

## Introduction

### Motivation

- Many stochastic processes arising in applications exhibit a range of possible behaviours depending the values of certain key parameters.
- The problem originate from a conjecture by *Paul Erdős* [2].
- The case of *simple symmetric random walk* is solved by Grill [2].
- We extend the result for a much larger class of random walk, with only *minor* moment assumptions.
- An Application in Physics is the *random polymer chain model*. The growth process is repelled or attracted by the centre of mass, depending if it is a poor or good solvent [1].

### Strong law of large numbers

Using standard techniques on functional limit theorems, we get the following *strong law of large numbers*.

**Proposition 1** (L., Wade, 2017). *If (μ) holds, then, as  $n \rightarrow \infty$ ,*

$$n^{-1}G_n \rightarrow \frac{1}{2}\mu, \text{ a.s.}$$

### Central limit theorem

With the help of Lindeberg–Feller theorem for triangular arrays, we have the following *central limit theorem*.

### The centre of mass of a random walk

- Let  $d \geq 1$ . Let  $X, X_1, X_2, \dots$  be a sequence of i.i.d. random variables on  $\mathbb{R}^d$ .

- Consider the random walk  $(S_n, n \in \mathbb{Z}_+)$  in  $\mathbb{R}^d$  defined by

$$S_0 := \mathbf{0} \quad \text{and} \quad S_n := \sum_{i=1}^n X_i \quad (n \geq 1).$$

- Our object of interest is the *centre of mass process*  $(G_n, n \in \mathbb{Z}_+)$  corresponding to the random walk, defined by

$$G_0 := \mathbf{0} \quad \text{and} \quad G_n := \frac{1}{n} \sum_{i=1}^n S_i \quad (n \geq 1).$$

### Asymptotic analysis

**Proposition 2** (L., Wade, 2017). *If (M) holds, then, as  $n \rightarrow \infty$ ,*

$$n^{-1/2} \left( G_n - \frac{n}{2} \mu \right) \xrightarrow{d} \mathcal{N}_d(\mathbf{0}, M/3).$$

### Local central limit theorem

For  $\mathbf{x} \in \mathbb{R}^d$ , define  $p_n(\mathbf{x}) := \mathbb{P}(n^{-1/2}G_n = \mathbf{x})$ , and

$$n(\mathbf{x}) := \frac{\exp\{-\frac{3}{2}\mathbf{x}^\top M^{-1}\mathbf{x}\}}{(2\pi)^{d/2} \sqrt{\det(M/3)}}.$$

### Notation and assumptions

Throughout we use the notation

$$\mu := \mathbb{E}X, \quad M := \mathbb{E}[(X - \mu)(X - \mu)^\top]$$

whenever the expectations exist; when defined,  $M$  is a symmetric  $d$  by  $d$  matrix. Now we need the following moment assumptions to proceed.

**(μ)** Suppose that  $\mathbb{E}\|X\| < \infty$ .

**(M)** Suppose that  $\mathbb{E}\|X\|^2 < \infty$  and  $M$  is positive-definite.

For our main results, we assume that  $X$  has a lattice distribution.

**(L)** Suppose that  $X$  is non-degenerate. Suppose that for a constant vector  $\mathbf{b} \in \mathbb{R}^d$  and a  $d$  by  $d$  matrix  $H$  with  $|\det H| = h > 0$ , we have  $\mathbb{P}(X \in \mathbf{b} + H\mathbb{Z}^d) = 1$ .

Also define

$$\mathcal{L}_n := \left\{ n^{-3/2} \left( \frac{1}{2}n(n+1)\mathbf{b} + H\mathbb{Z}^d \right) \right\}.$$

Our first main result is a *local* central limit theorem.

**Theorem 1** (L., Wade, 2017). *Suppose that (M), (L), and some technical assumptions hold. Then, as  $n \rightarrow \infty$ , we have*

$$\lim_{n \rightarrow \infty} \sup_{\mathbf{x} \in \mathcal{L}_n} \left| \frac{n^{3d/2}}{h} p_n(\mathbf{x}) - n \left( \mathbf{x} - \frac{(n+1)}{2n^{1/2}} \mu \right) \right| = 0. \quad (1)$$

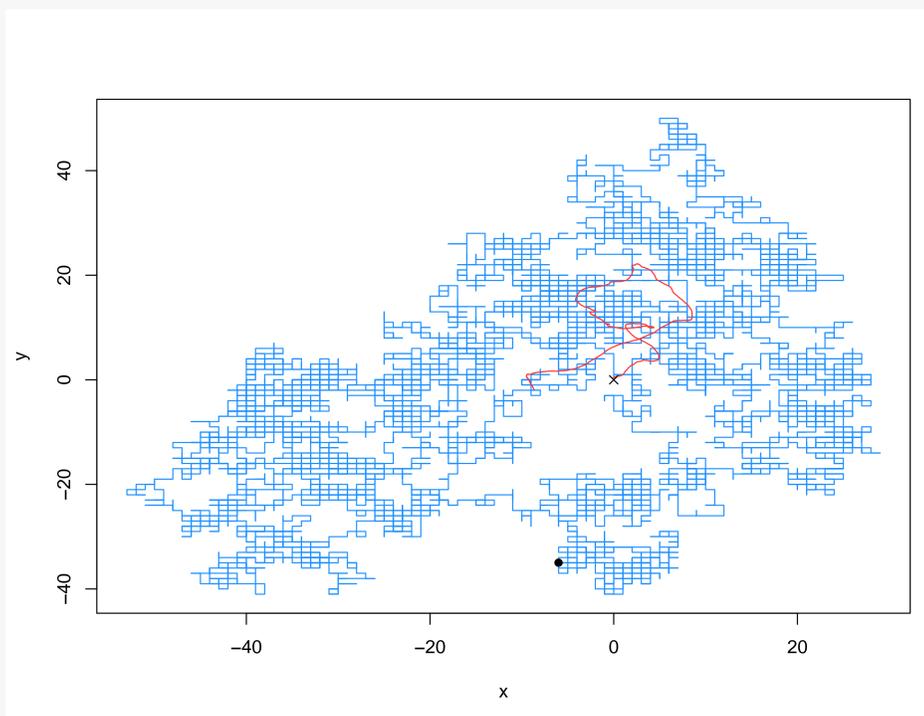
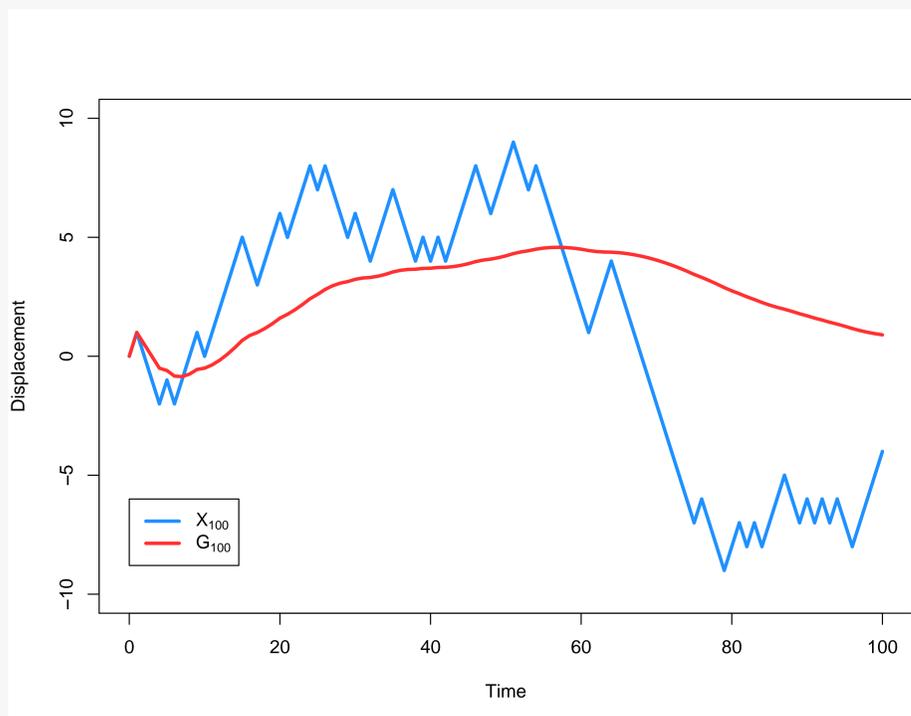


Figure 1: Two simulations of the centre of mass (red) and the corresponding random walk (blue) in one dimension, 100 steps (left) and two dimensions, 10000 steps (right)

## Recurrence classification

### One dimension

- Depending on different moment assumptions, we can get very different recurrence behaviour of the process. First we give a recurrence result in one dimension.
- In the case of SSRW the fact that  $G_n$  returns infinitely often to a neighbourhood of the origin is due to Grill [2, Theorem 1].

**Theorem 2** (L., Wade, 2017). *Suppose that  $d = 1$  and that either of the following two conditions holds.*

(i) *Suppose that  $\mathbb{E}|X| \in (0, \infty)$  and  $X \stackrel{d}{=} -X$ .*

(ii) *Suppose that (M) holds and that  $\mathbb{E}X = 0$ .*

*Then we have  $\liminf_{n \rightarrow \infty} G_n = -\infty$ ,  $\limsup_{n \rightarrow \infty} G_n = +\infty$  and  $\liminf_{n \rightarrow \infty} |G_n| = 0$ .*

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### References

- [1] F. Comets, M.V. Menshikov, S. Volkov, and A.R. Wade, Random walk with barycentric self-interaction, *J. Stat. Phys.* **143** (2011) 855–888.
- [2] K. Grill, On the average of a random walk, *Statist. Probab. Lett.* **6** (1988) 357–361.
- [3] I.A. Ibragimov and Y.V. Linnik, *Independent and Stationary Sequences of Random Variables*, Wolters-Noordhoff, Groningen, The Netherlands, 1971.

On the other hand, if the first moment does not exist,  $G_n$  may be transient. The condition we assume is as follows.

**(S)** Suppose that  $X \stackrel{d}{=} -X$  and  $X$  is in the domain of normal attraction of a symmetric  $\alpha$ -stable distribution with  $\alpha \in (0, 1)$ .

**Theorem 3** (L., Wade, 2017). *Suppose that  $d = 1$  and (L) holds, i.e.,  $\mathbb{P}(X \in b + h\mathbb{Z}^d) = 1$  for  $b \in \mathbb{R}$  and  $h > 0$ . Under some technical conditions we have  $\lim_{n \rightarrow \infty} |G_n| = \infty$ .*

### Two dimensions or more

- We have the following transience result in dimensions greater than one.
- We gives a diffusive rate of escape, implying  $\lim_{n \rightarrow \infty} \|G_n\| = +\infty$ .
- In the case of SSRW the result is due to Grill [2, Theorem 1].

**Theorem 4** (L., Wade, 2017). *Suppose that  $d \geq 2$  and that (M), (L), and*

*some technical condition hold, and that  $\mu = \mathbf{0}$ . Then*

$$\lim_{n \rightarrow \infty} \frac{\log \|G_n\|}{\log n} = \frac{1}{2}, \text{ a.s.}$$

### A conjecture

- Obtaining necessary and sufficient conditions for recurrence and transience of  $G_n$  is an open problem.
- For  $d \geq 2$ , we believe that  $G_n$  is always ‘at least as transient’ as the situation in Theorem 4:

**Conjecture 1** (L., Wade, 2017). *Suppose that  $\text{supp } X$  is not contained in a one-dimensional subspace of  $\mathbb{R}^d$ . Then*

$$\liminf_{n \rightarrow \infty} \frac{\log \|G_n\|}{\log n} \geq \frac{1}{2}, \text{ a.s.}$$