, \mu * \varphi_\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ . We write

$$\begin{aligned} d_L(\mu, \mu * \varphi_\varepsilon) &= \sup_{\psi \in \mathcal{L}} \left| \int \psi(y)\mu(dy) - \int \psi(y)\mu * \varphi_\varepsilon(y) dy \right| \\ &= \sup_{\psi \in \mathcal{L}} \left| \int \psi(y)\mu(dy) - \int \psi * \varphi_\varepsilon(y) \mu(dy) \right|. \end{aligned} \tag{6.1}$$

Since  $\psi \in \mathcal{L}$ ,  $\psi * \varphi_\varepsilon$  converges uniformly to  $\psi$  as  $\varepsilon \rightarrow 0$ . Hence the right hand side of (6.1) tends to 0. A similar argument, using Hypothesis 6.1(b), shows that  $d_L(\mu_n, \mu_n * \varphi_\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ , uniformly in  $n$ .

Secondly, we calculate, using Hypothesis 6.1(b),

$$\begin{aligned} \mu_n^\varepsilon(B(x, s)) &= n^{-d/2} \sum_{|z-x| \leq s\sqrt{n}} \sum_y \varphi_\varepsilon(y/\sqrt{n}) \mu_n(\{(z-y)/\sqrt{n}\}) \\ &\leq c_{6.4} n^{-d/2} \sum_y \varphi_\varepsilon(y/\sqrt{n}) s^{d-2+\gamma} \\ &\leq c_{6.4} n^{-d/2} \varepsilon^{-d} \|\varphi\|_\infty s^{d-2+\gamma} \#\{y: y/\varepsilon\sqrt{n} \in \text{support}(\varphi)\} \\ &\leq c_{6.5} s^{d-2+\gamma}, \end{aligned} \tag{6.2}$$

if  $1/2\sqrt{n} \leq s \leq 1$ . A similar calculation shows that  $\sup_n \mu_n^\varepsilon(\mathbb{R}^d) \leq c_{6.6}$ , independently of  $n$  and  $\varepsilon$ .

Thirdly, we show that for each  $\varepsilon > 0$ ,  $\mu_n^\varepsilon$  converges to  $\mu * \varphi_\varepsilon$  uniformly on compacts, as  $n \rightarrow \infty$ . Since  $\mu_n^\varepsilon(x) = \int \varphi_\varepsilon(x - y)\mu_n(dy)$ , the  $\mu_n$  are uniformly bounded, and  $\varphi_\varepsilon$  is smooth, then  $\{\mu_n^\varepsilon: n \geq 1\}$  is an equicontinuous family of functions of  $x$ . For each fixed  $x$ ,  $\mu_n^\varepsilon(x) \rightarrow \int \varphi_\varepsilon(x - y)\mu(dy) = \mu * \varphi_\varepsilon(x)$ , since  $\mu_n \xrightarrow{w} \mu$ .

In view of (6.2), Proposition 5.2, and the fact that  $d_L(\mu_n, \mu_n * \varphi_\varepsilon) \rightarrow 0$  as  $\varepsilon \rightarrow 0$ , for each  $\eta > 0$ ,

$$\sup_n P\left(\sup_{t \leq 1} |L_{[nt]}^{n, \mu_n} - L_{[nt]}^{n, \mu_n^\varepsilon}| \geq \eta\right) \rightarrow 0$$

as  $\varepsilon \rightarrow 0$ . By (6.1) and Propositions 2.1 and 2.8, for each  $\eta > 0$ ,

$$P\left(\sup_{t \leq 1} |L_t^\mu - L_t^{\mu * \varphi_\varepsilon}| \geq \eta\right) \rightarrow 0$$

as  $\varepsilon \rightarrow 0$ . So to show  $\{L_{[nt]}^{n,\mu_n}: 0 \leq t \leq 1\}$  converges weakly to  $\{L_t^\mu: 0 \leq t \leq 1\}$ , it suffices to show that  $L_{[nt]}^{n,\mu_n} \Rightarrow L_t^{\mu * \varphi_\varepsilon}$  for each  $\varepsilon$ . But

$$L_{[nt]}^{n,\mu_n^\varepsilon} = n^{-1} \sum_{j=0}^{[nt]} \varphi_\varepsilon * \mu_n(n^{-1/2} S_j). \tag{6.3}$$

Since  $\varphi_\varepsilon * \mu_n$  converges to  $\varphi_\varepsilon * \mu$  uniformly on compacts, the desired convergence follows immediately by Donsker’s theorem.

To complete the proof, it remains to establish tightness of  $\{L_{[nt]}^{n,\mu_n}: 0 \leq t \leq 1, \mu_n \in \mathfrak{M}_n\}$ . But this follows from Proposition 6.2.  $\square$

*B.  $3 + \rho$  moments.* We still assume  $d \geq 3$ , but now only require that

$$\mathbb{E}|X_1|^{3+\rho} < \infty,$$

for some  $\rho > 0$ .

**Theorem 6.4** *If Hypothesis 6.1 holds for some  $\beta \in (0, 1)$ , then the conclusion of Theorem 6.3 is still valid.*

*Proof.* Let  $\alpha = 1/8$ ,  $a_n = n^{1/2-\alpha}$ . If  $X_j = (X_j^1, \dots, X_j^d)$ , define  $\tilde{X}_j = (\tilde{X}_j^1, \dots, \tilde{X}_j^d)$  by

$$\tilde{X}_j^i = X_j^i 1_{(|X_j^i| < a_n)}, \quad i = 1, \dots, d.$$

Let  $e_i = \mathbb{E}\tilde{X}_1^i$  and define  $X'_j$  by  $(X'_j)^i = \tilde{X}_j^i - Y_j^i$ , where  $Y_j^i$  is a random variable independent of the  $X$ ’s that takes the value  $[a_n] \operatorname{sgn}(e_i)$  with probability  $|e_i|/[a_n]$ , and the value 0 with probability  $1 - |e_i|/[a_n]$ .

Since  $\mathbb{E}X_1^i = 0$ ,

$$\begin{aligned} |e_i| &= \left| \int_{[-a_n, a_n]^c} x \mathbb{P}(X_1^i \in dx) \right| \leq 2 \int_{a_n}^\infty \mathbb{P}(|X_1^i| \geq x) dx \\ &\leq 2a_n^{-(2+\rho)} \int_{a_n}^\infty x^{2+\rho} \mathbb{P}(|X_1^i| \geq x) dx \\ &\leq c_{6.7} a_n^{-(2+\rho)} \mathbb{E}|X_1^i|^{3+\rho}. \end{aligned} \tag{6.4}$$

If  $n$  is large enough,  $|e_i| < 1$ .

Note that the  $X'_j$  are mean 0, have finite  $3 + \rho$  moments (with a bound independent of  $n$ ), have covariance close to the identity matrix, are bounded by  $2a_n$  for  $n$  large, and still take values in  $\mathbb{Z}^d$  (which is why we did not simply define  $X'$  by  $\tilde{X} - \mathbb{E}\tilde{X}$ ). We have by Chebyshev’s inequality

$$\begin{aligned} \mathbb{P}(X'_j \neq X_j) &\leq \mathbb{P}(\tilde{X}_j \neq X_j) + \sum_{j=1}^d |e_j|/[a_n] \leq c_{6.8} a_n^{-(3+\rho)} \mathbb{E}|X_j|^{3+\rho} \\ &= o(1/n). \end{aligned} \tag{6.5}$$

Let  $S'_k = \sum_{j=1}^k X'_j$ . By Bernstein’s inequality,

$$\mathbb{P}(|S'_n| \geq |x|) \leq 2 \exp\left(\frac{-|x|^2}{2n + 4a_n|x|/3}\right), \quad |x| \geq 1.$$

The expression on the right hand side is largest when  $n$  is the largest, and so if  $f(x) = \lfloor |x|^2/k \log |x| \rfloor$ ,

$$\sum_{j=1}^{f(x)} \mathbb{P}(|S'_n| \geq |x|) \leq 2f(x) \exp\left(\frac{-|x|^2}{2f(x) + 4af(x)|x|/3}\right) \leq c_{6.9}|x|^{1-d} \tag{6.6}$$

if  $k$  is large enough.

We now use (6.6) in place of (3.8), and proceeding exactly as in the proofs of Proposition 3.2 and Corollary 3.3, we conclude that

$$|G'(0, x)| \leq c_{6.10}(1 \wedge |x|^{2-d}) \tag{6.7}$$

and that for each  $\beta \in (0, 1)$ , there exists a  $c_{6.11} = c_{6.11}(\beta)$  such that

$$|G'(0, x) - G'(0, y)| \leq c_{6.11} \left( \frac{|x - y|^{1-\beta}}{(|x| \wedge |y|)^{d-1-\beta}} + \frac{|x - y|}{(|x| \wedge |y|)^{d-1}} \right), \tag{6.8}$$

where  $G'$  is defined in terms of  $X'$  just as  $G$  was defined in terms of  $X$ .

Write  $(L_k^{n, \mu_n})' = n^{d/2-1} \sum_{j=0}^{k-1} \mu_n(\{S'_j/\sqrt{n}\})$ . Then for all  $\eta > 0$ ,

$$\begin{aligned} & \mathbb{P}\left(\sup_{t \leq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_t(\mu_n, \nu_n) \leq \delta}} |L_{[nt]}^{n, \mu_n} - L_{[nt]}^{n, \nu_n}| \geq \eta\right) \\ & \leq \mathbb{P}\left(\sup_{t \leq 1} \sup_{\substack{\mu_n, \nu_n \in \mathfrak{M}_n \\ d_t(\mu_n, \nu_n) \leq \delta}} |(L_{[nt]}^{n, \mu_n})' - (L_{[nt]}^{n, \nu_n})'| \geq \eta\right) \\ & \quad + \mathbb{P}(X_j \neq X'_j \text{ for some } j \leq n). \end{aligned} \tag{6.9}$$

Recall that the results of Sects. 4 and 5 (and hence Proposition 6.2) were valid provided the conclusions (a) and (b) of Corollary 3.3 held. Therefore, using (6.7) and (6.8), the first term on the right hand side of (6.9) can be made small, uniformly in  $n$ , by taking  $\delta$  small and using Proposition 6.2 (applied to  $L'$ ). To bound the second term, we write

$$\mathbb{P}(X_j \neq X'_j \text{ for some } j \leq n) \leq n\mathbb{P}(X_1 \neq X'_1) \rightarrow 0$$

as  $n \rightarrow \infty$  by (6.5). Tightness follows readily.

The proof of the convergence of the f.d.d.'s given in Proposition 6.3 goes through without change.  $\square$

*C.  $2 + \rho$  moments.* Still assuming  $d \geq 3$ , we now assume only that  $\mathbb{E}|X_1|^{2+\rho} < \infty$ , for some  $\rho > 0$ .

**Theorem 6.5** *Suppose Hypothesis 6.1 holds with  $\beta = 1 - \rho$ . Then the conclusion of Proposition 6.3 holds.*

*Proof.* Let  $\alpha = \rho/8$ ,  $a_n = n^{1/2-d}$ . Define  $\tilde{X}, X'$  as in subsection B. As in the proof of Theorem 6.4,

$$|e_i| \leq ca_n^{-(1+\rho)} \mathbb{E}|X_1|^{2+\rho},$$

and in place of (6.5) we get

$$\mathbb{P}(X'_j \neq X_j) \leq \mathbb{P}(\tilde{X}_j \neq X_j) + \sum_{j=1}^d |e_j|/[a_n] \leq ca_n^{-(2+\rho)} \mathbb{E}|X_1|^{2+\rho} = o(1/n). \quad (6.10)$$

Using Bernsteins’s inequality, we get (6.6) as before. However,

$$\begin{aligned} \mathbb{E}|\tilde{X}_j^i|^3 &\leq 3 \int_0^{a_n} x^2 \mathbb{P}(|X| \geq x) dx \\ &\leq 3a_n^{1-\rho} \int_0^{a_n} x^{1+\rho} \mathbb{P}(|X| \geq x) dx \leq c_{6.12} a_n^{1-\rho} \end{aligned}$$

and

$$\mathbb{E}|Y_j^i|^3 = [a_n]^3 |e_i|/[a_n] \leq c_{6.13} a_n^{1-\rho}.$$

So

$$\mathbb{E}|X'_j|^3 \leq c_{6.14} a_n^{1-\rho}.$$

Hence in Proposition 3.1 we can only conclude

$$\sup_{x \in \mathbb{Z}^d} |\mathbb{P}^x(S'_n = 0) - (2\pi n)^{-d/2} e^{-|x|^2/2n}| \leq c_{6.15} n^{-(d+\rho)/2-\varepsilon}$$

for some small  $\varepsilon > 0$ . Using this estimate in (3.11),

$$\sum_{n \geq f(x)} E(n, x) \leq c_{6.16} |x|^{-d-\rho+2}. \quad (6.11)$$

Following the proofs of Proposition 3.2 and Corollary 3.3, but using (6.6) in place of (3.8) and (6.11) in place of (3.11), we get (6.7) and (6.8) with  $\beta = 1 - \rho$ .

As in the  $3 + \rho$  moment case, using (6.10), we get tightness. No changes are needed to the proofs of the convergence of the f.d.d.’s.  $\square$

*D. Second moments.* When the  $X_j$ ’s have only finite second moments, our methods do not give uniform invariance principles. But we still can prove the convergence of the f.d.d.’s when  $d = 3$ .

**Theorem 6.6** *Suppose  $d = 3$ ,  $\mathbb{E}|X_j|^2 < \infty$ , and Hypothesis 6.1(a), (b) hold. For measures  $\mu^1, \dots, \mu^N \in \mathfrak{M}$ ,*

$$(L_{[m]}^{n, \mu^i} : 0 \leq t \leq 1, i = 1, \dots, N)$$

*converges weakly to  $(L_t^{\mu^i} : 0 \leq t \leq 1, i = 1, \dots, N)$ .*

*Proof.* We give the argument for  $N = 1$ , the general case being analogous. Examining the proof of Proposition 6.3, we see that we need only show that for each  $\eta > 0$

$$\mathbb{P}\left(\sup_{k \leq n} |L_k^{n, \mu_n} - L_k^{n, \mu_n^\varepsilon}| \geq \eta\right) \rightarrow 0 \quad (6.12)$$

as  $\varepsilon \rightarrow 0$ , uniformly for  $n \geq n_0(\eta)$ .

Let  $\theta > 0$ ,  $\zeta_n = \theta n^{1/2}$ , and  $K_n = \zeta_n^{2-d} = (\theta n^{1/2})^{2-d}$ . As in the proof of Proposition 4.4, define

$$\psi_n(x) = n^{d/2-1} G_{K_n}(xn^{1/2})$$

for  $x \in n^{-1/2} \mathbb{Z}^d$  and by a suitable interpolation procedure for  $x \notin n^{-1/2} \mathbb{Z}^d$ . Write  $v_n$  for  $\mu_n^\varepsilon$ . By Spitzer [S],  $|G(0, z)| \leq c_{6.17}(1 \wedge |z|^{2-d})$  and  $G(0, z) = g(0, z)(1 + o(1))$ , as  $|z| \rightarrow \infty$ . So given  $a$ , there exists  $M_1$  such that  $|G(0, x) - g(0, x)| \leq a|x|^{2-d}$  if  $x \in \mathbb{Z}^d$ ,  $|x| \geq M_1$ . Hence if  $|w|, |z| \geq M_1$ ,  $w, z \in \mathbb{Z}^d$ ,

$$|G(0, w) - G(0, z)| \leq \frac{c_{6.18}|w - z|}{(|w| \wedge |z|)^{d-1}} + \frac{2a}{(|w| \wedge |z|)^{d-2}}. \tag{6.13}$$

Now  $|\psi_n(y) - \psi_n(z)|$  will be largest if  $|y|, |z| \geq c_{6.19}\theta$ ,  $y, z \in n^{-1/2} \mathbb{Z}^d$  for some constant  $c_{6.19}$  independent of  $n$ . So suppose  $|y|, |z| \geq c_{6.19}\theta$ . Then for  $n$  large enough,  $|y|\sqrt{n}, |z|\sqrt{n} \geq M_1$ . So by (6.13), for  $n$  large enough,

$$\begin{aligned} |\psi_n(y) - \psi_n(z)| &\leq n^{d/2-1} \left[ \frac{c_{6.18}|y - z|\sqrt{n}}{(|y| \wedge |z|)^{d-1}n^{(d-1)/2}} + \frac{2a}{(|y| \wedge |z|)^{d-2}n^{(d-2)/2}} \right] \\ &\leq \frac{c_{6.20}|y - z|}{\theta^{d-1}} + \frac{2a}{\theta^{d-2}}. \end{aligned} \tag{6.14}$$

Let  $b > 0$ . Choose  $\theta$  small enough so that  $\theta^\gamma < b$ . Since the sequence  $\{\mu_n\}$  is tight, we choose  $M_2$  large so that  $\mu_n(B(0, M_2)^c) \leq b$ . By the estimate (6.14),

$$\begin{aligned} |\psi_n(y) - \psi_n * \varphi_\varepsilon(y)| &\leq \int |\psi_n(y) - \psi_n(y - \varepsilon x)| \varphi(x) dx \\ &\leq \frac{c_{6.20} \varepsilon |x|}{\theta^{d-1}} + \frac{2a}{\theta^{d-2}} \leq b \end{aligned}$$

if we take  $a$  and  $\varepsilon$  small, and  $n$  sufficiently large. Therefore,

$$|\int \psi_n(y)(\mu_n - v_n)(dy)| = |\int [\psi_n(y) - \psi_n * \varphi_\varepsilon(y)] \mu_n(dy)| \leq c_{6.21} b.$$

As in the proof of Proposition 4.4 (see (4.9)),

$$|\mathbb{E}^x L_\infty^{n, \mu_n} - \mathbb{E}^x L_\infty^{n, v_n}| \leq \theta^\gamma + |\int \psi_n(y)(\mu_n - v_n)(dy)| + c_{6.22}(d_L(\mu_n, v_n))^l. \tag{6.15}$$

So taking  $\varepsilon$  smaller if necessary, we can make the right hand side of (6.15) less than  $(2 + c_{6.21})b$ . Plugging the estimate (6.15) into the proof of Proposition 5.2 and using Chebyshev's inequality, we get finally

$$\begin{aligned} \mathbb{P}^0 \left( \sup_k |L_k^{n, \mu_n} - L_k^{n, \mu_n^\varepsilon}| \geq \eta \right) &\leq \eta^{-2} \mathbb{E}^0 \left[ \sup_k |L_k^{n, \mu_n} - L_k^{n, \mu_n^\varepsilon}|^2 \right] \\ &\leq c_{6.23} \eta^{-2} b \end{aligned}$$

if  $n$  is sufficiently large, which is precisely what we wanted.  $\square$

E.  $d = 1, 2$ . The results for  $d = 1, 2$  follow by the usual projection argument.

**Theorem 6.7** *Theorems 6.4, 6.5, and 6.6 hold for  $d = 1$  and 2.*

*Proof.* Fix  $M > 0$ . Given  $\mu$  defined on  $\mathbb{R}^d$ ,  $d = 1$  or 2, define  $\hat{\mu}$  on  $\mathbb{R}^3$  by

$$\hat{\mu}(A \times B) = \mu(A) |B \cap B(0, M)|, \quad A \subseteq \mathbb{R}^d, B \subseteq \mathbb{R}^{3-d},$$

$B(0, M)$  the ball in  $\mathbb{R}^{3-d}$ . Similarly, given  $\mu_n$  defined on  $n^{-1/2} \mathbb{Z}^d$ , define  $\hat{\mu}_n$  on  $n^{-1/2} \mathbb{Z}^3$ .

Define  $\widehat{X}_j = (X_j, Y_j)$ , where  $Y_j$  is simple random walk on  $\mathbb{Z}^{3-d}$ , independent of the  $X_j$ 's. Define

$$\widehat{L}_k^{n, \widehat{\mu}_n} = n^{1/2} \sum_{j=0}^{k-1} \widehat{\mu}_n(\{\widehat{S}_j/\sqrt{n}\}) .$$

Then by Theorem 6.4 or 6.6,  $\widehat{L}_k^{n, \widehat{\mu}_n}$  converges weakly to  $\widehat{L}_t^{\widehat{\mu}}$ , where  $\widehat{L}$  is the additive functional associated to  $\widehat{\mu}$ .

But it is clear that for all  $\mu_n$ ,  $\widehat{L}_k^{n, \widehat{\mu}_n} = L_k^{n, \mu_n}$  up until the first time  $n^{-1/2} |\sum_{i=1}^k Y_j|$  exceeds  $M$ , and for all  $\mu$ ,  $\widehat{L}_t^{\widehat{\mu}} = L_t^{\mu}$  up until the first time  $(3-d)$ -dimensional Brownian motion exceeds  $M$  in absolute value. Since  $M$  is arbitrary, the weak convergence of  $L_{[nt]}^{n, \mu_n}$  to  $L_t^{\mu}$  follows easily.  $\square$

### 7 Examples

#### A Classical additive functionals – $L^p$ functionals

Suppose  $p > d/2$ , and  $p^{-1} + q^{-1} = 1$ . Let  $\mathfrak{F}$  be a subset of  $\{f \in L^p(B(0, 1)) : f \geq 0\}$ . Let  $H_p$  denote the metric entropy of  $\mathfrak{F}$  with respect to  $d_p(f_1, f_2) = \|f_1 - f_2\|_p$ . Note in what follows we do not assume our  $f$ 's are continuous.

**Theorem 7.1** *If  $\sup_{f \in \mathfrak{F}} \|f\|_p < \infty$  and the exponent of metric entropy of  $H_p$  is less than  $1/2$ , then  $\int_0^t f(Z_s) ds$  is jointly continuous in  $t \in [0, 1]$  and  $f \in \mathfrak{F}$  (with respect to the  $d_p$  metric.)*

*Proof.* Here  $\mathfrak{M} = \{\mu : \mu \text{ has a density } f(x) \text{ with respect to Lebesgue measure, } f \in \mathfrak{F}\}$ , and  $L_t^{\mu} = \int_0^t f(Z_s) ds$ . By Hölder's inequality,

$$\mu(\mathbb{R}^d) = \int_{B(0, 1)} f(x) dx \leq c_{7.1} \|f\|_p ,$$

and  $\mu(B(x, s)) = \int_{B(0, 1)} 1_{B(x, s)}(y) f(y) dy \leq \|1_{B(x, s)}\|_q \|f\|_p \leq c_{7.2} s^{d/q}$ , for  $s \leq 1$  and  $f \in \mathfrak{F}$ .

So the total mass of the  $\mu$ 's is uniformly bounded and the index of  $\mathfrak{M}$  is  $d/q - d + 2 = 2 - d/p > 0$ . If  $\mu(dx) = f(x) dx$  and  $\nu(dx) = h(x) dx$ , then

$$d_G(\mu, \nu) = \sup_x \left| \int g(0, x) [f(x) - h(x)] dx \right| \leq \|f - h\|_p \|g(0, \cdot)\|_q \leq c_{7.3} d_p(f, h) ,$$

since  $g \in L^q(B(0, 1))$  when  $p > d/2$ .

Our result now follows by Theorem 2.2.  $\square$

Since changing  $f$  on a set of measure 0 does not affect  $L_t^{\mu}$  (here  $\mu(dx) = f(x) dx$ ), but can have a drastic effect on  $n^{-1} \sum_j f(n^{-1/2} S_j)$ , for an invariance principle one must have some additional regularity for  $f$  (cf. the next example).

#### B Classical additive functionals – indicators

Let  $\mathfrak{A}$  be a subset of  $\{A : A \subseteq B(0, 1)\}$ . Suppose that for almost every  $y \in \mathbb{R}^d$ ,  $r \in (0, 1]$ , and  $A \in \mathfrak{A}$ , as  $n \rightarrow \infty$ ,

$$n^{-d/2} \# \{n^{-1/2} \mathbb{Z}^d \cap A \cap B(y, r)\} \rightarrow |A \cap B(y, r)| . \tag{7.1}$$

Define  $d_S(A, B) = |A \Delta B|$ .

**Theorem 7.2** *Suppose the  $X_i$  satisfy the assumptions of sect. 6 and have  $2 + \rho$  moments. Let  $\beta = 1 - \rho$ , take  $\gamma = 2$ , and let  $l_\beta$  be defined by (4.8). Suppose  $\mathfrak{A}$  satisfies (7.1) and the exponent of metric entropy of  $\mathfrak{A}$  with respect to  $d_S$  is less than  $l_\beta/2$ . Then  $n^{-1} \sum_{i=0}^{[nt]} 1_A(S_j/\sqrt{n})$  converges weakly to  $\int_0^t 1_A(Z_s) ds$ , uniformly over  $t \in [0, 1]$  and  $A \in \mathfrak{A}$ .*

*Proof.* For  $A \in \mathfrak{A}$ , define  $\mu_A$  by  $\mu_A(dx) = 1_A(dx)$ . Define  $\mu_{A,n}$  by  $\mu_{A,n}(\{n^{-1/2}x\}) = n^{-d/2} 1_A(n^{-1/2}x)$ . That  $\mu_{A,n}$  converges to  $\mu_A$  follows by (7.1) and [Bi]. That Hypothesis 6.1(a) and (b) hold is easy. Hypothesis 6.1(c) follows from the crude estimate

$$|\int \psi(x)[1_A(x) - 1_B(x)] dx| \leq |A \Delta B|, \quad \psi \in \mathfrak{Q},$$

and a similar formula for  $d_L(\mu_{A,n}, \mu_{B,n})$ . Now apply Theorem 6.5.  $\square$

*C Local times on curves*

This example works for hypersurfaces in  $\mathbb{R}^d$  for any dimension  $d$ , but for simplicity we restrict ourselves to  $d = 2$  and the curves of form

$$C = \{(t, f(t)): t \in [0, 1], \|f\|_\infty \leq c_{7.4}\}. \tag{7.2}$$

We will use  $C$  to denote the graph of  $C$ . Let  $\mathfrak{C}$  be a collection of such curves. Let  $\mu_C(A) = |\{t: (t, f_C(t)) \in A\}|$ .

For such  $C \in \mathfrak{C}$ , we let  $f_{C,n}$  be a function from  $[0, 1]$  to  $[-2c_{7.4}, 2c_{7.4}]$ , such that  $f_{C,n}$  takes values in  $n^{-1/2}\mathbb{Z}$ , has jumps only at  $t$ 's in  $n^{-1/2}\mathbb{Z}$ , and  $f_n \rightarrow f$  in  $L^1$ -norm. Denote the curve and graph of  $\{(t, f_{C,n}(t)): t \in [0, 1]\}$  by  $C_n$ . If  $C^1$  and  $C^2$  denote two curves of the form (7.2) (corresponding to  $f_1$ , and  $f_2$ , resp.), let  $d_C(f_1, f_2) = \|f_{C^1} - f_{C^2}\|_1$ .

**Theorem 7.3** *Suppose the  $X_i$  satisfy the assumptions of sect. 6 and have  $2 + \rho$  moments. Suppose that for some  $c_{7.4}$  and  $\epsilon$  independent of  $n$*

$$H_C^n(x) \leq c_{7.4} x^{-(l_\beta/2 - \epsilon)}, \quad H_C(x) \leq c_{7.4} x^{-(l_\beta/2 - \epsilon)}, \quad x \in (0, 1),$$

where  $H_C^n(x)$  (resp.  $H_C(x)$ ) is the metric entropy of  $\mathfrak{C}$  (resp.  $\mathfrak{C}_n = \{C_n: C \in \mathfrak{C}\}$ ) with respect to  $d_C$ . Then  $n^{-1/2} \sum_{i=0}^{[nt]} 1_{C_n}(n^{-1/2}S_j)$  converges weakly to  $L_t^{\mu_C}$ , uniformly over  $t \in [0, 1]$ ,  $C \in \mathfrak{C}$ .

*Proof.* Define  $\mu_{C,n}(A) = n^{-1/2} \#\{k \leq \sqrt{n}: (n^{-1/2}k, f_{C,n}(n^{-1/2}k)) \in A\}$ . Since  $f_{C,n} \rightarrow f$  in  $L^1$ ,  $\mu_{C,n} \xrightarrow{w} \mu_C$ . Note that  $\mu_{C,n}(\mathbb{R}^d) \leq 1$ , while

$$\mu_{C,n}(B(x, r)) \leq c_{7.5} r,$$

so the index of  $\mathfrak{C}_n$  is 1. The result follows from Theorem 6.5.  $\square$

*D Local times in  $\mathbb{R}^1$*

Even for local times in  $\mathbb{R}^1$ , our results are fairly strong. For  $x \in \mathbb{R}^1$ , let  $\mu_x$  be the point mass at  $x$ . Then  $L_t^{\mu_x}$  (usually written as  $L_t^x$ ) is just local time at  $x$ . Clearly the  $\mu_x$  are uniformly bounded with index 1. By Example 2.4,  $H_L(\delta) \leq c_{7.6} |\log(\delta)|$ .



Define  $\Gamma(n, x)$  to be  $n^{-1/2}$  times the unique integer lying in the interval  $[x\sqrt{n}, x\sqrt{n+1}]$ .

**Theorem 7.4** *If the  $X_i$  have finite  $2 + \rho$  moments for some  $\rho > 0$ , and are as in Sect. 6, then  $n^{-1/2} \sum_{j=1}^{[nt]} 1_{[\sqrt{nx}, \sqrt{nx+1})}(S_j)$  converge weakly to  $L_t^x$ , uniformly over all levels  $x$ .*

*Proof.* It suffices to prove the result uniformly over  $x \in [-M, M]$  for each  $M$ . Define  $\mu_{n,x}$  to be point mass at  $\Gamma(n, x)$ . Clearly  $\mu_{n,x} \xrightarrow{w} \mu_x$  as  $n \rightarrow \infty$ , the  $\mu_{n,x}$  are uniformly bounded, have index 1, and entropy is of order  $|\log(x)|$ . The result follows by Theorem 6.5 and the observation that  $S_j \in [\sqrt{nx}, \sqrt{nx+1}]$  if and only if  $\mu_{n,x}(\{n^{-1/2}S_j\}) = 1$ .  $\square$

The question of invariance principles for local time has a long history, dating back to [CH]. Using techniques highly specific to one-dimensional Brownian motion, Borodin [Bo1] has proved Theorem 7.4 when the  $X_i$ 's have finite second moments. For a slightly different notion of local time, [P] has a uniform invariance principle if the  $X_i$  have  $1 + \sqrt{3} \approx 2.732$  moments. See [BK2] for results and references to the corresponding strong invariance principles.

*E Fractals*

For simplicity, we confine ourselves to  $d = 2$  and fractals of the following form: let  $F_0 = [0, 1]^2$ , let  $F_1$  be the union of  $R$  closed squares with sides of size  $a$ , such that the interiors are pairwise disjoint. To form  $F_2$ , replace each of the squares making up  $F_1$  by replicas of  $F_1$ , and continue.

To be more precise, if  $S$  is any square, let  $\Psi_S$  be the orientation preserving affine map that takes  $S$  to  $F_0$ . Let

$$F_2 = \bigcup \{ \Psi_S^{-1}(F_1) : S \text{ is one of the } R \text{ squares with sides of size } a \text{ making up } F_1 \},$$

$$F_{k+1} = \bigcup \{ \Psi_S^{-1}(F_1) : S \text{ is one of the } R^k \text{ squares with sides of size } a^k \text{ making up } F_k \},$$

Let  $F = \bigcap_{k=0}^{\infty} F_k$ .

For example, if  $F_1 = [0, 1]^2 - (1/3, 2/3)^2$ ,  $F$  will be the Sierpinski carpet. If  $F_1 = ([0, 1/3] \cup [2/3, 1])^2$ , we get the 2-dimensional Cantor set.

Let  $\mu$  be the Hausdorff-Besicovitch measure on  $F$ , normalized to have total mass 1.

**Theorem 7.5** *If the Hausdorff dimension of  $F > 0$ ,*

$$\frac{1}{|F_n|} \int_0^t 1_{F_n}(Z_s) ds \xrightarrow{\text{a.s.}} L_t^\mu .$$

*Remark 7.6* The convergence in probability is a consequence of results in [B].

*Proof.* It is not hard to see that  $\sup_x \mu(B(x, s)) \leq c_{7.7} s^\gamma$ , where  $\gamma$  is the Hausdorff dimension of  $F$ .

Suppose  $\psi \in \mathcal{Q}$ , and let  $S_1, \dots, S_R$  be the squares making up  $F_1$ . Let  $\mu_n(dx) = |F_n|^{-1} 1_{F_n}(x) dx$ . Let  $x_i$  be the lower left corner of  $S_i$ . Since  $\mu_2, \mu_1$  both have total mass 1,

$$\begin{aligned} \int_{S_i} \psi(x) [\mu_2(dx) - \mu_1(dx)] &= |F_1|^{-1} \int_{S_i} [\psi(x) - \psi(x_i)] [|F_1|^{-1} 1_{F_2}(x) - 1_{F_1}(x)] dx \\ &= |F_1|^{-1} a^2 \int_{F_0} \psi_i(x) [\mu_1(dx) - \mu_2(dx)], \end{aligned} \tag{7.2}$$

where  $\psi_i(x) = \psi \circ \Psi_{S_i}^{-1}(x) - \psi \circ \Psi_{S_i}^{-1}(x_i)$ . Since  $|\nabla \psi_i(x)| \leq a$ , and  $\psi_i(0) = 0$ , then  $\|\psi_i\|_\infty \leq \sqrt{2}a$ . So the right hand side of (7.2) is bounded above by  $|F_1|^{-1} \sqrt{2}a^3 d_L(\mu_0, \mu_1)$ . Summing over  $i$ , and taking the supremum over  $\psi \in \mathcal{Q}$ ,

$$d_L(\mu_2, \mu_1) \leq \sqrt{2}a^3 R |F_1|^{-1} d_L(\mu_2, \mu_1) = \sqrt{2}ad_L(\mu_1, \mu_0).$$

By an induction argument,

$$d_L(\mu_{k+1}, \mu_k) \leq (\sqrt{2}a)^k d_L(\mu_1, \mu_0). \tag{7.3}$$

If  $R = 1$ , so that  $F_1$  is a single square, then  $F$  is a single point. This case is ruled out by the assumption that the dimension of  $F$  is strictly bigger than 0. So  $R > 1$ , and hence  $a < 1/2$ .

Let  $\mathfrak{M} = \{\mu_n\}_{n=1}^\infty \cup \{\mu\}$ . To cover  $\mathfrak{M}$  with  $d_L$ -balls of radius  $\delta$ , first put a ball  $B$  of radius  $\delta$  around  $\mu$ . Since  $\mu_k \xrightarrow{w} \mu$ , (7.3) shows that  $d_L(\mu_k, \mu) \leq c_{7.8}(\sqrt{2}a)^k$ . So  $B$  covers all but  $|\log(\delta/c_{7.8})/\log(\sqrt{2}a)| + 1$  of the  $\mu_n$ 's. So at most  $c_{7.9}|\log(\delta)|$  balls are needed, hence  $H_L(\delta) \approx |\log|\log(\delta)||$ .

By Theorem 2.2,  $L_t^v$  is continuous with respect to  $d_L$ , for  $v \in \mathfrak{M}$ . This implies our result.  $\square$

*F Intersection local time – double points*

Let  $S_n^1$  and  $S_n^2$  be two independent identically distributed random walks converging in law to two independent Brownian motions,  $Z_t^1$ , and  $Z_t^2$ . By redefining these processes on a suitable probability space, we may assume that the convergence is almost sure.

Define  $\mu_{u,x}(A) = |\{t \in [0, u] : Z_t^2 + x \in A\}|$ . In [BK1], it is shown that  $\alpha(x, s, u) = L_s^{\mu_{u,x}}$  is the intersection local time for  $(Z^1, Z^2)$ . Let us consider the corresponding invariance principle. We discuss the case  $d = 3$  first. (If  $d \geq 4$ , the paths of  $Z^1$  and  $Z^2$  do not intersect.)

If  $x = (x^1, x^2)$ , let  $\Gamma_2(n, x) = (\Gamma(n, x^1), \Gamma(n, x^2))$ , where  $\Gamma$  is defined in subsection  $D$ . Define

$$\mu_{u,x,n}(A) = n^{-1} \sum_{k=1}^{[nu]} 1_A(S_j^2/\sqrt{n} + \Gamma_2(n, x)).$$

**Lemma 7.7** *There exists  $\gamma > 0$  such that for each  $M$ , with probability one,*

$$\mu_{u,x,n}(B(y, s) - \{y\}) \leq c_{7.10} s^{1+\gamma}, \quad x, y \in B(0, M), s \leq 1,$$

where  $c_{7.10}$  depends on  $M$  and  $\omega$ .

*Proof.* For simplicity, we prove this when  $x = 0$ , the general case being similar.

$$\begin{aligned} \mathbb{E}^z \mu_{\infty,0,n}(B(y,s) - \{y\}) &\leq n^{-1} \sum_{w \neq y} G(z,w) 1_{B(y,s)}(n^{-1/2}w) \\ &\leq n^{-1} \sum_{k=0}^{\infty} \sum_{2^k \leq |w-z| < 2^{k+1}} 2^{-k} n^{-1/2} \\ &\quad \# \{B(y\sqrt{n}, s\sqrt{n}) \cap \mathbb{Z}^d \cap [B(z, 2^{k+1}) - B(z, 2^k)]\} \\ &\leq c_{7.11} s^{1+\gamma}, \end{aligned}$$

for  $\gamma = 1/2$ .

This estimate is uniform in  $z$ , hence the potential of  $\mu_{\infty,0,n}(B(y,s) - \{y\})/a$  is bounded above by 1, where  $a = \sup_z \mathbb{E}^z \mu_{\infty,0,n}(B(y,s) - \{y\})$ . By [DM], p. 193,

$$\mathbb{P}^z \{ \mu_{\infty,0,n}(B(y,s) - \{y\}) > c_{7.11} s^{9/8} \} \leq c_{7.12} \exp(-c_{7.13} s^{-1/8}).$$

For each  $k$ , we can choose  $N_k = c_{7.14} 2^{3k}$  balls, each of radius  $2^{-k+2}$ , so that for every  $y \in B(0, M)$  and every  $s \leq 2^{-k+1}$ ,  $B(y, s)$  is covered by one of these  $N_k$  balls. Hence,

$$\begin{aligned} \mathbb{P}^z \{ \mu_{\infty,0,n}(B(y,s) - \{y\}) > c_{7.11} s^{9/8} \text{ for some } y \in B(0, M) \\ \text{and some } s \in [2^{-k}, 2^{-k+1}] \} \\ \leq N_k c_{7.12} \exp(-c_{7.13} 2^{k/8}). \end{aligned}$$

Summing over  $k$  and using the Borel–Cantelli lemma, we conclude

$$\sup_{\substack{y \in B(0, M) \\ 0 < s \leq 1}} \mu_{\infty,0,n}(B(y,s) - \{y\}) \leq c_{7.14} s^{9/8}, \quad \text{a.s.} \quad \square$$

**Theorem 7.8** *Let  $X_i^1, X_i^2$  be two independent sequences of i.i.d. r.v.’s, identically distributed, and satisfying the assumptions of sect. 6 with  $2 + \rho$  moments. If  $d = 3$ ,  $L_{[ns]}^{n, \mu_{u,n,x}}$  converges weakly to  $\alpha(x, s, u)$ , uniformly over  $x \in \mathbb{R}^3, s, u \in [0, 1]$ .*

*Proof.* We apply Theorem 6.5. Since  $\sup_{y,n} \mathbb{P}^y \{ \sup_{j \leq n} |S_j^2|/\sqrt{n} \geq M \} \rightarrow 0$  as  $M \rightarrow \infty$ , it suffices to look at the  $\mu_{u,n,x}$  restricted to  $B(0, M)$ .

For each  $u$ , the metric entropy of  $\{ \mu_{u,x,n} : x \in B(0, M) \}$  is bounded above by  $c_{7.15} \delta^{-3}$ . For each  $x$ , the total variation of  $\mu_{u_2,x,n} - \mu_{u_1,x,n}$  is bounded above by  $u_2 - u_1$ . So Hypothesis 6.1 ( $c_\beta$ ) holds for every  $\beta > 0$ . Hypothesis 6.1(a) is clear and 6.1(b) is Lemma 7.7.

Since  $S_n^2/\sqrt{n}$  converges uniformly to  $Z_t^2, \mu_{u,n,x} \xrightarrow{w} \mu_{u,x}$ . The result follows.  $\square$

To handle the case  $d = 2$ , we use the projection technique of sect. 6E, and get Theorem 7.8 for the case  $d = 2$  as well.

For both the  $d = 2$  and  $d = 3$  cases, weak convergence at a single level  $x$  follows by Theorem 6.6 or 6.7 under the assumption of finite variance only.

### G Intersection local time – multiple points

In [BK1], we gave a method for constructing intersection local time for the intersection of  $k + 1$  independent Brownian motions in  $\mathbb{R}^2$  from the intersection

local time of  $k$  independent planar Brownian motions. A completely analogous construction can be made for the number of intersections of  $k$  random walks. We then can get the analogue of Theorem 7.8: for  $d = 2$  only, the number of intersections converges weakly to the  $k$ -tuple intersection local time, uniformly over all the variables, provided the  $X$ 's have  $2 + \rho$  moments. As in the proof of Theorem 7.8, the only work is in finding the index of the family of measures, and as in [BK1], the estimates needed for  $k + 1$ -intersection local time follow from those obtained for  $k$ -intersection local time.

For multiple points, we cannot use a projection argument, and must work with 2-dimensional random walks killed off at a geometric rate. So it is necessary to rework the results of sect. 3 for  $d = 2$  with  $G$  replaced by the  $\lambda$ -resolvent of  $S_n$ . We leave the (numerous) details to the interested reader.

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