

Inequalities for critical exponents in d -dimensional sandpiles

Sandeep Bhupatiraju* Jack Hanson† Antal A. Járai‡

Abstract

Consider the Abelian sandpile measure on \mathbb{Z}^d , $d \geq 2$, obtained as the $L \rightarrow \infty$ limit of the stationary distribution of the sandpile on $[-L, L]^d \cap \mathbb{Z}^d$. When adding a grain of sand at the origin, some region, called the avalanche cluster, topples during stabilization. We prove bounds on the behaviour of various avalanche characteristics: the probability that a given vertex topples, the radius of the toppled region, and the number of vertices toppled. Our results yield rigorous inequalities for the relevant critical exponents. In $d = 2$, we show that for any $1 \leq k < \infty$, the last k waves of the avalanche have an infinite volume limit, satisfying a power law upper bound on the tail of the radius distribution.

Keywords: Abelian sandpile; critical exponent; wave; uniform spanning tree; loop-erased random walk.

AMS MSC 2010: Primary 60K35, Secondary 82B20.

Submitted to EJP on February 21, 2016, final version accepted on September 20, 2017.

Supersedes arXiv:1602.06475.

Contents

1 Introduction	2
2 Definitions and background	8
2.1 Graphs and sandpiles	8
2.2 Toppling numbers and waves	10
2.3 Spanning trees and the burning bijection	11
2.4 Intermediate configurations	12
2.5 Bijection for intermediate configurations	13
2.6 Random walk notation and basic facts	15
2.7 Wilson’s algorithm	17
2.8 Wiring o to the boundary	17

*Indiana University, USA. The research of S.B. was supported by NSF grants DMS-1007244 and DMS-1612363. E-mail: sbhupati@indiana.edu

†City College of New York, USA. The research of J.H. was supported by an AMS-Simons travel grant and NSF grant DMS-1612921. E-mail: jhanson@ccny.cuny.edu

‡University of Bath, United Kingdom. E-mail: A.Jarai@bath.ac.uk

3 Toppling probability bounds in low dimensions	18
3.1 Preliminary setup	18
3.2 SRW and LERW steering	20
3.2.1 Estimates for a single walk	20
3.2.2 Separation lemma	21
3.2.3 Estimate on $\Gamma_{z,L}$	23
3.3 Proof of Theorem 1.1	25
3.3.1 Proof of Theorem 1.1 when $d = 2$	25
3.3.2 Proof of Theorem 1.1 when $d = 3$	25
3.3.3 Proof of Theorem 1.1 when $d = 4$	26
4 Low-dimensional radius bounds	26
5 The last k waves in 2D	27
6 High-dimensional radius bounds	35
6.1 Radius bounds on transitive unimodular graphs	35
6.2 Radius lower bound when $d \geq 5$	36
6.3 Radius upper bound when $d \geq 5$	37
7 Bounds on the size	38
7.1 Upper bounds on the size when $d = 3, 4$	38
7.2 Lower bounds on the size when $2 \leq d \leq 4$	39
7.3 Proof of the moment bounds in $d = 4$	41
7.3.1 Proof of the first moment bounds	41
7.3.2 Proof of the second moment bounds	43
7.4 The size of the past in invariant forests	48
7.5 Lower bounds: $d \geq 5$	48
7.6 Upper bounds: $d \geq 5$	49
References	50

1 Introduction

The Abelian sandpile model is a particle system defined in terms of simple local redistribution events, called topplings, which give rise to non-local dynamical events called avalanches. The model has received a lot of attention in the theoretical physics literature (see [9, 3]) due to its remarkable self-organized critical state, conjectured to be characterized by power-law behavior of various quantities related to avalanches. Starting with the seminal work of Dhar, much mathematical progress has been made toward understanding this self-organized critical state. The surveys [33] and [14] collect some of this. However, establishing power law behavior for many fundamental avalanche characteristics on \mathbb{Z}^d appears difficult in general. The purpose of this paper is to establish new rigorous inequalities, which in high dimensions come close to identifying the correct tail behavior, for these quantities.

Given a finite set $V \subset \mathbb{Z}^d$, a **sandpile** on V is a collection of indistinguishable particles, given by a map $\eta : V \rightarrow \{0, 1, \dots\}$. We say that η is **stable**, if $\eta(x) < 2d$ for all $x \in V$. If η is unstable at x , that is $\eta(x) \geq 2d$, we say that x is **allowed to topple**. On toppling, x sends one particle along each edge incident with it, resulting in the new

sandpile

$$\eta'(y) = \begin{cases} \eta(y) - 2d & \text{if } y = x; \\ \eta(y) + 1 & \text{if } y \sim x, y \in V; \\ \eta(y) & \text{otherwise.} \end{cases} \quad (1.1)$$

Particles sent to vertices in $\mathbb{Z}^d \setminus V$ are lost. It is well-known [8] that given any sandpile η , carrying out all possible topplings in any sequence, results in a uniquely defined stable sandpile η° . The **sandpile Markov chain** on V is the Markov chain with state space equal to the set of stable sandpiles on V , where at each time step a particle is added at a uniformly chosen vertex of V , and the sandpile is stabilized, if necessary. The unique stationary distribution [8] is denoted ν_V .

We will be interested in sandpiles on \mathbb{Z}^d , where “stable” and “toppling” are defined the same way as for finite V . Athreya and Járai [2] proved that if $V(L) = [-L, L]^d \cap \mathbb{Z}^d$, $d \geq 2$, then $\nu_L := \nu_{V(L)}$ converges weakly, as $L \rightarrow \infty$, to a limit measure ν , called the **sandpile measure**. Let $\eta : \mathbb{Z}^d \rightarrow \{0, \dots, 2d - 1\}$ be a sample configuration from the measure ν . Let us add a particle to η at the origin o , and let $\text{Av} = \text{Av}(\eta)$ denote the set of vertices that topple, called the **avalanche cluster**. The set of all topplings, with multiplicity, is called the **avalanche**. In this paper we study various characteristics of avalanches.

The concept of waves, introduced by Ivashkevich, Ktitarev and Priezzhev [13] in the context of finite graphs, will play an important role. Waves provide a decomposition of an avalanche into smaller sets of topplings: $\mathcal{W}_{L,1}, \dots, \mathcal{W}_{L,N} \subset V(L)$; see Section 2 for precise definitions. In each wave, every vertex topples at most once, and the union of the waves includes all topplings of the avalanche with the correct multiplicity. The paper [13] analyzed the last wave $\mathcal{W}_{N,L}$ in particular when $d = 2$.

Our first set of results concern the probability that a given vertex topples. Based on an analysis of the last wave, we prove the following rigorous lower bounds on the **toppling probability**.

Theorem 1.1.

(i) Let $d = 2$. Then

$$\nu(z \in \text{Av}) \geq |z|^{-3/4+o(1)}, \quad \text{as } |z| \rightarrow \infty.$$

(ii) Let $d = 3$. There exist constants $\zeta < 1/2$ and $c > 0$ such that

$$\nu(z \in \text{Av}) \geq c|z|^{-2\zeta-1}, \quad \forall z \in \mathbb{Z}^3.$$

(iii) Let $d = 4$. There exists a constant $c > 0$ such that

$$\nu(z \in \text{Av}) \geq c|z|^{-2} (\log |z|)^{-1/3}, \quad \forall z \in \mathbb{Z}^4.$$

In Theorem 1.1 (ii), ζ can be taken to be any value such that a random walk in \mathbb{Z}^3 of length n does not hit the loop-erasure of an independent random walk of length n with probability $\geq c'n^{-\zeta}$ for some $c' > 0$. We prefer to write the bound (ii) in this form to emphasize the dependence on this exponent, whose value is of interest in the theory of loop-erased walks. The exponent ζ is known to satisfy the bound $\zeta < 1/2$; see [24, Sections 10.3 and 11.5].

The rigorous upper bound $\nu(z \in \text{Av}) \leq C|z|^{2-d}$, for some $C = C(d)$, follows from Dhar’s formula (see [15, Eqn. (3.5)]). In dimensions $d \geq 5$, Járai, Redig and Saada [16, Section 6.2] proved that $\nu(z \in \text{Av}) \geq c|z|^{2-d}$, for some $c = c(d)$, also based on an analysis of the last wave. [16] introduced the critical exponent θ to quantify the departure from Dhar’s formula, assuming that $\nu(z \in \text{Av}) \approx |z|^{2-d-\theta}$, as $|z| \rightarrow \infty$. This

means that $\theta = 0$ when $d \geq 5$. Our Theorem 1.1 shows that if θ exists in the sense that $\lim \log(\nu(z \in \text{Av}))/\log(|z|)$ exists as $|z| \rightarrow \infty$ (“logarithmic equivalence”), then

$$0 \leq \theta \begin{cases} \leq 3/4 & \text{when } d = 2, \\ < 1 & \text{when } d = 3, \\ = 0 & \text{when } d = 4. \end{cases}$$

In particular, Theorem 1.1(iii) establishes that $\theta = 0$ when $d = 4$, with at most a logarithmic correction.

The reason behind the fact that $\theta = 0$ for $d \geq 5$ is that in these dimensions loop-erased walk and independent simple random walk do not intersect with positive probability. The difference in behavior when $d \geq 5$ also shows up in our other results, and $d = 4$ is expected to be the upper critical dimension of the model, in the sense that critical exponents are no longer expected to depend on dimension when $d \geq 5$ [32]. We expect that θ is positive in dimensions two and three, in analogy with other statistical physics models below the upper critical dimension. However, it seems difficult to get rigorous upper bounds improving on Dhar’s formula, since any such bound would have to control all waves of the avalanche. For the last wave, we have a precise characterization in terms of loop-erased walk; however, we lack a convenient description of the joint distribution of all waves of the avalanche. For similar reasons, we do not expect the bounds coming from last waves in low dimensions to be tight.

Our next set of results concern the radius of the toppled region. Let $R = R(\eta) = \sup\{|z| : z \in \text{Av}(\eta)\}$ be the **radius** of the avalanche. As we explain below, some of the following inequalities are easy consequences of Theorem 1.1, while some others follow from known results on uniform spanning forests of \mathbb{Z}^d .

Theorem 1.2.

(i) Let $d = 2$. Then,

$$r^{-3/4+o(1)} \leq \nu(R \geq r), \quad \text{as } r \rightarrow \infty.$$

(ii) Let $d = 3$. There are constants $c > 0$ and C such that with ζ as in Theorem 1.1 we have

$$cr^{-(2\zeta+1)} \leq \nu(R \geq r) \leq Cr^{-1/6}, \quad \forall r \geq 1.$$

(iii) Let $d = 4$. Then there exist constants $c > 0$ and C such that

$$cr^{-2}(\log r)^{-1/3} \leq \nu(R \geq r) \leq Cr^{-1/4}, \quad \forall r \geq 1.$$

(iv) Let $d \geq 5$. There is a constant $c = c(d) > 0$ such that

$$cr^{-2} \leq \nu(R \geq r) \leq r^{-2}(\log r)^{3+o(1)}, \quad \forall r \geq 1.$$

The lower bounds of Theorem 1.2 in dimensions 2, 3, and 4 follow from taking $z = re_1$ in Theorem 1.1, where $e_1 = (1, 0, \dots, 0) \in \mathbb{Z}^d$. In dimensions $d \geq 3$, upper bounds can be derived from results of Lyons, Morris and Schramm [25]. They analyzed, using the “conductance martingale” of Morris [30], the wired uniform spanning forest measure **WSF** on transient graphs, including \mathbb{Z}^d for $d \geq 3$, as well as a related measure **WSF**_{*o*}, obtained by “wiring *o* to infinity”. See the book [26] for detailed background on wired spanning forests. Let \mathfrak{F}_o denote the component of *o* under **WSF**_{*o*}. The proof of [25, Theorem 4.1] shows that for $d \geq 3$,

$$\mathbf{WSF}_o(\text{diam}(\mathfrak{F}_o) > r) \leq C(d)r^{-\frac{1}{2}+\frac{1}{d}}. \tag{1.2}$$

The measure \mathbf{WSF}_o can be related to waves in sandpiles; in particular, this was used by Járai and Redig [15] to show that when $d \geq 3$, avalanches are finite ν -a.s. We derive the upper bounds in Theorem 1.2(ii)–(iii) from (1.2).

Above the critical dimension, $d \geq 5$, Priezzhev [32] gave heuristic arguments for the mean-field behaviour $\nu(R \geq r) \approx r^{-2}$. Both the lower bound $\nu(R \geq r) \geq \nu(re_1 \in Av)$ and the upper bound of (1.2) can be sharpened to establish this rigorously, in the sense of logarithmic equivalence. On the other hand, Theorem 1.2(ii)–(iii) establishes that, if a critical exponent α satisfying $\nu(R > r) \approx r^{-\alpha}$ governs the tail of R in low dimensions, then this α is different from the mean-field value 2.

We deduce the lower and upper bounds in Theorem 1.2(iv) from very general mass transport arguments, stated in Theorem 1.3 below; see [26, Chapter 8] for background on mass transport. While the main focus of this paper is sandpile models on \mathbb{Z}^d , we believe this result may be useful on other graphs and for other models. The proof is in Section 6.1, and is independent of the rest of the paper. Let $G = (V, E)$ be a graph and let $\Gamma \subset \text{Aut}(G)$ be a transitive subgroup of the group of automorphisms of G , under the topology of pointwise convergence. It is well known that every closed subgroup of $\text{Aut}(G)$ has a Borel measure which is invariant under the left multiplication by $\gamma \in \Gamma$. The group Γ is called unimodular if this measure is also invariant under right multiplication. In addition, we call the graph G unimodular if $\text{Aut}(G)$ has some unimodular transitive closed subgroup. In this setting the mass transport principle states that for $o \in V(G)$ and a non-negative function $f : V \times V \rightarrow [0, \infty]$, which is invariant under the diagonal action of Γ , we have $\sum_{x \in V} f(o, x) = \sum_{x \in V} f(x, o)$. Let d be a Γ -invariant metric on V , and write $\text{diam}(A; x) = \sup\{d(v, x) : v \in A\}$, and let $\text{diam}(A) = \text{diam}(A; o)$. Write $D_x(r) = \{y \in V : d(y, x) \leq r\}$. We say that an infinite tree T has **one end**, if any two infinite self-avoiding paths in T have a finite symmetric difference. Given $x \in T$, we denote by past_x the set of vertices $y \in T$ such that the unique infinite self-avoiding path in T starting at y contains x . By a **percolation** on (V, E) , we mean a probability measure on subgraphs of (V, E) . Given a percolation, we write \mathfrak{C}_x for the connected component of x . When the percolation is supported on spanning forests, we write \mathfrak{F}_x for \mathfrak{C}_x .

Theorem 1.3. *Let (V, E) be a graph with a transitive unimodular group of automorphisms Γ , and let $o \in V$ be a fixed vertex. Let μ be a Γ -invariant percolation on (V, E) .*

(i) *If μ is supported on spanning forests with one-ended components, then*

$$\mu(\text{diam}(\text{past}_o) > r) \geq \sum_{x \in V: r < d(x, o) \leq 2r} \frac{\mu(o \in \text{past}_x)^2}{\mathbb{E}_\mu \left[|\mathfrak{F}_o \cap D_o(4r)| \mathbf{1}_{o \in \text{past}_x} \right]}.$$

(ii) *We have*

$$\mu(\text{diam}(\mathfrak{C}_o) > 4r) = \sum_{x \in V: r < d(x, o) \leq 4r} \mu(o \in \mathfrak{C}_x) \mathbb{E}_\mu \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{C}_x; x) > 4r}}{|\mathfrak{C}_x \cap D_x(4r) \setminus D_x(r)|} \middle| o \in \mathfrak{C}_x \right].$$

(iii) *Suppose that $\mathbf{WSF}_o(|\mathfrak{F}_o| < \infty) = 1$. Then*

$$\begin{aligned} & \mathbf{WSF}_o(\text{diam}(\mathfrak{F}_o) > 4r) \\ &= \sum_{x \in V: r < d(x, o) \leq 4r} \mathbf{WSF}_x(o \in \mathfrak{F}_x) \mathbb{E}_{\mathbf{WSF}_x} \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{F}_x; x) > 4r}}{|\mathfrak{F}_x \cap D_x(4r) \setminus D_x(r)|} \middle| o \in \mathfrak{F}_x \right]. \end{aligned}$$

Regarding the upper bound in Theorem 1.2 (iv), Lyons, Morris and Schramm state the result

$$\mathbf{WSF}(\text{diam}(\text{past}_o) > r) \leq r^{-2}(\log r)^{O(1)}; \tag{1.3}$$

see [25, page 1710]. However, since a proof of (1.3) is not included in [25], and we need a sharpening of (1.2) for our results, we deduce a diameter estimate for \mathfrak{F}_o under

\mathbf{WSF}_o from Theorem 1.3(iii). (This implies (1.3) due to a stochastic comparison; see [25, Lemma 3.2]). In order to deal with the fact that \mathbf{WSF}_o is not translation invariant, we restrict attention to \mathfrak{T}_o , which is unimodular; see Section 6.1.

We do not have an upper bound on $\nu(R \geq r)$ in $d = 2$, and it is an open problem whether $\nu(R < \infty) = 1$. It follows from Theorem 1.2(i) that $\mathbb{E}_\nu R = \infty$, when $d = 2$. It may be of independent interest that a short proof of the weaker statement, that $\mathbb{E}_{\nu_L} R$ diverges, can be given without reference to spanning trees or the burning bijection described in Section 2.3. We state this as a separate result.

Proposition 1.4. *If $d = 2$, then $\lim_{L \rightarrow \infty} \mathbb{E}_{\nu_L} R = \infty$.*

Our last set of results concern the number of topplings in the avalanche. Let S denote the total number of topplings in the avalanche (that is, elements of Av are counted with multiplicity). Recall that $\mathcal{W}_{N,L}$ denotes the last wave in the finite graph V_L . Based on the fractal dimension of loop-erased walk and scaling assumptions, Ivashkevich, Kvitarev and Priezzhev [13] derived the exponent $\nu_L(|\mathcal{W}_{N,L}| \geq t) \approx t^{-3/8}$, in the limit $L \rightarrow \infty$. We prove a rigorous lower bound with the same exponent, which we also extend to higher dimensions. Above the critical dimension, $d \geq 5$, we also have an upper bound on the total number of topplings with an exponent which is independent of d . The upper bounds of Theorem 1.2 on the radius in $d = 3, 4$ provide upper bounds on the size of the avalanche cluster.

Since the size of the avalanche cluster could be measured in two different ways – namely, via $|\text{Av}|$ and via S – there are in principle two different possible critical exponents $\tau_{S'}$ and τ_S given (if they exist) by $\nu(|\text{Av}| > t) \approx t^{-\tau_{S'}}$, $\nu(S > t) \approx t^{-\tau_S}$. As in the preceding cases, our theorems give corresponding bounds on the possible values of $\tau_S, \tau_{S'}$. These bounds, as well as the best current bounds on the exponents θ and α described above, are summarized in Table 1.

Theorem 1.5.

(i) *Let $d = 2$. Then*

$$t^{-3/8+o(1)} \leq \nu(|\text{Av}| \geq t), \quad \text{as } t \rightarrow \infty.$$

(ii) *Let $d = 3$. With ζ as in Theorem 1.1, and for some constants C and $c > 0$, we have*

$$ct^{-(2\zeta+1)/3} \leq \nu(|\text{Av}| \geq t) \leq Ct^{-1/18}, \quad \forall t \geq 1.$$

Moreover, $\nu(S \geq t) \leq Ct^{-1/19}, \forall t \geq 1$.

(iii) *Let $d = 4$. There exists C and $c > 0$ such that*

$$ct^{-1/2}(\log t)^{-5/6} \leq \nu(|\text{Av}| \geq t) \leq Ct^{-1/16}, \quad \forall t \geq 1.$$

Moreover, $\nu(S \geq t) \leq Ct^{-1/17}, \forall t \geq 1$.

(iv) *Let $d \geq 5$. There exist $c = c(d) > 0$ such that*

$$ct^{-1/2} \leq \nu(|\text{Av}| \geq t) \leq \nu(S \geq t) \leq t^{-2/5+o(1)}, \quad \forall t \geq 1.$$

We establish the lower bounds in dimensions $d = 2, 3$ by showing that once a vertex at distance $t^{1/d}$ from o is in the last wave, at least ct other vertices in its neighbourhood will also be in the last wave. In $d = 4$, ct is replaced by $ct/\log t$. Given this, parts (i)–(iii) of Theorem 1.5 can be deduced from parts (i)–(iii) of Theorem 1.1. For the lower bound in $d \geq 5$, in Theorem 1.5(iv), we use the following analogue of Theorem 1.3(i). Write

$$\mathfrak{T}_o(r) = \mathfrak{T}_o \cap D_o(r) \quad \text{and} \quad \tilde{\mathfrak{T}}_o(r) = \left\{ x \in \mathfrak{T}_o(r) : \begin{array}{l} \text{the path from } o \text{ to } x \text{ in} \\ \mathfrak{T}_o \text{ stays inside } D_o(r) \end{array} \right\}.$$

	Toppling probability	Radius	Avalanche cluster size	Avalanche size
	$\nu(x \in \text{Av})$ $\approx x ^{2-d-\theta}$	$\nu(R > r)$ $\approx r^{-\alpha}$	$\nu(\text{Av} > t)$ $\approx t^{-\tau_{S'}}$	$\nu(S > t)$ $\approx t^{-\tau_S}$
$d = 2$	[0, 3/4]	[0, 3/4]	[0, 3/8]	[0, 3/8]
$d = 3$	[0, 1)	[1/6, 2)	[1/18, 2/3)	[1/19, 2/3)
$d = 4$	= 0	[1/4, 2]	[1/16, 1/2]	[1/17, 1/2]
$d \geq 5$	= 0	= 2	[2/5, 1/2]	[2/5, 1/2]

Table 1: The best known bounds on the critical exponents introduced in this introduction. Bounds are expressed in interval form: e.g., $1/6 \leq \alpha < 2$ when $d = 3$.

Theorem 1.6. *Let (V, E) be a graph with a transitive unimodular group of automorphisms Γ . Let μ be a Γ -invariant percolation on (V, E) supported on spanning forests with one-ended components. For all $t, r \geq 1$ we have*

$$\mu(|\text{past}_o| > t) \geq \sum_{x:r < d(x,o) \leq (3/2)r} \frac{\mu(o \in \text{past}_x, |\tilde{\mathfrak{F}}_o(r/2)| > t)^2}{\mathbb{E}[|\mathfrak{F}_o(4r)| \mathbf{1}_{\{o \in \text{past}_x\}}]}$$

The proof of Theorem 1.6 is in Section 7.4 and does not rely on the rest of the paper. We believe, as it has been argued by Priezzhev [32], that the exponent 1/2 is sharp in Theorem 1.5(iv). But it seems challenging to establish a matching upper bound for S or Av . The main difficulty lies in extracting useful information on the dependence between the waves from the bijection with WSF_o . Instead of such an approach, we control the number of waves using our upper bound on the radius in Theorem 1.2(iv), which allows us to use a union bound instead of estimating the dependence. This leads to the upper bounds in Theorem 1.5(ii)–(iv).

All our results have analogues in large finite $V(L)$, or indeed are derived therefrom. Passing to the limit of \mathbb{Z}^d is not too difficult when $d \geq 3$, due to the result of Jarai and Redig [15, Theorem 3.11] showing that $\nu(|S| < \infty) = 1$. When $d = 2$, this is not known. We bypass this problem with a more technical argument, that we believe is of interest in its own right. We show that for any $1 \leq k < \infty$, the **last k -waves** (when they exist) have a finite limit as $V(L) \uparrow \mathbb{Z}^d$. Recall that for $\eta \in \mathcal{R}_L$ the waves occurring during the stabilization of $\eta + \mathbf{1}_o$ are denoted $\mathcal{W}_{L,1}, \dots, \mathcal{W}_{L,N}$. Waves can also be defined on \mathbb{Z}^d , denoted $\mathcal{W}_1, \mathcal{W}_2, \dots$; see Section 2.2. On \mathbb{Z}^d , the number of waves N may take the value infinity.

Let η_{N-k+1} denote a random configuration on $V(L)$ with law $\nu_L(\cdot | N \geq k)$. Given a configuration η on $V(L)$ such that $\eta(o) = 2d$, let $\mathcal{W}(\eta)$ denote the set of sites that can be toppled with every site toppling at most once. We extend this also to configurations ξ on \mathbb{Z}^d such that $\xi(o) \geq 2d$.

Theorem 1.7. *Assume $d = 2$.*

(i) *For all $k \geq 1$, the law of η_{N-k+1} converges weakly to the law of a random configuration ξ_k in \mathbb{Z}^2 . Let ρ_k denote the law of ξ_k . The law of $\mathcal{W}_{N-k+1} = \mathcal{W}(\eta_{N-k+1})$ converges to the law of $\mathcal{W}_k^* := \mathcal{W}(\xi_k)$, that is a.s. finite under ρ_k .*

- (ii) For all $k \geq 0$ we have $\lim_{L \rightarrow \infty} \nu_L(N = k) = \nu(N = k)$.
- (iii) For every $k \geq 1$, the joint law of $\mathcal{W}_{L,1}, \dots, \mathcal{W}_{L,k}$ under $\nu_L(\cdot | N = k)$, converges weakly, as $L \rightarrow \infty$, to the law of $\mathcal{W}_1, \dots, \mathcal{W}_k$ under $\nu(\cdot | N = k)$, and under this conditioning we have ν -a.s. $|\mathcal{W}_\ell| < \infty$, $\ell = 1, \dots, k$.

See Section 5 for more detailed statements. Our argument in fact gives a power law upper bound on the radii of the last k waves, and leads to the following extension of Theorem 1.2(i). Let $R_k = \sup\{|x| : x \in \mathcal{W}_k^*\}$.

Theorem 1.8. Assume $d = 2$.

- (i) There are constants $\alpha_1 > \alpha_2 > \dots > 0$ and C_1, C_2, \dots such that

$$\rho_k(R_k > r) \leq C_k r^{-\alpha_k}, \quad \forall r \geq 1, \forall k \geq 1.$$

- (ii) We have

$$\nu(R \geq r, N \leq k) \leq C_k r^{-\alpha_k}, \quad \forall r \geq 1.$$

We will also use Theorem 1.7 to prove the following theorem.

Theorem 1.9. Suppose $d = 2$. Then $\mathbb{E}_\nu N = \infty$.

This is a strengthening of the statement $\mathbb{E}_\nu R = \infty$, due to a simple comparison proved in Lemma 2.3, and therefore also of Proposition 1.4.

Organization of the paper. In Section 2, we give definitions and background on sandpiles, spanning trees, and random walks; we also prove Proposition 1.4. In Section 2.2, we prove Theorem 1.1 modulo a technical argument required for the two-dimensional case, which we defer to Section 5.

Sections 4, 5, and 6 are devoted to the various radius bounds above. In Section 4, we prove Theorem 1.2 (i) – (iii). Section 5 contains additional arguments for the two-dimensional case; here we prove Theorems 1.7, 1.8, and 1.9. In Section 6, we complete the proof of Theorem 1.2 by proving the high-dimensional bounds (Theorem 1.2 (iv) and Theorem 1.3).

Section 7 contains the proofs of the size bounds above: Theorems 1.5 and 1.6.

A note on constants. All our constants will be positive and finite, and they may depend on the dimension d . Other dependence will always be indicated. Constants denoted C and c may change from line to line; those with index (such as c_1) stay the same within the same proof.

2 Definitions and background

In this section, we provide definitions and collect useful facts about the basic objects we use: toppling numbers, waves, spanning trees, bijections and Wilson’s algorithm.

2.1 Graphs and sandpiles

We will work with finite connected graphs of the form $H = (U \cup \{s\}, F)$, where s is a distinguished vertex, called the **sink**. We allow multiple edges, so in general H is a multigraph, but we exclude loop-edges. If $U = V$ is a finite subset of \mathbb{Z}^d , we let $G_V = (V \cup \{s\}, E)$ denote the **wired subgraph** induced by V — i.e., where all vertices in $\mathbb{Z}^d \setminus V$ are identified to the single vertex s , and loops at s are removed. Of prime importance will be the **standard exhaustion** $V(L) = [-L, L]^d \cap \mathbb{Z}^d$, $L \geq 1$; the wired subgraph induced by $V(L)$ will be denoted G_L . In general, a subscript L will be shorthand for subscript $V(L)$. We write $x \sim y$ to denote that vertices x and y of a graph (understood from context) are neighbours. Given a graph $H = (U \cup \{s\}, F)$, we write $\deg_H(x)$ for the degree of the vertex $x \in U \cup \{s\}$ in H . When $V \subset \mathbb{Z}^d$, we write $\deg_V(x)$ for the degree of vertex $x \in V$ in the subgraph of \mathbb{Z}^d induced by V .

Given $H = (U \cup \{s\}, F)$, the discrete Dirichlet Laplacian Δ_H is given by

$$\Delta_H(x, y) = \begin{cases} \deg_H(x) & \text{if } x = y; \\ -a_{xy} & \text{if } x \neq y; \end{cases} \quad x, y \in U, \tag{2.1}$$

where a_{xy} equals the number of edges connecting x and y . In particular, when V is a finite subset of \mathbb{Z}^d , we have Δ_V given by

$$\Delta_V(x, y) = \begin{cases} 2d & \text{if } x = y; \\ -1 & \text{if } x \sim y; \\ 0 & \text{otherwise,} \end{cases} \quad x, y \in V. \tag{2.2}$$

We denote the inverse matrix by $g_H = (\Delta_H)^{-1}$, $g_V = (\Delta_V)^{-1}$.

We write Δ for the matrix defined as in (2.2), but with $x, y \in \mathbb{Z}^d$, and $g = \Delta^{-1}$ when $d \geq 3$. Up to a factor $(2d)^{-1}$, these matrices are the Green function of simple random walk. Namely, let $(S(n))_{n \geq 0}$ denote a simple random walk in \mathbb{Z}^d , and let

$$\sigma_A = \inf\{n \geq 0 : S(n) \notin A\}.$$

Then $2d g_V(\cdot, \cdot) = G_V(\cdot, \cdot)$ and $2d g(\cdot, \cdot) = G(\cdot, \cdot)$, where we define

$$G_V(x, y) := \mathbb{E}_x \left[\sum_{0 \leq n < \sigma_V} \mathbf{1}_{S(n)=y} \right], \quad x, y \in V;$$

$$G(x, y) := \mathbb{E}_x \left[\sum_{0 \leq n < \infty} \mathbf{1}_{S(n)=y} \right], \quad x, y \in \mathbb{Z}^d.$$

Whenever G appears with arguments, it refers to a Green function as defined above; when it appears without arguments, it refers to graphs as in the notation introduced previously.

Let us fix a finite connected graph $H = (U \cup \{s\}, F)$. A **sandpile** on H is a function $\eta : U \rightarrow \{0, 1, 2, \dots\}$. We say that η is **unstable** at $x \in U$, if $\eta(x) \geq \deg_H(x)$. In this case x is **allowed to topple**, which means that x sends one particle along each edge incident with it. Particles arriving at s are lost. Toppling x has the effect of subtracting row $\Delta_H(x, \cdot)$ from $\eta(\cdot)$. It is a basic property of the model that if unstable vertices are toppled in any order until there are no such vertices, the stable sandpile obtained is independent of the order chosen (called the Abelian property) [8]. Hence for any sandpile η there is a well-defined **stabilization** of η , denoted η° . The sandpile Markov chain is defined as follows. The state space is $\Omega_H = \prod_{x \in U} \{0, \dots, \deg_H(x) - 1\}$. Given that the current state is η , a single step is defined by choosing a vertex $X \in U$ uniformly at random, and moving to state $(\eta + \mathbf{1}_X)^\circ$. The maps $a_x : \Omega_H \rightarrow \Omega_H$ defined by $a_x : \eta \mapsto (\eta + \mathbf{1}_x)^\circ$, $x \in U$, are called the **addition operators**. It follows from the uniqueness of stabilization that $a_x a_y = a_y a_x$, $x, y \in U$. When it is necessary to emphasize the graph H on which the operator a_x is applied, we write $a_{x,H}$ for a_x ; we also use $a_{x,L}$ for a_{x,G_L} . We denote the set of recurrent states of the sandpile Markov chain by \mathcal{R}_H . It is known that the unique stationary distribution is given by the uniform distribution on \mathcal{R}_H [8], and that each (restricted) map $a_x : \mathcal{R}_H \rightarrow \mathcal{R}_H$ preserves this measure.

In the special case when the graph arises from a finite $V \subset \mathbb{Z}^d$, we denote the state space by $\Omega_V = \{0, \dots, 2d - 1\}^V$, and the set of recurrent states by \mathcal{R}_V .

Now, let $\eta : \mathbb{Z}^d \rightarrow \{0, 1, 2, \dots\}$ be a sandpile on \mathbb{Z}^d . If x is unstable in η , we define the toppling of x using the matrix Δ , that is, $\eta \mapsto \eta(\cdot) - \Delta(x, \cdot)$. Let us call a finite or infinite sequence consisting of topplings of unstable vertices **exhaustive**, if any vertex that is

unstable at some point, is toppled at a later time. It can be shown, similarly to the finite graph case, that for all $x \in \mathbb{Z}^d$, x topples the same number of times (possibly infinity) in any exhaustive sequence.

We write e_i for the unit vector in the i -th positive coordinate direction, $|\cdot|$ for the Euclidean norm, and $\|\cdot\|$ for the ℓ^∞ norm on \mathbb{Z}^d . We denote by o the origin in \mathbb{Z}^d , and we let

$$V_x(n) = \{y \in \mathbb{Z}^d : \|y - x\| \leq n\}; \quad B_x(n) = \{y \in \mathbb{Z}^d : |y - x| \leq n\},$$

and write $V(n) = V_o(n)$ and $B(n) = B_o(n)$.

Given $A, B \subset \mathbb{Z}^d$, we let $\text{dist}(A, B)$ denote their Euclidean distance, and write $\text{dist}(x, B)$ when $A = \{x\}$. If $A \subset \mathbb{Z}^d$, we let $\partial A = \{x \in \mathbb{Z}^d \setminus A : \text{dist}(x, A) = 1\}$. When considering A as a subset of G_L , we will often use ∂A to denote the boundary restricted to G_L — that is, excluding from ∂A any $x \notin G_L$ — the meaning will be clear in context.

If z_1, z_2 are two elements of \mathbb{R}^d , let $\text{ang}(z_1, z_2)$ denote the angle between z_1 and z_2 . We will make use of the “little o ” notation: $a_n = n^{a+o(1)}$ if $\lim_n [\log a_n / \log n] = a$.

2.2 Toppling numbers and waves

Given a sandpile η on a finite connected graph $H = (U \cup \{s\}, F)$, we write $n(x, y) = n(x, y; \eta)$ for the **toppling numbers**, i.e., number of times y topples when stabilizing $\eta + \mathbf{1}_x$. A useful ordering of topplings, introduced by Ivashkevich, Ktitarev and Priezzhev [13], for stabilizing $\eta + \mathbf{1}_x$ is in terms of **waves**, which we now define. If $\eta + \mathbf{1}_x$ is stable, there are no waves. Otherwise, topple x and carry out any further topplings that are possible without toppling x a second time. It is easy to verify that in doing so, every vertex in U topples at most once. We write $\mathcal{W}_{1,H} = \mathcal{W}_{1,H}(x; \eta)$ for the set of sites toppled so far; this is the **first wave**. If x is still unstable after the first wave, which happens if and only if all neighbours of o are in $\mathcal{W}_{1,H}$, topple x a second time, and carry out any further topplings that are possible without toppling x a third time. The set of vertices that topple in doing so is denoted $\mathcal{W}_{2,H} = \mathcal{W}_{2,H}(x; \eta)$; this is the **second wave**. We define further waves analogously.

When the graph arises from a finite $V \subset \mathbb{Z}^d$, we write $\mathcal{W}_{1,V}, \mathcal{W}_{2,V}, \dots$ for the waves. We make similar definitions for sandpiles on \mathbb{Z}^d . On \mathbb{Z}^d , the toppling numbers $n(x, y; \eta)$ are possibly infinite. Waves $\mathcal{W}_1, \mathcal{W}_2, \dots$ are defined analogously to the finite case, and their number may be infinite. The following lemma is straightforward to verify.

Lemma 2.1.

- (i) Let $H = (U \cup \{s\}, F)$ be a connected finite graph. The number of waves containing y equals $n(x, y; \eta)$. In particular, the number of waves is $N = N(x; \eta) := n(x, x; \eta)$.
- (ii) In the case of \mathbb{Z}^d , the number of waves containing y equals $n(x, y; \eta)$ (possibly infinite). In particular, the number of waves is $N = N(x; \eta) := n(x, x; \eta)$.

The next lemma gives the expected number of topplings in an avalanche.

Lemma 2.2 (Dhar’s formula [8]). Let $H = (U \cup \{s\}, F)$ be finite. We have

$$\mathbb{E}_{\nu_H} n(x, y) = g_H(x, y), \quad x, y \in U.$$

For cubes in \mathbb{Z}^d , we have the following deterministic comparison between $n(o, o; \eta)$ and the radius $R(\eta)$.

Lemma 2.3.

- (i) Let η be any sandpile configuration on \mathbb{Z}^d , $d \geq 1$. Then

$$n(o, o; \eta) \leq R(\eta).$$

- (ii) The same statement holds when η is a sandpile configuration in G_L for any $L \geq 1$.

Proof. (i) In the proof below, it will be convenient to denote $n[\eta] := n(o, o; \eta)$. We will also use the notion of “restricted topplings”. For $L > 0$, let $\eta|_L$ denote the restriction of η to $V(L)$; the restricted toppling number $n_L[\eta]$ will denote the number of topplings occurring at o in the stabilization of $\eta|_L + \mathbf{1}_o$ on G_L . Note that such a stabilization amounts to taking the configuration $\eta + \mathbf{1}_o$ on \mathbb{Z}^d and toppling only sites $x \in V(L)$ until all such x are stabilized, neglecting any sites in $\mathbb{Z}^d \setminus V(L)$ that may become unstable. In particular, for any stable sandpile configuration η on \mathbb{Z}^d , $n[\eta] \geq n_L[\eta]$ for any L .

Assume η and $R < \infty$ are as in the statement of the lemma. We claim that the equality $n[\eta] = n_R[\eta]$ holds. Indeed, by the above observation, n_R is exactly the number of topplings at o required to stabilize only the sites of $V(R)$, but by assumption these are the only sites which need to be toppled to stabilize $\eta + \mathbf{1}_o$ in all of \mathbb{Z}^d . Therefore, it suffices to show that $n_R[\eta] \leq R$.

For this, we will use a special “maximal” configuration ϕ_R from [11, Lemma 4.2]:

$$\phi_R(x) = \begin{cases} 2d - 1, & x \in V(R) \\ 2d - 2 & \text{otherwise.} \end{cases}$$

Note that $n_R[\eta] \leq n_R[\phi_R]$, since $\eta|_R \leq \phi_R$ pointwise. Moreover, $n_R[\phi_R] \leq n[\phi_R]$. It is proven in [11, Lemma 4.2], and not difficult to see by computing each wave, that $n[\phi_R] = R$. This completes the proof of (i).

(ii) The above proof applies here as well, with only minor changes. □

Proof of Proposition 1.4. By Dhar’s formula [8], we have

$$\lim_{L \rightarrow \infty} \mathbb{E}_{\nu_L} n(o, o) = \lim_{L \rightarrow \infty} g_{V_L}(o, o) = \infty, \quad \text{when } d = 2.$$

By Lemma 2.3(ii), this implies $\lim_{L \rightarrow \infty} \mathbb{E}_{\nu_L} R = \infty$. □

Remark 2.4. We do not see a simple way to deduce the statement $\mathbb{E}_{\nu} n(o, o) = \infty$ from the simpler statement that $\lim_{L \rightarrow \infty} \mathbb{E}_{\nu_L} n(o, o) = \infty$, when $d = 2$. This requires ruling out the possibility that $\mathbb{E}_{\nu_L} n(o, o)$ is dominated by rare events with many waves. Our proof of $\mathbb{E}_{\nu} n(o, o) = \infty$ in Theorem 1.9 will build on quite a few other results; in particular, results from Section 5.

2.3 Spanning trees and the burning bijection

In 1990, Dhar [8] introduced a method for checking whether a particular stable configuration η lies in \mathcal{R}_H , called the “burning algorithm”. Application of the burning algorithm provides a bijection φ between \mathcal{R}_H and the set of all spanning trees of H , that we denote by \mathcal{T}_H ; see [28]. We briefly describe this bijection here. In Section 2.5, we give a version of it for “waves” which will be necessary in our analysis.

Recall that a_{xy} denotes the number of edges between vertices x and y .

Lemma 2.5 (Burning algorithm [8], [12, Lemma 4.1]). *Let η be a stable sandpile on a connected finite graph $H = (U \cup \{s\}, F)$. At each $x \in U$, add a_{xs} grains of sand, and stabilize. We have $\eta \in \mathcal{R}_H$ if and only if each vertex in U topples exactly once.*

Note that instead of adding sand as in the lemma, we may initiate the toppling process by placing all $\sum_{x \in U} a_{xs} = \deg_H(s)$ grains at s , and toppling s first. Suppose we carry out any possible topplings in parallel. We say that x **burns at time** k , if it is toppled in the k -th parallel toppling step, where we regard s to have burnt at time 0.

The bijection is defined as follows. For each $y \in U$, fix an arbitrary ordering \prec_y of the edges adjacent to y . Given $\eta \in \mathcal{R}_H$, for each $y \in U$, we adjoin to the tree T an edge connecting y to a neighbour burnt one time step before, chosen as follows. If P_y is the

number of edges joining y to neighbours burnt before y , and A_y is the subset of such edges leading to sites burnt one step before y , then the burning rule implies

$$\eta(y) = \deg_H(y) - P_y + i \quad \text{for some } 0 \leq i < |A_y| .$$

We add to T the i -th edge in A_y in the ordering \prec_y .

The resulting graph T will be a spanning tree (the fact that it spans — i.e., that every site topples in this procedure — is part of the content of Lemma 2.5), and we set $\varphi(\eta) = T$. The map $\varphi : \mathcal{R}_H \rightarrow \mathcal{T}_H$ is usually referred to as the “burning bijection”, and the toppling procedure used to construct φ will be referred to as the “burning procedure”.

2.4 Intermediate configurations

Now, we give a description of waves in terms of recurrent configurations on an auxiliary graph, as introduced in [13]. As in the previous section, we describe this in an arbitrary finite connected graph $H = (U \cup \{s\}, F)$. Suppose we are interested in waves started by the addition of a particle at a fixed vertex $w \in U$. Consider the graph H' obtained from H by adding the edge $f' := \{w, s\}$. For readability, we denote $\Delta'_H := \Delta_{H'}$ and $\mathcal{R}'_H := \mathcal{R}_{H'}$. Let $a'_x = a'_{x,H}$, $x \in U$ denote the addition operators on H' . We reserve the notation η° for stabilization on the original graph H . When we need to emphasize that addition is applied on the graph derived from H , we prefer the $a'_{x,H}$ notation.

The burning algorithm (Lemma 2.5) implies that $\mathcal{R}_H \subset \mathcal{R}'_H$. The following lemma compares the sizes of these two sets.

Lemma 2.6. *For any finite connected graph $H = (U \cup \{s\}, F)$, and $w \in U$, we have $|\mathcal{R}'_H| = (1 + g_H(w, w))|\mathcal{R}_H|$.*

Proof. Let $\mathbf{1}_{w,w}$ denote the $U \times U$ matrix whose only non-zero entry is a 1 at (w, w) . We have

$$\begin{aligned} |\mathcal{R}'_H| &= \det(\Delta'_H) = \det(\Delta_H + \mathbf{1}_{w,w}) = (1 + (\Delta_H)^{-1}(w, w)) \det(\Delta_H) \\ &= (1 + g_H(w, w)) |\mathcal{R}_H|. \end{aligned} \tag{2.3}$$

□

The following immediate corollary will be useful in controlling avalanches. A version of part (i) was proved in [15, Lemma 7.5].

Corollary 2.7. *Consider the sequence $(V_L)_L$.*

- (i) *Suppose $d \geq 3$. There exists a constant $C(d)$ such that $|\mathcal{R}'_L| \leq C(d) |\mathcal{R}_L|$ for all $L \geq 1$.*
- (ii) *Suppose $d = 2$. There exists a constant C such that $|\mathcal{R}'_L| \leq C \log L |\mathcal{R}_L|$ for all $L \geq 2$.*

Proof. Both statements follow from Lemma 2.6, the equality $g_L(o, o) = (2d)^{-1}G_L(o, o)$, and known properties of the Green function $G_L(o, o)$; see e.g. [24]. □

We now describe the interpretation of waves as elements of $\mathcal{R}'_H \setminus \mathcal{R}_H$; introduced in [13]. Let $\eta \in \mathcal{R}_H$, and suppose that $\eta + \mathbf{1}_w$ is unstable at w in H , i.e. $\eta(w) = \deg_H(w) - 1$. Consider the waves occurring in stabilizing $\eta + \mathbf{1}_w$ in H . Recall that $N = N(\eta)$ denotes the number of waves. For $1 \leq k \leq N$, let $\eta_k = a'_w \eta_{k-1}$, where $\eta_0 = \eta$. It is straightforward to check that η_k is the configuration seen just before the k -th wave is carried out, and $a_w \eta = (a'_w)^{N+1} \eta$. Note that the latter statement also holds, trivially, when $\eta + \mathbf{1}_w$ is stable, in which case $N = 0$.

Definition 2.8. *Let $\eta \in \mathcal{R}_H$ be such that $\eta + \mathbf{1}_w$ is unstable at w . We call the sequence $\alpha(\eta) := (\eta_1, \dots, \eta_N)$ the **intermediate configurations** corresponding to η .*

We record here the characterization of $\mathcal{R}'_H \setminus \mathcal{R}_H$; a similar statement was shown in [16] for a continuous height model.

Lemma 2.9. *The collection $\{\alpha(\eta) : \eta \in \mathcal{R}_H, \eta(w) = \deg_H(w) - 1\}$ forms a partition of $\mathcal{R}'_H \setminus \mathcal{R}_H$.*

Proof. Since $\eta_k(w) = \deg_H(w)$, $k = 1, \dots, N$, we have $\eta_k \in \mathcal{R}'_H \setminus \mathcal{R}_H$, $k = 1, \dots, N$. This and the relation $a_w \eta = (a'_w)^{N+1} \eta$ imply that $\alpha(\eta)$ has distinct entries (the order of a'_w is at least $N + 1$). By Dhar's formula, the average number of waves per recurrent configuration is $g_H(w, w)$, so $g_H(w, w)|\mathcal{R}_H|$ elements of \mathcal{R}'_H correspond to intermediate configurations.

It is similarly easy to check that $\eta \in \mathcal{R}'_H$, accounting for another $|\mathcal{R}_H|$ elements of \mathcal{R}'_H . Comparing this with Lemma 2.6, we see that every element of \mathcal{R}'_H is either an intermediate configuration or recurrent on H , completing the proof. \square

Given $\eta_* \in \mathcal{R}'_H \setminus \mathcal{R}_H$, we denote by $\mathcal{W}(\eta_*)$ the set of vertices that topple in the stabilization $a'_{w,H}(\eta_*)$. It is immediate from this definition that if $\alpha(\eta) = (\eta_1, \dots, \eta_N)$, then $\mathcal{W}(\eta_k)$ is the k -th wave corresponding to η , i.e. $\mathcal{W}_{k,H}(w; \eta)$. Of particular interest will be the last wave $\mathcal{W}(\eta_N)$. The following corollary follows directly from the definitions.

Corollary 2.10. *An intermediate configuration $\eta_* \in \mathcal{R}'_H \setminus \mathcal{R}_H$ is a last wave if and only if there exists $y \sim w$, $y \in U \cup \{s\}$, such that $y \notin \mathcal{W}(\eta_*)$.*

We will also need the following lemma.

Lemma 2.11. *We have*

$$\frac{1}{\deg_H(w)} |\mathcal{R}_H| \leq |\{\eta_* \in \mathcal{R}'_H : \eta_* \text{ is a last wave}\}| \leq |\mathcal{R}_H|.$$

Proof. The upper bound is obvious. To see the lower bound, we assign to $\eta \in \mathcal{R}_H$ the last intermediate configuration in the stabilization of $\eta + (\deg_H(w) - \eta(w))\mathbf{1}_w$. This map is at most $\deg_H(w)$ to 1, proving the lower bound. \square

2.5 Bijection for intermediate configurations

We now specialize to the set-up where $H = G_V$, $V \subset \mathbb{Z}^d$ finite, $o \in V$. In this section, we describe a version of the burning bijection on G'_V , that will allow us to control topplings occurring in a wave. To the best of our knowledge, such a bijection was first introduced by Ivashkevich, Ktitarov and Priezzhev [13]. See also [32, 15, 16], where it played a key role. For many of our results, the variant in [13] would suffice. However, a more careful choice of the burning process will be needed in Section 5, so we introduce here the version we need. Our burning process is similar to burning processes introduced in [17] and [10].

Let $\eta_* \in \mathcal{R}'_V \setminus \mathcal{R}_V$. We define a pair of vertex-disjoint trees $(T_o, T_s) = \varphi'(\eta_*)$, such that $T_o \cup T_s$ spans G_V . Send one grain of sand from s to o , resulting in $2d$ grains at o . We sequentially topple vertices in the balls $B(0) \cap V, B(1) \cap V, B(2) \cap V, \dots$, and build a tree rooted at o , similarly to the usual burning bijection. The precise definitions of burnt and unburnt sets are as follows. We let

$$\begin{aligned} \text{Bt}_0^{(0)} &= \{o\} & \text{Ut}_0^{(0)} &= V \cup \{s\} \setminus \{o\} \\ \text{Bt}_k^{(0)} &= \emptyset, \quad k \geq 1, & \text{Ut}_k^{(0)} &= V \cup \{s\} \setminus \{o\}, \quad k \geq 1. \end{aligned}$$

For $r \geq 1$, inductively, we set

$$\begin{aligned} \text{Bt}_0^{(r)} &= \cup_{\ell \geq 0} \text{Bt}_\ell^{(r-1)} & \text{Ut}_0^{(r)} &= V \cup \{s\} \setminus \text{Bt}_0^{(r)} \\ \text{Bt}_k^{(r)} &= \left\{ x \in B(r) \cap \text{Ut}_{k-1}^{(r)} : \eta_*(x) \geq \deg_{\text{Ut}_{k-1}^{(r)}}(x) \right\} & \text{Ut}_k^{(r)} &= \text{Ut}_{k-1}^{(r)} \setminus \text{Bt}_k^{(r)}, \quad k \geq 1. \end{aligned}$$

For each $r \geq 1$ there exists a smallest index $J = J(r) \geq 1$ such that $\text{Bt}_J^{(r)} = \emptyset$, and there is a smallest index $R \geq 1$ such that $J(R + 1) = 1$. Then $\text{Bt}_0^{(R+1)} = \mathcal{W}(\eta_*)$ is the set of vertices toppled in the wave represented by η_* .

We complete the burning process by sending a_{sx} grains of sand from s to x for each $x \in V$, and follow the usual burning rule. That is, we set:

$$\begin{aligned} \widetilde{\text{Bt}}_0 &= \cup_{r \geq 0} \cup_{\ell \geq 0} \text{Bt}_\ell^{(r)} & \widetilde{\text{Ut}}_0 &= V \setminus \widetilde{\text{Bt}}_0 \\ \widetilde{\text{Bt}}_k &= \left\{ x \in \widetilde{\text{Ut}}_{k-1} : \eta_*(x) \geq \deg_{\widetilde{\text{Ut}}_{k-1}}(x) \right\} & \widetilde{\text{Ut}}_k &= \widetilde{\text{Ut}}_{k-1} \setminus \widetilde{\text{Bt}}_k, \quad k \geq 1. \end{aligned}$$

We now define the bijection. If $o \neq u \in V \cap \text{Bt}_0^{(R+1)}$, then there exists a unique pair (r, k) with $r \geq 1$ and $k \geq 1$ such that $u \in \text{Bt}_k^{(r)}$. Due to the definition of the burning rule, there exists at least one $y \sim x$ such that $y \in \text{Bt}_{k-1}^{(r)}$. We select an edge that connects u to one of these vertices, using the ordering \prec_u , as in Section 2.3. Namely, if P_u is the number of edges joining u to neighbours in $\cup_{\ell < k} \text{Bt}_\ell^{(r)}$, and A_u is the subset of such edges leading to vertices in $\text{Bt}_{k-1}^{(r)}$, then necessarily

$$\eta_*(u) = 2d - P_u + i \quad \text{for some } 0 \leq i < |A_u|.$$

We add to T_o the i -th edge in A_u in the ordering \prec_u .

If $u \in \text{Ut}_0^{(R+1)} = \widetilde{\text{Ut}}_0$, there is a unique $k \geq 1$ such that $u \in \widetilde{\text{Bt}}_k$, and there exists at least one $y \sim x$ with $y \in \widetilde{\text{Bt}}_{k-1}$. We select an edge to one of these vertices, using the ordering \prec_u as before. Namely, if P_u is the number of edges joining u to neighbours in $\cup_{\ell < k} \widetilde{\text{Bt}}_\ell$, and A_u is the subset of such edges leading to vertices in $\widetilde{\text{Bt}}_{k-1}$, then necessarily

$$\eta_*(u) = 2d - P_u + i \quad \text{for some } 0 \leq i < |A_u|.$$

We add to T_s the i -th edge in A_u in the ordering \prec_u . Let $\varphi'(\eta_*) := (T_o, T_s)$ denote the two components spanning forest obtained by the above construction. Let us write $\mathcal{T}_{V,o}$ for the set of all spanning forests of G_V rooted at $\{s, o\}$.

Lemma 2.12.

- (i) The map φ' is a bijection between $\mathcal{R}'_V \setminus \mathcal{R}_V$ and $\mathcal{T}_{V,o}$.
- (ii) For any $\eta_* \in \mathcal{R}'_V \setminus \mathcal{R}_V$, the vertex set of $T_o(\eta_*)$ equals $\mathcal{W}(\eta_*)$.
- (iii) We have the following property:

$$\begin{aligned} & \text{If there is a path from } o \text{ to a vertex } x \in V \text{ in } T_o = \varphi'(\eta_*) \text{ that} \\ & \text{stays inside } B(r), \text{ then starting from } \eta_* + \mathbf{1}_o \text{ there is a sequence} \\ & \text{of topplings in } B(r) \text{ that topples } x. \end{aligned} \tag{2.4}$$

Proof. (i) Let $\eta_* \neq \widehat{\eta}_* \in \mathcal{R}'_V \setminus \mathcal{R}_V$. Tracing the burning process to the first time when a vertex with $\eta_*(x) \neq \widehat{\eta}_*(x)$ is encountered, we see that φ' is injective on $\mathcal{R}'_V \setminus \mathcal{R}_V$. It follows from the definitions that $\varphi'(\mathcal{R}'_V \setminus \mathcal{R}_V)$ is a subset of the set of spanning forests of G_V rooted at $\{o, s\}$. By the matrix-tree theorem applied to G'_V , the number of spanning forests of G_V rooted at $\{o, s\}$ equals $\det(\Delta'_V) - \det(\Delta_V)$. This also equals $|\mathcal{R}'_V \setminus \mathcal{R}_V|$ [8], so statement (i) follows.

(ii) The burning process that was used to define $\text{Bt}_0^{(R+1)}$ can be identified with topplings in the wave corresponding to $a'_{o,V}(\eta_*)$. This implies that $\text{Bt}_0^{(R+1)} = \mathcal{W}(\eta_*)$, and this is the vertex set of T_o .

(iii) This again follows directly from the interpretation of the burning process in terms of topplings in the wave. □

2.6 Random walk notation and basic facts

Many of our techniques require a detailed analysis of spanning trees via Wilson’s algorithm. For this reason, we will often have to consider collections of simple random walks (“SRW”) and loop-erased random walks (“LERW”) on (subsets of) \mathbb{Z}^d .

We will denote by $S_x = (S_x(0), S_x(1), \dots)$ an infinite simple random walk on \mathbb{Z}^d started at x , so that $S_x(0) = x$. We will suppress the subscript when the choice of x is clear or when $x = o$. In general, S_x and S_y will be assumed independent when $x \neq y$. When multiple independent walks from one site are necessary, we write S_x, S'_x (etc).

If $A \subset \mathbb{Z}^d$ we define the standard stopping times

$$\sigma_A = \inf\{n \geq 0 : S_x(n) \notin A\} \qquad \xi_A = \inf\{n \geq 0 : S_x(n) \in A\} , \qquad (2.5)$$

$$\bar{\sigma}_A = \inf\{n \geq 1 : S_x(n) \notin A\} \qquad \bar{\xi}_A = \inf\{n \geq 1 : S_x(n) \in A\} . \qquad (2.6)$$

Note that we suppress any dependence of these stopping times on the starting point x ; when we write (for instance) $S_x(\xi_A)$, we are referring to the location of S_x at its first hitting time on A . When the starting vertex may be ambiguous, we use subscripts on the symbol \mathbb{P} ; for instance,

$$\mathbb{P}_x(\xi_A < \xi_B) = \mathbb{P}(S_x[0, \xi_A] \cap B = \emptyset) .$$

We will abbreviate $\sigma_{V(n)}$ to σ_n , and write ξ_o for $\xi_{\{o\}}$.

The trace of a walk S_x between two times $a < b$ (where b can be infinite) will be denoted

$$S_x[a, b] := \{S_x(j) : a \leq j \leq b\},$$

and similar notation will be used for other intervals—e.g., $S(a, b)$ and so on. We will sometimes abuse notation and treat $S_x[a, b]$ as a sequence instead of an unordered set.

If $A \subset \mathbb{Z}^d$ is a finite connected set and if $x, y \in \mathbb{Z}^d$, recall the Green function

$$G_A(x, y) := \sum_{0 \leq j < \sigma_A} \mathbb{P}(S_x(j) = y) . \qquad (2.7)$$

As before, $G_{V(n)}$ is abbreviated G_n . We will use the following standard asymptotics for the Green function inside a large ball and the probability of hitting o before exiting a large ball:

Theorem 2.13 (See [23], Prop. 1.5.9 and 1.6.7).

(i) If $d = 2$, we have uniformly in $x \in B(n)$:

$$G_{B(n)}(o, x) = \frac{2}{\pi} [\log n - \log |x|] + O(|x|^{-1} + 1/n)$$

$$\mathbb{P}_x(\xi_o < \sigma_{B(n)}) = \frac{1}{\log n} [\log n - \log |x| + O(|x|^{-1} + 1/n)] .$$

(ii) For $d \geq 3$, there exist $c_1 = c_1(d), c_2 = c_2(d) > 0$ such that, uniformly in n and $x \in B(n)$:

$$G_{B(n)}(o, x) = c_1 [|x|^{2-d} - n^{2-d}] + O(|x|^{1-d})$$

$$\mathbb{P}_x(\xi_o < \sigma_{B(n)}) = c_2 [|x|^{2-d} - n^{2-d}] + O(|x|^{1-d}) .$$

We also note the following simple observation about G . If $K_1 \subset K_2$ and $x, y \in K_1$,

$$G_{K_1}(x, y) \leq G_{K_2}(x, y) . \qquad (2.8)$$

We will need the following result, usually called the Beurling estimate, which gives an upper bound on the probability that a path in \mathbb{Z}^2 is not hit by SRW.

Lemma 2.14 (Beurling estimate [19], [24, Section 6.8]). *Consider \mathbb{Z}^2 . There is a constant C such that the following bound holds, uniformly in x , n , and lattice paths α connecting o to $\partial B(n)$:*

$$\mathbb{P}_x(\sigma_{B(n)} < \xi_\alpha) \leq C \left(\frac{|x|}{n} \right)^{1/2}.$$

Given a transient random walk S_x , we denote by $\mathcal{L}S_x$ the loop-erasure of S_x , where loops are erased in forward chronological order. A similar definition is made for $\mathcal{L}S_x[a, b] := \mathcal{L}(S_x[a, b])$, etc.; note that for a finite segment of a random walk, the loop erasure can be defined even in the case that the walk is recurrent. See [24, Chapter 9] and [23, Chapter 7] for background on LERW; in particular, for properties not detailed below.

If $A \ni x$ is a finite subset of \mathbb{Z}^d , let \widehat{S}_x^A denote a finite loop-erased random walk killed at the boundary of A :

$$\widehat{S}_x^A = \mathcal{L}S_x[0, \sigma_A].$$

We will also make use of infinite loop-erased random walks \widehat{S}_x on \mathbb{Z}^d . When $d \geq 3$, the definition

$$\widehat{S}_x := \mathcal{L}S_x$$

is unambiguous with probability 1, as noted in the last paragraph.

For $d = 2$, the loop-erasure of the infinite random walk S_x is not well-defined; in this case, \widehat{S}_x is defined by taking limits. It is known (see [23, Section 7.4]) that for any finite lattice path γ with $|\gamma| = k$,

$$\lim_{n \rightarrow \infty} \mathbb{P}(\widehat{S}_x^{B(n)}[0, k] = \gamma) =: \mathbb{P}(\widehat{S}_x[0, k] = \gamma)$$

exists, and the extension of this to a measure on infinite paths gives a definition of the distribution of \widehat{S}_x . The rate of convergence of $\widehat{S}_x^{B(n)}$ to \widehat{S}_x is well-controlled; see Lemma 3.4 below. We will refer to all of the processes $\widehat{S}_x, \widehat{S}_x^A$ as loop-erased random walks or LERW. We will also assume as usual (unless stated otherwise) that a LERW is independent of any other walks appearing in a given statement.

We define LERW stopping times $\widehat{\xi}, \widehat{\sigma}$ analogously to ξ and σ ; for instance,

$$\widehat{\sigma}_K = \inf\{n \geq 0 : \widehat{S}_x(n) \notin K\}.$$

As before, we will use x as a subscript on \mathbb{P} when considering stopping times to indicate the starting point. When the LERW is finite (i.e., we are considering \widehat{S}_x^A), we also will use the superscript A on \mathbb{P} , writing (for instance)

$$\mathbb{P}_x^A(\widehat{\xi}_{K_1} < \widehat{\xi}_{K_2})$$

to avoid confusing the set in which the LERW lives with the set it is hitting. When the LERW is infinite, we omit superscripts altogether.

One result important for analyzing LERW is the following ‘‘Domain Markov Property’’ (DMP). This roughly says that the terminal segment of a LERW can be built by starting a SRW at the tip of the initial segment, conditioning it not to hit the initial segment, then erasing loops.

Lemma 2.15 (Domain Markov Property; see [24, Chapter 11]). *Let \widehat{S}_x^K be a loop-erased walk in K , and let α be a finite path of length m such that*

$$\mathbb{P}(\widehat{S}_x^K[0, m] = \alpha) > 0.$$

Then for all paths β ,

$$\mathbb{P}(\widehat{S}_x^K[m, \widehat{\sigma}_K] = \beta \mid \widehat{S}_x^K[0, m] = \alpha) = \mathbb{P}(\mathcal{L}S_{\alpha(m)}[0, \sigma_K] = \beta \mid \sigma_K < \bar{\xi}_\alpha).$$

Note that Lemma 2.15 is in fact much more general: it holds for infinite LERW (on \mathbb{Z}^d for $d \geq 3$) and finite LERW on graphs which are not necessarily subsets of \mathbb{Z}^d .

2.7 Wilson’s algorithm

Let $H = (U \cup \{s\}, F)$ be a connected finite graph. Since ν_H is uniform on \mathcal{R}_H , the bijection φ maps it to the uniform measure on spanning trees of H , which we denote by μ_H . In this section, we collect basic facts about Wilson’s algorithm, that we will use to analyze μ_H . For an in-depth introduction to uniform spanning trees (UST) see the book [26].

We denote a sample from μ_H by \mathfrak{T}^H . We will usually identify a spanning tree of H with the set of edges it contains, so $\mathfrak{T}^H \subset F$. Pemantle [31] proved that for $x, y \in U$, the path in \mathfrak{T}^H between x and y is distributed as a LERW from x to y (i.e. as $\mathcal{L}S_x[0, \xi_y]$). Wilson’s algorithm [35] provides a method for constructing the full UST (a sample from μ_H) from LERWs.

Wilson’s algorithm. Let v_1, \dots, v_n be an enumeration of the vertices in U . We construct a random sequence of tree subgraphs $\mathfrak{F}_0 \subset \dots \subset \mathfrak{F}_n$. Let \mathfrak{F}_0 have vertex set $\{s\}$ and empty edge set. If \mathfrak{F}_{i-1} has been defined for some $1 \leq i \leq n$, we let S_{v_i} be a simple random walk on H started at v_i , and let $\xi_{V(\mathfrak{F}_{i-1})}$ be the first hitting time of the vertex set of \mathfrak{F}_{i-1} by S_{v_i} . We set $\mathfrak{F}_i = \mathfrak{F}_{i-1} \cup \mathcal{L}S_{v_i}[0, \xi_{V(\mathfrak{F}_{i-1})}]$ —that is, the edges in the loop-erasure of $S_{v_i}[0, \xi_{V(\mathfrak{F}_{i-1})}]$ are added to \mathfrak{F}_{i-1} . The output of the algorithm is \mathfrak{F}_n . Wilson’s theorem [35] implies that \mathfrak{F}_n is uniform, i.e. distributed as \mathfrak{T}^H .

The measure $\mu_L := \mu_{G_L}$ is known as the UST in $V(L)$ with the **wired boundary condition**. For studying the $L \rightarrow \infty$ limit of sandpiles on G_L , as well as sandpiles on \mathbb{Z}^d , it will be useful to consider the weak limit $\lim_{L \rightarrow \infty} \mu_L =: \mathbf{WSF}$, called the **wired spanning forest measure**. Existence of the limit is implicit in [31]; see [26] for an in-depth treatment. We denote a sample from **WSF** by \mathfrak{T} .

It is well known that **WSF** concentrates on spanning forests of \mathbb{Z}^d all whose components are infinite. Pemantle [31] showed that for $d \leq 4$, \mathfrak{T} is a tree **WSF**-a.s, while for $d \geq 5$, \mathfrak{T} has infinitely many connected components **WSF**-a.s. This dichotomy is the underlying fact behind mean-field behaviour of the sandpile model for $d \geq 5$; which is reflected in some of our results and proofs. We write \mathfrak{T}_x for the component of \mathfrak{T} containing $x \in \mathbb{Z}^d$.

It is possible to construct \mathfrak{T} more directly, using an appropriate extensions of Wilson’s algorithm. Let v_1, v_2, \dots be an enumeration of \mathbb{Z}^d . When $d \geq 3$, we set $\mathfrak{F}_1 = \mathcal{L}S_{v_1}[0, \infty)$. Then for $i \geq 2$, we inductively define $\mathfrak{F}_i = \mathfrak{F}_{i-1} \cup \mathcal{L}S_{v_i}[0, \xi_{V(\mathfrak{F}_{i-1})}]$, where the stopping time may be finite or infinite. See [6, 26] for a proof that $\cup_{i \geq 1} \mathfrak{F}_i$ has the distribution of \mathfrak{T} given by **WSF**. This is called Wilson’s method rooted at infinity.

When $d = 2$, a method analogous to that in finite volume can be used. We set $\mathfrak{F}_1 = \{v_1\}$, and for $i \geq 2$ we inductively define $\mathfrak{F}_i = \mathfrak{F}_{i-1} \cup \mathcal{L}S_{v_i}[0, \xi_{V(\mathfrak{F}_{i-1})}]$; see [6, 26].

It will be important for our proofs that **WSF**-a.s. each component of \mathfrak{T} has **one end**, for all $d \geq 2$. This means that any two infinite self-avoiding paths lying in the same component of \mathfrak{T} have a finite symmetric difference. For $2 \leq d \leq 4$ this was proved by Pemantle [31]. For $d \geq 5$ this was first shown by Benjamini, Lyons, Peres and Schramm [6] (who generalized it to a much larger class of infinite graphs).

2.8 Wiring o to the boundary

Let $\mu_{L,o}$ be the uniform measure on $\mathcal{T}_{L,o}$. Under $\mu_{L,o}$, we denote the components containing o and s , respectively, by $\mathfrak{T}_{L,o}$ and $\mathfrak{T}_{L,s}$, respectively.

Due to monotonicity, the weak limit $\lim_{L \rightarrow \infty} \mu_{L,o} =: \mathbf{WSF}_o$ exists; see [6, 26]. We also use \mathfrak{T} to denote a sample from **WSF** $_o$, and write \mathfrak{T}_x for its component containing

$x \in \mathbb{Z}^d$. In particular, \mathfrak{T}_o is the component containing o under the measure \mathbf{WSF}_o . It was shown by Lyons, Morris and Schramm [25] that for $d \geq 3$ (and more generally under a suitable isoperimetric condition), we have $\mathbf{WSF}_o(|\mathfrak{T}_o| < \infty) = 1$, and also that this is equivalent to the one-end property of \mathbf{WSF} . Járai and Redig [15] used the finiteness of \mathfrak{T}_o to show that $\nu(S < \infty) = 1$ when $d \geq 3$.

3 Toppling probability bounds in low dimensions

In this section, we give a proof of our toppling probability lower bounds stated in Theorem 1.1. These will follow from the following theorem, which gives a lower bound in terms of a non-intersection probability between a loop-erased walk and a simple random walk, and a random walk hitting probability. Our arguments also apply to $d \geq 5$ with an identical statement, however, this case is already known from [16, Section 6.2].

Recall that S_x and S'_x are independent simple random walks from the site x .

Theorem 3.1. *Assume $2 \leq d \leq 4$. There is a constant $c = c(d) > 0$ such that, for all $z \in \mathbb{Z}^d$ we have*

$$\nu(z \in \text{Av}) \geq c \mathbb{P} \left(S'_o[0, \sigma_{|z|}] \cap \widehat{S}_o(0, \widehat{\sigma}_{|z|}] = \emptyset \right) \mathbb{P}(z \in S_o[0, \infty)) . \tag{3.1}$$

Moreover, the right hand side of (3.1) is a lower bound on $\nu_L(z \in \text{Av})$ for all $L \geq 4\|z\|$.

In Sections 3.1–3.2, we state and prove preliminary results which are useful for establishing Theorem 3.1, and in Section 3.3 we use these to prove the theorem. In Section 3.3, we also give a corollary which will be useful for proofs of later theorems.

3.1 Preliminary setup

Our strategy for proving Theorem 3.1 will be to work in large finite volume $V(L)$. That is, given a particular $z \in \mathbb{Z}^d$, we will choose some L_0 sufficiently large, so that the probability $\nu_L(z \in \text{Av})$ is close to the claimed value for all $L \geq L_0$. We will require L_0 to be on the order of some large multiple of $\|z\|$.

The main idea of the proof is to show a lower bound for the probability that z is in the last wave of the avalanche. By Corollary 2.10,

$$\begin{aligned} |\{\eta \in \mathcal{R}_L : z \in \text{Av}\}| &\geq |\{\eta \in \mathcal{R}_L : \eta(o) = 2d - 1, z \in \mathcal{W}_{N(\eta)}\}| \\ &= |\{\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L : z \in \mathcal{W}(\eta_*), v \notin \mathcal{W}(\eta_*) \text{ for some } v \sim o\}| . \end{aligned}$$

Dividing by $|\mathcal{R}_L|$, using Lemma 2.11 and symmetry of $V(L)$, for any fixed $e \sim o$ we get

$$\begin{aligned} \nu_L(z \in \text{Av}) &\geq \mu_{L,o}(z \in \mathfrak{T}_{L,o} \mid v \notin \mathfrak{T}_{L,o} \text{ for some } v \sim o) \\ &\geq (2d)^{-1} \mu_{L,o}(z \in \mathfrak{T}_{L,o} \mid e \notin \mathfrak{T}_{L,o}) \end{aligned} \tag{3.2}$$

uniformly in z, L .

We analyze the event

$$A(z, e) = \{z \in \mathfrak{T}_{L,o}, e \notin \mathfrak{T}_{L,o}\}.$$

Let us apply Wilson’s algorithm in the graph $G_{L,o}$, starting with a walk S_e from e , followed by a walk S_z from z . This gives that for fixed $e \sim o$, the occurrence of $A(z, e)$ is equivalent to a LERW from e to exit $V(L)$ without hitting o , and a SRW from z to hit o before hitting the LERW. We will bound the right-hand side of (3.2) from below by analyzing this random walk event. By time reversal, we will be able to consider the SRW going from o to z . The lower bound then contains two factors: the LERW and SRW avoiding each other near o , and the SRW subsequently hitting z . This leads to the two probabilities in Theorem 3.1.

We begin by expressing the probability in (3.2) in terms of the random walk construction specified above (Lemma 3.2). We then lower bound the probability of the resulting walk event by something amenable to analysis by walk intersection techniques that we give in Section 3.2. Let $\pi = \mathcal{L}S_e[0, \sigma_L]$.

Lemma 3.2.

(i) We have

$$\mu_{L,o}(A(z, e)) = \mathbb{P} \left(\pi \cap S_z[0, \xi_o] = \emptyset, \xi_o^{S_z} < \sigma_L^{S_z} \right). \tag{3.3}$$

(ii) There are constants $\kappa(d) > 0, d \geq 2$, such that as $L \rightarrow \infty$, we have

$$\mu_{L,o}(e \notin \mathfrak{T}_{L,o}) = \mathbb{P} (o \notin S_e[0, \sigma_{V(L)}]) \sim \begin{cases} \kappa(2)(\log L)^{-1} & \text{when } d = 2; \\ \kappa(d) & \text{when } d \geq 3. \end{cases} \tag{3.4}$$

Proof. Note that $e \notin \mathfrak{T}_{L,o}$ if and only if S_e exits $V(L)$ before hitting o . This implies the equality in (ii). The asymptotics in (ii) for $d = 2$ follow from Theorem 2.13.

Given that the event $e \notin \mathfrak{T}_{L,o}$ has occurred, z will be in $\mathfrak{T}_{L,o}$ if and only if S_z hits o before exiting $V(L)$, and does so avoiding π . This implies statement (i). \square

In the sequel, we make use of the event $\Gamma_{z,L}$, defined as

$$\Gamma_{z,L} = \left\{ \pi \cap V_z(\|z\|/10) = \emptyset, \xi_z^{S_o} < \sigma_{4\|z\|}^{S_o}, \pi \cap S_o[0, \xi_z] = \emptyset \right\}.$$

Lemma 3.3. For all $L \geq 100|z|$, we have

$$\mathbb{P} \left(\pi \cap S_z[0, \xi_o] = \emptyset, \xi_o^{S_z} < \sigma_L^{S_z} \right) \geq \begin{cases} c \mathbb{P}(\Gamma_{z,L}) \log |z|, & d = 2, \\ (2d)^{-1} \mathbb{P}(\Gamma_{z,L}), & d > 2. \end{cases} \tag{3.5}$$

Proof. Using reversibility of the random walk, we can rewrite the probability in the left hand side of (3.5) as follows:

$$\begin{aligned} & \mathbb{P} \left(\pi \cap S_z[0, \xi_o] = \emptyset, \xi_o^{S_z} < \sigma_L^{S_z} \right) \\ &= \mathbb{E} \left(\mathbb{P} \left(\pi \cap S_z[0, \xi_o] = \emptyset, \xi_o^{S_z} < \sigma_L^{S_z} \mid \pi \right) \right) \\ &= \mathbb{E} \left(\frac{G_{V(L) \setminus \pi}(z, z)}{G_{V(L) \setminus \pi}(o, o)} \mathbb{P} \left(\pi \cap S_o[0, \xi_z] = \emptyset, \xi_z^{S_o} < \sigma_L^{S_o} \mid \pi \right) \right) \\ &\geq \mathbb{E} \left(\mathbf{1}_{\pi \cap V_z(\|z\|/10) = \emptyset} \frac{G_{V(L) \setminus \pi}(z, z)}{G_{V(L) \setminus \pi}(o, o)} \mathbb{P} \left(\pi \cap S_o[0, \xi_z] = \emptyset, \xi_z^{S_o} < \sigma_L^{S_o} \mid \pi \right) \right). \end{aligned} \tag{3.6}$$

In the presence of the indicator, (2.8) implies $G_{V(L) \setminus \pi}(z, z) \geq G_{V_z(\|z\|/10)}(z, z)$. Since $e \in \pi$, we also have

$$G_{V(L) \setminus \pi}(o, o) \leq G_{\mathbb{Z}^d \setminus \{e\}}(o, o) \leq 2d,$$

since after each visit to o , the random walk next hits e with probability $(2d)^{-1}$. It follows that the right-hand side of (3.6) is at least

$$\begin{aligned} & (2d)^{-1} G_{V_z(\|z\|/10)}(z, z) \mathbb{E} \left(\mathbf{1}_{\pi \cap V_z(\|z\|/10) = \emptyset} \mathbb{P} \left(\pi \cap S_o[0, \xi_z] = \emptyset, \xi_z^{S_o} < \sigma_L^{S_o} \mid \pi \right) \right) \\ & \geq (2d)^{-1} G_{V_z(\|z\|/10)}(z, z) \mathbb{P}(\Gamma_{z,L}). \end{aligned} \tag{3.7}$$

Using Theorem 2.13 and (2.8), we have

$$G_{V_z(\|z\|/10)}(z, z) \geq \begin{cases} c \log |z|, & d = 2, \\ 1, & d > 2, \end{cases}$$

Inserting this estimates into (3.7) completes the proof. \square

3.2 SRW and LERW steering

In order to control the spanning tree event in (3.2), we need to give a lower bound on the probability of the event $\Gamma_{z,L}$ of Lemma 3.3. This will be achieved by “steering” the two walks so that they become well separated as they reach distance of order $\|z\|$, which allows us to arrange that π avoids the box $V_z(\|z\|/10)$, and it does not influence very much the probability that S_o hits z . In Section 3.2.1 we collect results we need for a single walk. In Section 3.2.2 we prove the required separation estimate. In Section 3.3 we prove a lower bound on $\mathbb{P}(\Gamma_{z,L})$ and complete the proof of Theorem 3.1.

3.2.1 Estimates for a single walk

The first lemma we need, shown by Masson, holds for general d , and compares an infinite LERW to a finite LERW, allowing us to control probabilities by restricting to finite balls. Masson stated this in the case $K \supset B(4n)$, however, his proof applies in the slightly more general case we use here.

Lemma 3.4. [29, Corollary 4.5] *Let $d \geq 2$ be arbitrary. For any $\delta > 0$, we have*

$$\mathbb{P}(\widehat{S}_o[0, \widehat{\sigma}_n] = \alpha) \asymp_\delta \mathbb{P}(\widehat{S}_o^K[0, \widehat{\sigma}_n] = \alpha),$$

for all α , all $n \geq 1/\delta$, and all $K \supset V((1 + \delta)n)$, where the constants implied by the \asymp_δ notation only depend on δ and d .

We will need the following “Boundary Harnack inequality”, to control a LERW after it has reached the boundary of a box. Estimates of this flavour were proved in [29, Proposition 3.5], [34, Proposition 6.1.1], [4] and [5, Section 3]. The variant we need here is a simplified version of [5, Lemma 3.8]. We define

$$H_n = \{x \in \mathbb{Z}^d : x \cdot e_1 = n\}, \quad H_n^+ = \{x \in \mathbb{Z}^d : x \cdot e_1 \geq n\}, \quad H_n^- = \{x \in \mathbb{Z}^d : x \cdot e_1 \leq n\}.$$

Lemma 3.5. *There exists $c(d) > 0$ such that the following holds. Let $\pi \subset V(n/2)$ and $x \in \partial V(n/2) \cap H_{n/2}$. Let $1 \leq m \leq n/4$, and $L \geq 4n$. We have*

$$\mathbb{P}_x(S(\sigma_{V_x(m)}) \in H_{n/2+m} \mid \sigma_L < \bar{\xi}_\pi) \geq c(d) \frac{m}{n}.$$

Proof. Let $\pi' = \pi \cap V_x(m)$. It is shown in [5, Section 3] that

$$\mathbb{P}_x(S(\sigma_{V_x(m)}) \in H_{n/2+m}, \sigma_{V_x(m)} < \bar{\xi}_{\pi'}) \geq \frac{1}{2d} \mathbb{P}_x(\sigma_{V_x(m)} < \bar{\xi}_{\pi'}).$$

This yields

$$\begin{aligned} & \mathbb{P}_x(S(\sigma_{V_x(m)}) \in H_{n/2+m}, \sigma_L < \bar{\xi}_\pi) \\ & \geq \mathbb{P}_x(S(\sigma_{V_x(m)}) \in H_{n/2+m}, \sigma_{V_x(m)} < \bar{\xi}_{\pi'}) \\ & \quad \min_{w \in (\partial V_x(m)) \cap H_{n/2+m}} \mathbb{P}_w(\sigma_{2n} < \xi_{H_{n/2}}) \min_{z \in \partial V(2n)} \mathbb{P}_z(\sigma_L < \xi_\pi) \\ & \geq \frac{1}{2d} \mathbb{P}_x(\sigma_{V_x(m)} < \bar{\xi}_{\pi'}) c \frac{m}{n} \max_{z \in \partial V(2n)} \mathbb{P}_z(\sigma_L < \xi_\pi) \\ & \geq \frac{c(m/n)}{2d} \mathbb{P}_x(\sigma_{V(2n)} < \bar{\xi}_\pi) \max_{z \in \partial V(2n)} \mathbb{P}_z(\sigma_L < \xi_\pi) \geq c(d) \frac{m}{n} \mathbb{P}_x(\sigma_L < \bar{\xi}_\pi). \end{aligned}$$

Here we used a gambler’s ruin estimate and the Harnack principle in the second inequality. □

3.2.2 Separation lemma

Separation lemmas for a loop-erased walk and a simple random walk appeared in [29] for $d = 2$ and in [34] for $d = 3$. We give here a unified proof that works for all $d \geq 2$. Let us write $\widehat{S}_{L,o} = \mathcal{L}S_o[0, \sigma_L]$. Recall that S'_x generally denotes a simple random walk independent of the walk S_x . We define $A_n = \{\widehat{S}_{L,o}(0, \widehat{\sigma}_n] \cap S'_o[0, \sigma_n] = \emptyset\}$, $n \leq L$. Let

$$D_n = \min \left\{ \text{dist}(\widehat{S}_{L,o}(\widehat{\sigma}_n), S'_o[0, \sigma_n]), \text{dist}(\widehat{S}_{L,o}[0, \widehat{\sigma}_n], S'_o(\sigma_n)) \right\}.$$

Lemma 3.6 (Separation Lemma). *Let $d \geq 2$. There exists $\delta = \delta(d) > 0$ and $c = c(d) > 0$ such that for all $1 \leq n \leq L/4$ we have*

$$\mathbb{P}(D_n \geq \delta n \mid A_n) \geq c.$$

In the proof of the separation lemma, a basic step is to show that given that non-intersection occurred to distance $n/2$, there is small probability that separation at distance n is bad.

Lemma 3.7. *Let $d \geq 2$. There exists a function $r : (0, 1/2] \rightarrow (0, \infty)$ with $\lim_{\delta \rightarrow 0} r(\delta) = 0$, and for all $\delta \in (0, 1/2]$ there exists $n_0(\delta)$ such that we have we have*

$$\mathbb{P}(A_n \cap \{D_n < \delta n\} \mid A_{n/2}) \leq r(\delta), \quad n \geq n_0(\delta), 0 < \delta \leq 1/2.$$

Proof. We condition on $\pi_0 := \widehat{S}_{L,o}[0, \widehat{\sigma}_{n/2}]$ and $S'_o[0, \sigma_{n/2}]$, and denote $x_1 = \widehat{S}_{L,o}(\widehat{\sigma}_{n/2})$, $x_2 = \widehat{S}_{L,o}(\widehat{\sigma}_n)$, $y_1 = S'_o(\sigma_{n/2})$, $y_2 = S'_o(\sigma_n)$. We distinguish two cases:

- (a) $S'_o[\sigma_{n/2}, \sigma_n]$ visits $B_{x_2}(\delta n)$;
- (b) $\widehat{S}_{L,o}[\widehat{\sigma}_{n/2}, \widehat{\sigma}_n]$ visits $B_{y_2}(\delta n)$.

We bound the probabilities of the two cases separately, showing that each is bounded by a suitable $r(\delta)$.

Case (a). Let us further condition on x_2 . Since Brownian motion in the cube $\{u \in \mathbb{R}^d : \|u\| \leq 1\}$ has continuous paths, and the path tends to its exit point, and the probability of any given exit point is 0, there exists $r_1(\delta)$, tending to 0, such that given any boundary point w , the Brownian path intersects $\{u \in \mathbb{R}^d : \|u\| \leq 1, |u - w| \leq 2\delta\}$ with probability $\leq r_1(\delta)$. Hence the required bound follows from the invariance principle.

Case (b). Let us condition on y_2 . Due to the Domain Markov Property (Lemma 2.15), the path $\widehat{S}_{L,o}[\widehat{\sigma}_{n/2}, \widehat{\sigma}_L]$ has the law of $\mathcal{L}(X[0, \sigma_L])$, where X has the law of S_{x_1} conditioned on the event $\{\sigma_L < \xi_{\pi_0}\}$. On the event in Case (b), the path $X[\sigma_{3n/4}, \sigma_L]$ has to visit $B_{y_2}(\delta n)$. Conditioning on the point $x' = X(\sigma_{3n/4})$, the probability of this event is

$$\mathbb{P}(X[\sigma_{3n/4}, \sigma_L] \cap B_{\delta n}(y_2) \neq \emptyset \mid X(\sigma_{3n/4}) = x') = \frac{\mathbb{P}_{x'}(\sigma_L < \xi_{\pi_0}, \xi_{B_{y_2}(\delta n)} < \sigma_L)}{\mathbb{P}_{x'}(\sigma_L < \xi_{\pi_0})}. \quad (3.8)$$

When $d \geq 3$, the right hand side is at most

$$\frac{\mathbb{P}_{x'}(\xi_{B_{\delta n}(y_2)} < \infty)}{\mathbb{P}_{x'}(\xi_{B_{n/2}(o)} = \infty)} \leq C\delta^{d-2}.$$

When $d = 2$, consider

$$h(w) := \mathbb{P}_w(\sigma_L < \xi_{\pi_0}, \xi_{B_{y_2}(\delta n)} < \sigma_L) \quad \text{and} \quad w^* := \operatorname{argmax}_{w \in V(2n) \setminus V_{y_2}(n/8)} h(w).$$

Using the Harnack principle, we have

$$\begin{aligned} h(w^*) &\leq \mathbb{P}_{w^*}(\xi_{B_{y_2}(\delta n)} < \sigma_{4n}) \max_{v \in \partial V(4n)} \mathbb{P}_v(\sigma_L < \xi_{\pi_0}) \\ &\quad + \mathbb{P}_{w^*}(\sigma_{4n} < \xi_{\pi_0} \wedge \xi_{B_{y_2}(\delta n)}) \max_{w \in \partial V(2n)} h(w) \\ &\leq C(\log 1/\delta)^{-1} C \mathbb{P}_{x'}(\sigma_L < \xi_{\pi_0}) + c_1 h(w^*), \end{aligned}$$

where $c_1 < 1$. Therefore, we get

$$(1 - c_1)h(x') \leq (1 - c_1)h(w^*) \leq C(\log 1/\delta)^{-1} \mathbb{P}_{x'}(\sigma_L < \xi_{\pi_0}).$$

This completes the proof. \square

The next step is to show that given a “good” separation at distance $n/2$, the probability that the paths can be “very well” separated at distance n is at least a constant. We denote $\mathcal{G}_n := \sigma(\widehat{S}_o^L[0, \widehat{\sigma}_n], S'_o[0, \sigma_n])$ the information about the paths up to the exit from V_n .

Lemma 3.8. *For any $0 < \delta \leq 1/2$ there exists $c(\delta) > 0$ and $n_1(\delta)$ such that for all $n \geq n_1(\delta)$ and $L \geq 4n$ the following hold.*

(i) We have

$$\mathbb{P}\left(A_n \cap \left\{ \widehat{S}_o^L(\widehat{\sigma}_{3n/4}, \widehat{\sigma}_n) \in H_{-n/2}^-, S'_o(\sigma_{3n/4}, \sigma_n) \in H_{n/2}^+ \right\} \middle| \mathcal{G}_{n/2}\right) \geq c(\delta) \quad (3.9)$$

everywhere on the event $A_{n/2} \cap \{D_{n/2} \geq \delta n/2\}$.

(ii) Moreover, we may further require the event $\widehat{S}_o^L[\widehat{\sigma}_{2n}, \widehat{\sigma}_L] \cap V_n = \emptyset$ in (3.9).

Proof. We condition on $\pi_0 = \widehat{S}_o^L[0, \widehat{\sigma}_{n/2}]$ and $S'_o[0, \sigma_{n/2}]$. Let us write $x_1 = \widehat{S}_o^L(\widehat{\sigma}_{n/2})$ and $y_1 = S'_o(\sigma_{n/2})$. Due to the conditioning in (3.9), we have $|x_1 - y_1| \geq \delta(n/2)$. We write X for a random walk starting at x_1 conditioned on the event $\{\sigma_L < \bar{\xi}_{\pi_0}\}$, so that $\widehat{S}_o^L[\widehat{\sigma}_{n/2}, \widehat{\sigma}_L]$ has the law of $\mathcal{L}X[0, \sigma_L]$.

An application of Lemma 3.5 yields that with probability $\geq c\delta$, the process X exits $V_{x_1}(\delta n/8)$ on the face furthest from $V(n/2)$. Also, there is probability $\geq (2d)^{-1}$ that $S'_o[\sigma_{n/2}, \infty)$ exits $V_{y_1}(\delta n/8)$ on the face furthest from $V(n/2)$. Using the Harnack principle for X , and appropriate disjoint corridors of width of order δn for the LERW and the SRW, respectively (see Figure 1 below), there is probability $\geq c(\delta)$ that:

- (a) S'_o exits $V(n)$ in H_n , with the appropriate portion in the required halfspace;
- (b) X exits $V(2n)$ in H_{-2n} with $X[\sigma_{3n/4}, \sigma_{2n}] \subset [-2n, -n/2] \times [-3n/4, 3n/4]^{d-1}$;
- (c) $S'_o[0, \sigma_n] \cap X[0, \sigma_n] = \emptyset$.

In order to further ensure that $\mathcal{L}X$ first exits $V(n)$ at a point in H_{-n} , and that A_n occurs, we show that for $w \in \partial V(2n)$ we have

$$\mathbb{P}_w\left(\sigma_L^X < \xi_{V(n)}^X\right) \geq c > 0. \quad (3.10)$$

This is indeed sufficient, since the events in point (b) and (3.10) imply that the last visit of X to $\partial V(n)$ must occur at a point in H_{-n} . In order to see (3.10), first note that the statement is clear for $d \geq 3$, since then

$$\mathbb{P}_w(\sigma_L < \xi_{V(n)}) \geq \mathbb{P}_w(\xi_{V(n)} = \infty) \geq c \geq c \mathbb{P}_w(\sigma_L < \xi_{\pi_0}).$$

When $d = 2$, let $w_* := \operatorname{argmax}_{w \in \partial V(2n)} \mathbb{P}_w(\sigma_L < \xi_{\pi_0})$. Then we have

$$\mathbb{P}_{w_*}(\sigma_L < \xi_{\pi_0}) \leq \mathbb{P}_{w_*}(\sigma_L < \xi_{V(n)}) + \max_{y \in \partial V(n)} \mathbb{P}_y(\sigma_{2n} < \xi_{\pi_0}) \mathbb{P}_{w_*}(\sigma_L < \xi_{\pi_0}).$$

The maximum over y is $\leq c_2 < 1$, which after rearranging gives

$$\mathbb{P}_{w_*}(\sigma_L < \xi_{\pi_0}) \leq \frac{1}{1 - c_2} \mathbb{P}_{w_*}(\sigma_L < \xi_{V(n)}).$$

For other $w \in \partial V(2n)$ we obtain the statement from the Harnack principle.

The construction we gave ensures both the event in (i) and (ii), and thus completes the proof of the lemma. \square

Lemma 3.9. *Let $d \geq 2$. There exists $c = c(d) > 0$ such that for any $z \in \mathbb{Z}^d \setminus \{o\}$ and $L > 4\|z\|$ we have*

$$\mathbb{P}(\Gamma_{z,L}) \geq \begin{cases} c \operatorname{Es}^L(\|z\|) \mathbb{P}_o(\xi_{\{z\}} < \sigma_{4\|z\|}) & \text{if } d \geq 3; \\ c (\log L)^{-1} \operatorname{Es}^L(\|z\|) \mathbb{P}_o(\xi_{\{z\}} < \sigma_{4\|z\|}) & \text{if } d = 2. \end{cases} \quad (3.13)$$

Proof. We may assume that $\|z\|$ is sufficiently large. Without loss of generality, we also assume that the first coordinate of z is positive and has maximal absolute value among all coordinates.

We first require the occurrence of the event

$$B(e) := \{\pi \cap \{o\} = \emptyset\}. \quad (3.14)$$

We have

$$q_L := \mathbb{P}(B(e)) = \mathbb{P}_e(\sigma_L < \xi_o) \geq \begin{cases} c(\log L)^{-1} & \text{when } d = 2; \\ c & \text{when } d \geq 3. \end{cases} \quad (3.15)$$

Conditional on $B(e)$, the law of π is the same as the law of $\widehat{S}_o^L[1, \widehat{\sigma}_L]$ conditional on $\widehat{S}_o^L(1) = e$. Therefore, we will express properties of π conditional on the event $B(e)$ in terms of the properties of $\widehat{S}_o^L[1, \widehat{\sigma}_L]$ conditional on $\widehat{S}_o^L(1) = e$.

Let us require the occurrence of the event

$$A_{\|z\|/4} \cap \{D_{\|z\|/4} \geq \delta\|z\|/4\}, \quad (3.16)$$

where $\delta = \delta_0$ is the constant chosen in Lemma 3.6. According to that lemma, the event in (3.16) has unconditional probability $\geq c \operatorname{Es}^L(\|z\|/4) \geq c \operatorname{Es}^L(\|z\|)$. Due to \mathbb{Z}^d -symmetry, the conditional probability of (3.16) given $\widehat{S}_o^L(1) = e$ is the same as the unconditional probability.

Let us further require the event in Lemma 3.8 with $n = \|z\|/2$, that is, that the paths extend disjointly to opposite faces of $V(\|z\|/2)$, with the random walk landing on $H_{\|z\|/2}$, and the LERW landing on $H_{-\|z\|/2}$. According to Lemma 3.8, this happens with conditional probability $\geq c$. It follows from Lemma 3.8, that there is probability bounded away from 0 that the LERW can be further extended to land on $H_{-8\|z\|}$, in such a way that π is contained in $H_{3\|z\|/8}^- \cup V(4\|z\|)^c$. Since $S'_o(\sigma_{\|z\|/2}) \in H_{\|z\|/2}$, the conditional probability, given π and $S'_o[0, \sigma_{\|z\|/2}]$ that S'_o hits z before $\xi_\pi \wedge \sigma_{4\|z\|}$ is $\geq c|z|^{2-d}$ when $d \geq 3$, and $\geq (\log|z|)^{-1}$ when $d = 2$. Combining the estimates for each part of the construction yields:

$$\mathbb{P}(\Gamma_{z,L}) \geq \begin{cases} c \operatorname{Es}^L(\|z\|) |z|^{2-d} & \text{when } d \geq 3; \\ c (\log L)^{-1} \operatorname{Es}^L(\|z\|) \frac{1}{\log|z|} & \text{when } d = 2. \end{cases}$$

This completes the proof of the lemma when $\|z\|$ is sufficiently large. □

The following proposition summarizes the result of the finite L arguments we made in Sections 3.1–3.2.

Proposition 3.10. *Let $d \geq 2$. For any $z \in \mathbb{Z}^d$ and all $L \geq 4\|z\|$ we have*

$$\begin{aligned} \nu_L(z \in \operatorname{Av}) &\geq c \operatorname{Es}^L(\|z\|) G_{V_z(\|z\|/10)}(z, z) \mathbb{P}_o(\xi_z < \sigma_{4\|z\|}) \\ &\asymp \operatorname{Es}^{\|z\|}(\|z\|) \mathbb{P}_o(\xi_z < \infty). \end{aligned} \quad (3.17)$$

Proof. Combining (3.2), Lemma 3.2, Lemma 3.3, and Lemma 3.9, we obtain the statement in both $d \geq 3$ and $d = 2$. Lemma 3.4 implies that $\operatorname{Es}^L(\|z\|)$ is comparable to $\operatorname{Es}^{4\|z\|}(\|z\|)$. Masson [29, Lemma 5.1] showed that the latter is comparable to $\operatorname{Es}^{\|z\|}(\|z\|)$. □

3.3 Proof of Theorem 1.1

In passing to the limit $L \rightarrow \infty$, we use the following proposition. We prove only the $d \geq 3$ case here; the proof of the more technical $d = 2$ case is deferred to the end of Section 5; see Lemma 5.10 there.

Proposition 3.11.

Assume $d \geq 2$. Then we have $\nu(z \in Av) = \lim_{L \rightarrow \infty} \nu_L(z \in Av)$.

Proof of Proposition 3.11, $d \geq 3$ case. Due to [15, Theorem 3.11] we have $\nu(|Av| < \infty) = 1$. Therefore, given $\varepsilon > 0$, we can find $|z| < M < \infty$ such that $\nu(Av \subset V(M)) > 1 - \varepsilon$. Due to the weak convergence $\nu_L \rightarrow \nu$, there exists $M < L_0 < \infty$ such that for all $L \geq L_0$ we have

$$\begin{aligned} |\nu(z \in Av) - \nu_L(z \in Av)| &\leq |\nu(z \in Av) - \nu(z \in Av, Av \subset V(M))| \\ &\quad + |\nu(z \in Av, Av \subset V(M)) - \nu_L(z \in Av, Av \subset V(M))| \\ &\quad + |\nu_L(z \in Av, Av \subset V(M)) - \nu_L(z \in Av)| \\ &< 3\varepsilon. \end{aligned}$$

This completes the proof. □

We can now complete the proof of Theorem 3.1.

Proof of Theorem 3.1. Lemma 3.4 implies that $\text{Es}^L(\|z\|)$ in Proposition 3.11 is comparable to the avoidance probability in the theorem, so the statement follows using Proposition 3.11. □

For later use we state here a corollary of the construction.

Corollary 3.12. Let $d \geq 2$. There exists a constant $c = c(d) > 0$ such that the following holds. For each $z \in \mathbb{Z}^d$ and $L \geq 4\|z\|$ and $e \sim o$, with π denoting the path in $\mathfrak{T}_{L,s}$ connecting e to s , we have

$$\begin{aligned} \mu_{L,o}(z \in \mathfrak{T}_{L,o}, \pi \cap V_z(\|z\|/10) = \emptyset \mid e \notin \mathfrak{T}_{L,o}) \\ \geq c \mathbb{P}(S_o[0, \sigma_{|z|}] \cap \widehat{S}'_o(0, \widehat{\sigma}_{|z|}] = \emptyset) \mathbb{P}_o(\xi_z < \infty). \end{aligned}$$

We now turn to the proofs of the explicit bounds for $2 \leq d \leq 4$.

3.3.1 Proof of Theorem 1.1 when $d = 2$

By Theorem 3.1, Theorem 1.1(i) will hold once we know

$$\mathbb{P}\left(S'_o[0, \sigma_n] \cap \widehat{S}(0, \widehat{\sigma}_n] = \emptyset\right) \geq n^{-3/4+o(1)}. \tag{3.18}$$

The exponent $3/4 + o(1)$ was first proved by Kenyon [18], who stated it for simple random walk in the half plane. A proof for more general walks was given by Masson, who derived it from results on SLE₂ [29, Theorem 5.7]. He established the analogue of (3.18) for a SRW and a finite LERW – via [29, Lemma 5.1], where the intersection probabilities for finite and infinite LERW are related. This implies Theorem 1.1(i). □

3.3.2 Proof of Theorem 1.1 when $d = 3$

In this section, we complete the proof of the explicit lower bound in Theorem 1.1(ii) by showing that $\mathbb{P}(S'_o[0, \sigma(n)] \cap \widehat{S}_o(0, \widehat{\sigma}(n)] = \emptyset) \geq cn^{-2\zeta}$ (this suffices, by the previously proven Theorem 3.1). Since \widehat{S}_o is the loop-erasure of S , it is enough to show

$$\mathbb{P}(S'_o[0, \sigma(n)] \cap S(0, \infty) = \emptyset) \geq cn^{-2\zeta}. \tag{3.19}$$

This is a simple adaptation of the results of [22]. Indeed, there exists a $c_1 > 0$ such that (uniformly in m)

$$\mathbb{P}(S'_o[0, m] \cap S(0, \infty) = \emptyset) \geq c_1 m^{-\zeta}$$

by [22, discussion after (3)]. On the other hand, by [22, Lemma 4.7], there exists $C_2, c_2 > 0$ such that (uniformly in a, n)

$$\mathbb{P}(S'_o[0, \sigma_n] \cap S(0, \infty) = \emptyset, \sigma_n > an^2) \leq C_2 \exp(-a/c_2)n^{-2\zeta}.$$

Choosing a sufficiently large (relative to c_1, c_2) and $m = an^2$ completes the proof of (3.19). From this Theorem 1.1(ii) follows. \square

3.3.3 Proof of Theorem 1.1 when $d = 4$

In four dimensions, the avoidance probability was determined by Lawler [21], who showed that

$$\text{Es}(n) \asymp (\log n)^{-1/3}.$$

This and Theorem 3.1 yield Theorem 1.1(iii).

4 Low-dimensional radius bounds

In this section we prove the radius bounds stated in Theorem 1.2(i)–(iii) for dimensions $2 \leq d \leq 4$. We start with the lower bounds.

Proof of Theorem 1.2(i)–(iii), lower bounds. Observe that

$$\nu(R \geq r) \geq \nu(re_1 \in \text{Av}). \tag{4.1}$$

Therefore, the claimed lower bounds follow immediately from Theorem 1.1(i)–(iii). \square

Proof of Theorem 1.2(ii)–(iii), upper bounds. Let us first consider a finite volume $V(L)$. Recall that for $\eta \in \mathcal{R}_L$ we write $\alpha(\eta) = (\eta_1, \dots, \eta_N)$ for the waves in the stabilization $S_o(\eta + \mathbf{1}_o)$. Let us extend the notation for the radius to waves and two-component spanning trees in the natural way:

$$\begin{aligned} R(\eta_*) &= \sup\{|z| : z \in \mathcal{W}(\eta_*)\} \\ R(T_o) &= \sup\{|z| : z \in T_o\} \end{aligned}$$

Under the bijection of Section 2.5 these two notions coincide. We have

$$\begin{aligned} |\{\eta \in \mathcal{R}_L : R(\eta) > r\}| &= |\{\eta \in \mathcal{R}_L : R(\eta_i) > r \text{ for some } 1 \leq i \leq N(\eta)\}| \\ &\leq |\{\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L : R(\eta_*) > r\}|. \end{aligned}$$

Hence we get

$$\nu_L(R > r) \leq \frac{|\mathcal{R}'_L \setminus \mathcal{R}_L|}{|\mathcal{R}_L|} \mu_{L,o}(R(\mathfrak{T}_{L,o}) > r) = g_L(o, o) \mu_{L,o}(R(\mathfrak{T}_{L,o}) > r),$$

where the last equality uses Lemma 2.6.

Since $\{R > r\}$ and $\{R(\mathfrak{T}_{L,o}) > r\}$ are both cylinder events, we can take the limit $L \rightarrow \infty$ on both sides to get

$$\nu(R > r) \leq C(d) \mathbf{WSF}_o(R(\mathfrak{T}_{L,o}) > r). \tag{4.2}$$

Lyons, Morris and Schramm [25] proved that

$$\mathbf{WSF}_o(R(\mathfrak{T}_{L,o}) > r) \leq C(d) r^{-\beta_d}, \quad d \geq 3,$$

with $\beta_d = \frac{1}{2} - \frac{1}{d}$. Inserting this into (4.2) yields the upper bounds $C r^{-1/6}$ and $C r^{-1/4}$ in dimensions $d = 3$ and $d = 4$, respectively. \square

5 The last k waves in 2D

In this section we prove that for any $1 \leq k < \infty$, as $L \rightarrow \infty$, the last k waves on G_L (when they exist) have a weak limit. We introduce some notation for the last k waves. Recall that for $\eta \in \mathcal{R}_L$, we denote by $N = N(\eta) = n_L(o, o)$ the number of times o topples during the stabilization of $\eta + \mathbf{1}_o$. Recall from Section 2.4, that given $\eta \in \mathcal{R}_L$, we write $\alpha(\eta) = (\eta_1, \dots, \eta_N)$ for the set of intermediate configurations (right before each wave). Let

$$\xi_0^L = \xi_0^L(\eta) := a_{o,L}\eta$$

and whenever $N(\eta) \geq k$, define

$$\xi_k^L = \xi_k^L(\eta) := \eta_{N-k+1}.$$

Observe that when ξ_k^L is defined, we have $\xi_{k-1}^L = a'_{o,L}\xi_k^L$. Analogously to the finite graph case, a'_{o,\mathbb{Z}^2} is the operator which adds a particle at the origin and stabilizes the resulting configuration in the graph $(\mathbb{Z}^2)'$.

Theorem 5.1. *Assume $d = 2$. We have:*

- (i) *For every $1 \leq k < \infty$ the limit $b_k = \lim_{L \rightarrow \infty} \nu_L(N \geq k)$ exists.*
- (ii) *For every $1 \leq k < \infty$, the law of ξ_k^L under the measure $\nu_L(\cdot | N \geq k)$ converges weakly to the law ρ_k of a configuration ξ_k .*
- (iii) *The configuration $\xi_k + \mathbf{1}_o$ can be stabilized in $(\mathbb{Z}^2)'$ with finitely many topplings ρ_k -a.s.*
- (iv) *The transformation $\xi_k \mapsto a'_{o,\mathbb{Z}^2}\xi_k$ is measure-preserving between $b_k\rho_k$ and the restriction of $b_{k-1}\rho_{k-1}$ to the image. (Here $b_0 = 1, \rho_0 := \nu$.)*
- (v) *With ν -probability 1 on the event $\{N = k\}$, all k waves are finite, and we have $\nu(N = k) = b_k - b_{k+1}$.*

Remark 5.2. It is not difficult to construct, for every $1 \leq k < \infty$, an explicit finite configuration around o showing that $\liminf_{L \rightarrow \infty} \nu_L(N = k) > 0$.

Recall the bijection for intermediate configurations from Section 2.5. This bijection was between $\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L$ and the set of spanning forests of G_L with two components $T_o = T_o(\eta_*)$ and $T_s = T_s(\eta_*)$, where $o \in T_o$ and $s \in T_s$. Recall the following property of the bijection from Section 2.5, rephrased for the configurations ξ_k^L .

$$\begin{aligned} &\text{If there is a path in } T_o(\xi_k^L(\eta)) \text{ from } o \text{ to a vertex } x \text{ that stays} \\ &\text{inside } B(r), \text{ then starting from } \xi_k^L + \mathbf{1}_o \text{ there is a sequence of} \qquad (5.1) \\ &\text{topplings in } B(r) \text{ that topples } x. \end{aligned}$$

We write $\mathfrak{T}_{L,o,k} = T_o(\xi_k^L)$ and $\mathfrak{T}_{L,s,k} = T_s(\xi_k^L)$ for short. When $x \in \mathfrak{T}_{L,o,k}$, we denote $\pi_{L,k}(x)$ the unique self-avoiding path in $\mathfrak{T}_{L,o,k}$ from o to x .

Our control on the size of waves will be in terms of the following random variables.

$$\begin{aligned} R_{\text{in},k}^L &= \sup\{r \geq 0 : B(r) \subset \mathfrak{T}_{L,o,k}\}, \quad k \geq 1; \\ R_{\text{out},k}^L &= \inf\{r \geq 0 : \mathfrak{T}_{L,o,k} \subset B(r)\}, \quad k \geq 1; \\ P_k^L &= \inf\left\{r \geq 0 : \pi_{L,k}(x) \subset B(r) \text{ for all } x \in B\left(R_{\text{in},(k-1)}^L + 1\right)\right\}, \quad k \geq 2. \end{aligned} \tag{5.2}$$

All quantities in (5.2) are defined to be 0 when $N < k$. The following lemma states a basic inequality we will need.

Lemma 5.3. *We have:*

- (i) $R_{\text{in},1}^L = 0$.
- (ii) $R_{\text{in},k}^L \leq P_k^L, k \geq 2$.

Proof. (i) This follows directly from Corollary 2.11.

(ii) Write $r = P_k^L$ for short, and assume for a proof by contradiction, that we had $R_{\text{in},k}^L \geq r + 1$. This implies that in the stabilization $a'_{o,L}(\xi_k^L)$ all vertices in $B(r + 1)$ topple, and hence $(\xi_{k-1}^L)_{B(r)} = (\xi_k^L)_{B(r)}$. Starting from the configuration $\xi_{k-1}^L + \mathbf{1}_o$, let us topple all sites in $B(r)$ we can. The definition of P_k^L and property (5.1) imply that all vertices in $B(R_{\text{in},(k-1)}^L + 1)$ topple. However, this contradicts the definition of $R_{\text{in},(k-1)}^L$. \square

In the following proposition we show that the in-radius of the k -th last wave is tight, with a power law upper bound on the tail.

Proposition 5.4. *There exist constants $\alpha'_1 > \alpha'_2 > \dots > 0$ and C_1, C_2, \dots such that*

$$\limsup_{L \rightarrow \infty} \nu_L(R_{\text{in},k}^L > r) \leq C_k r^{-\alpha'_k}, \quad \forall r \geq 1, \forall k \geq 1. \tag{5.3}$$

In particular, for all $1 \leq k < \infty$, the sequence $\{R_{\text{in},k}^L\}_{L \geq 1}$ is tight.

Proof. We prove the statement by induction on k . The case $k = 1$ holds trivially due to Lemma 5.3(i). Assume $k \geq 2$, and that (5.3) holds for $k - 1$. Let $1 \leq r_0 < \infty$ be fixed, and find $L_0 = L_0(r_0) < \infty$ such that for $L \geq L_0$ we have

$$\nu_L(R_{\text{in},(k-1)}^L > r_0) \leq 2C_{k-1} r_0^{-\alpha'_{k-1}}. \tag{5.4}$$

It is sufficient to bound $R_{\text{in},k}^L$ when the event $\{R_{\text{in},(k-1)}^L \leq r_0\}$ occurs, and due to Lemma 5.3(ii), it is enough to bound P_k^L on this event. In what follows, we assume the event $\{R_{\text{in},(k-1)}^L \leq r_0\}$.

For $\ell \geq 1$ we are going to bound the probability that $r_0 2^\ell < P_k^L \leq r_0 2^{\ell+1}$. Due to Lemma 5.3(ii), this event implies that $(\mathfrak{T}_{L,o,k}, \mathfrak{T}_{L,s,k})$ belongs to the following event $E(x, r_0, \ell)$ for some $x \in \partial B(r_0 + 1)$:

$$E(x, r_0, \ell) = \{(T_o, T_s) \in \mathcal{T}_{L,o} : x \in T_o, \pi(x) \text{ visits } B(r_0 2^\ell)^c \text{ and } T_s \cap \partial B(r_0 2^{\ell+1}) \neq \emptyset\},$$

where $\pi(x)$ is the path in T_o from x to o . Therefore, using Corollary 2.7(ii), we have

$$\begin{aligned} \nu_L \left(R_{\text{in},(k-1)}^L \leq r_0, P_k^L > r_0 2^{\ell_0} \right) &\leq \sum_{\ell \geq \ell_0} \frac{|\{\eta \in \mathcal{R}_L : r_0 2^\ell < P_k^L(\eta) \leq r_0 2^{\ell+1}\}|}{|\mathcal{R}_L|} \\ &\leq C (\log L) \sum_{\ell \geq \ell_0} \sum_{x \in \partial B(r_0+1)} \mu_{L,o}(E(x, r_0, \ell)). \end{aligned} \tag{5.5}$$

We use Wilson's algorithm to get an upper bound on the probability of $E(x, r_0, \ell)$. Let the first random walk start at x . Let τ be the time of the last visit, before $\xi_{\{o\}}$, to a vertex in $\partial B(r_0 2^\ell)$. Let us condition on the path $S_x[0, \tau]$. Let $\gamma = \mathcal{L}S_x[0, \tau]$, and let γ_0 be the initial segment of γ from x to the first visit of γ to $\partial B(r_0 2^\ell)$. The walk

$$S'(m) = S_x(\tau + m), \quad m = 0, \dots, \xi_o - \tau;$$

is a simple random walk on \mathbb{Z}^2 conditioned on $\xi_o < \bar{\xi}_{B(r_0 2^\ell)^c}$. On the event $E(x, r_0, \ell)$, S' cannot hit γ_0 , so we bound the probability that S' hits $B(r_0 + 1)$ before γ_0 .

The walk S' has to successively cross from $\partial B(r_0 2^q)$ to $B(r_0 2^{q-1})$ for $q = \ell - 1, \dots, 1$. During each crossing, it has a fixed constant probability of hitting γ_0 , since this holds for simple random walk, and the Harnack principle [24] then implies it holds for S' . Hence the probability that S' reaches $B(r_0 + 1)$ before hitting γ_0 is less than $(1 - c_1)^{\ell-1}$ for some $0 < c_1 < 1$. This bounds from above the probability that $x \in \mathfrak{T}_{L,o}$ and $\pi(x)$ visits $B(r_0 2^\ell)^c$. Assuming that this event occurs, we now bound the conditional probability that $\mathfrak{T}_{L,s}$

contains a vertex $y \in \partial B(r_0 2^{\ell+1})$. For this, continue Wilson's algorithm with a walk S_{y_0} starting at any $y_0 \in \partial B((r_0 2^{\ell+1})^4)$, followed by walks starting at y_1, \dots, y_M , where the latter is an enumeration of all vertices in $\partial B(r_0 2^{\ell+1})$. Let \mathfrak{F}_j denote the tree generated by the walks $S_x, S_{y_0}, \dots, S_{y_j}$. Denote

$$E_j = \left\{ \sigma_L^{S_{y_j}} < \xi_{\mathfrak{F}_{j-1}}^{S_{y_j}} \right\}, \quad j = 0, 1, \dots, M,$$

where $\mathfrak{F}_{-1} := \pi(x)$. Then on the event $\{x \in \mathfrak{T}_{L,o}, \pi(x) \cap B(r_0 2^\ell)^c \neq \emptyset\}$ we have

$$\mu_{L,o}(\mathfrak{T}_{L,s} \cap \partial B(r_0 2^{\ell+1}) \neq \emptyset \mid \mathfrak{F}_{-1}) \leq \sum_{j=0}^M \mathbb{E} \left(\mathbb{P}(E_j \mid \mathfrak{F}_{j-1}) \mathbf{1}_{E_0^c} \dots \mathbf{1}_{E_{j-1}^c} \right).$$

Application of Theorem 2.13(ii) yields that the $j = 0$ term is at most $C \log(r_0 2^{\ell+1})^4 / \log L$ when $L > (r_0 2^{\ell+1})^4$. For $1 \leq j \leq M$, using Beurling's estimate (Lemma 2.14), on the event $E_0^c \cap \dots \cap E_{j-1}^c$ we have

$$\begin{aligned} \mathbb{P}(E_j \mid \mathfrak{F}_{j-1}) &\leq \mathbb{P}_{y_j}(\sigma_{B((r_0 2^{\ell+1})^4)} < \xi_{\mathfrak{F}_0}) \max_{w \in \partial B((r_0 2^{\ell+1})^4)} \mathbb{P}_w(\sigma_L < \xi_o) \\ &\leq C(r_0 2^{\ell+1})^{-1} (\log r_0 2^{\ell+1}) / \log L. \end{aligned}$$

Since $M \leq C r_0 2^{\ell+1}$, putting the $j = 0$ and $1 \leq j \leq M$ cases together we get that the sum over $0 \leq j \leq M$ is bounded by $C (\log r_0 2^{\ell+1}) / \log L$. Together with the earlier bound on $\pi(x)$ leaving $B_o(r_0 2^\ell)$ this gives

$$\mathbb{P}(E(x, r_0, \ell)) \leq C (\log r_0 2^{\ell+1}) \frac{(1 - c_1)^\ell}{\log L}.$$

Substituting into (5.5), and summing over $\ell \geq \ell_0$ implies, for L sufficiently large that

$$\nu_L(R_{\text{in},(k-1)}^L \leq r_0, P_k^L > r_0 2^{\ell_0}) \leq C r_0 (\log r_0 2^{\ell_0}) (1 - c_1)^{\ell_0}. \tag{5.6}$$

We apply (5.6) with $2^{\ell_0} = r_0^{\beta'}$, for some $\beta' > 0$. The expressions in the right hand sides of (5.6) and (5.4) are of equal order (up to logarithms), when

$$\beta' = -(1 + \alpha'_{k-1}) \frac{\log 2}{\log(1 - c_1)}.$$

Since $R_{\text{in}}^{(k)} \leq P^{(k)}$, the bounds (5.4) and (5.6) imply, for all large enough L , that

$$\nu_L(R_{\text{in},k}^L > r_0^{1+\beta'}) \leq \nu_L(R_{\text{in},k-1}^L > r_0) + \nu_L(R_{\text{in},k-1}^L \leq r_0, P^{(k)} > r_0^{1+\beta'}) \leq C_k (\log r_0) r_0^{-\alpha'_{k-1}}.$$

Hence we get (5.3) for k with a choice of $0 < \alpha'_k < \alpha'_{k-1} / (1 + \beta')$. □

We next prove that the out-radius of the k -th last wave is also tight, and satisfies a power law upper bound. We are going to need the following lemma.

Lemma 5.5. *There exist constants C such that the following holds. Let $1 \leq r < r' < L$, and let $K \subset V(L) \cup \{s\}$ be a connected set of edges that contains a path connecting $B(r)$ to s . We have*

$$\mu_{L,o}(\mathfrak{T}_{L,o} \not\subset B(r') \mid K \subset \mathfrak{T}_{L,s}) \leq C r^{3/2} (r')^{-1/2}.$$

Proof. Let us use Wilson's algorithm in the contracted graph G_L/K , that is, the edges in K are already present at the start of the algorithm. We let walks start at $\{x_1, \dots, x_M\} = \partial B(r)$. If $\mathfrak{T}_{L,o} \not\subset B(r')$, then at least one of these walks has to reach $\partial B(r')$ before hitting K . Beurling's estimate (Lemma 2.14) implies that for each x_j , this has probability at most $C (r'/r)^{-1/2}$. Since $M = O(r)$, the statement follows. □

Proposition 5.6. *There exist constants $\alpha_1 > \alpha_2 > \dots > 0$ and C_1, C_2, \dots such that*

$$\limsup_{L \rightarrow \infty} \nu_L(R_{\text{out},k}^L > r) \leq C_k r^{-\alpha_k}, \quad \forall r \geq 1, \forall k \geq 1. \tag{5.7}$$

In particular, for all $1 \leq k < \infty$ the sequence $\{R_{\text{out},k}^L\}_{L \geq 1}$ is tight.

Proof. Fix $1 \leq k < \infty$, and $1 \leq r_0 < \infty$. From Proposition 5.4 we have that there exists $L_0 = L_0(r_0) < \infty$ such that for all $L \geq L_0$ we have

$$\nu_L(R_{\text{in},k}^L > r_0) \leq 2 C_k r_0^{-\alpha'_k} \tag{5.8}$$

Assume the event $\{R_{\text{in},k}^L \leq r_0\}$, which implies that $\mathfrak{T}_{L,s,k} \cap \partial B(r_0) \neq \emptyset$. We bound the probability that $R_{\text{out},k}^L > r_0^{1+\beta}$, where the parameter $\beta > 0$ will be chosen at the end.

Similarly to (5.5), we have:

$$\nu_L \left(R_{\text{in},k}^L \leq r_0, R_{\text{out},k}^L > r_0^{1+\beta} \right) \leq C (\log L) \sum_{x \in \partial B(r_0)} \mu_{L,o} \left(x \in \mathfrak{T}_{L,s}, \mathfrak{T}_{L,o} \not\subset B(r_0^{1+\beta}) \right). \tag{5.9}$$

Let \mathcal{K} be the set of edges in $\mathfrak{T}_{L,s}$ on the path from x to s . Conditioning on the value $\mathcal{K} = K$, the right hand side of (5.9) equals

$$C (\log L) \sum_{x \in \partial B(r_0)} \mu_{L,o} (x \in \mathfrak{T}_{L,s}) \sum_K \mu_{L,o} (\mathcal{K} = K | x \in \mathfrak{T}_{L,s}) \mu_{L,o} \left(T_o \not\subset B(r_0^{1+\beta}) \mid \mathcal{K} = K \right). \tag{5.10}$$

We have $\mu_{L,o}(x \in \mathfrak{T}_{L,s}) \leq C(\log r_0)/(\log L)$. Applying Lemma 5.5 to the conditional probability in (5.10) gives that the expression in (5.10) is at most

$$C (\log r_0) r_0^{3/2} r_0^{-(1+\beta)/2} \sum_{x \in \partial B(r_0)} \sum_K \mu_{L,o} (\mathcal{K} = K | x \in \mathfrak{T}_{L,s}) = C (\log r_0) r_0^{2-\beta/2}. \tag{5.11}$$

Choose β so that $2 - \beta/2 < -\alpha'_k$, so that (5.11) together with (5.8) gives

$$\nu_L(R_{\text{out},k}^L > r_0^{1+\beta}) \leq C_k r_0^{-\alpha'_k}.$$

This implies the statement of the proposition with $\alpha_k = \alpha'_k/(1 + \beta)$. □

The next proposition shows that $\mu_{L,o}(A \subset \mathfrak{T}_{L,s}, B \subset \mathfrak{T}_{L,o})$ for fixed finite sets of vertices A and B , satisfies a certain normalization as $L \rightarrow \infty$. Tightness of the in-radius established in Proposition 5.4 will allow us to apply this proposition, and subsequently prove Theorem 5.1. We introduce the notation $q_L := \mu_{L,o}(e \notin \mathfrak{T}_{L,o})$, where $e \sim o$. Due to symmetry, q_L does not depend on e . In fact, since q_L is the escape probability of random walk from o , we have $q_L = G_L(o, o)^{-1}$. We remark that for the square grid, the quantity $G_L(o, o)$ has an explicit formula:

$$G_L(o, o) = \sum_{\ell=0}^L \frac{T'_{L+1}(2 - \cos \theta_\ell)}{T_{L+1}(2 - \cos \theta_\ell)}, \tag{5.12}$$

where $\theta_\ell = 2\pi(2\ell + 1)/(4L + 4)$, and T_{L+1} is the degree $L + 1$ Tchebyshev polynomial. The formula (5.12) can be derived via contour integration. However, we will not need it, and we omit the proof.

Proposition 5.7. *Assume $d = 2$. Let $A, B \subset \mathbb{Z}^2$ be disjoint, non-empty finite sets, with $o \in B$. Then the limit $p_{A,B} := \lim_{L \rightarrow \infty} q_L^{-1} \mu_{L,o}(A \subset \mathfrak{T}_{L,s}, B \subset \mathfrak{T}_{L,o})$ exists.*

We first need the following lemma. In its statement, $a(x)$ is the potential kernel for simple random walk on \mathbb{Z}^2 ; see [24].

Lemma 5.8. Fix $x \in A$.

(i) We have $\lim_{L \rightarrow \infty} q_L^{-1} \mu_{L,o}(x \notin \mathfrak{T}_{L,o}) = \frac{a(x)}{a(e)}$, where $e \sim o$.

(ii) Conditional on $x \notin \mathfrak{T}_{L,o}$, the law of the path from x to s has a weak limit, as $L \rightarrow \infty$.

Proof. (i) We use Wilson’s algorithm in the graph $G_{L,o}$ with a walk starting at x . Considering the limit of the bounded martingale $\{a(S_x(t \wedge \sigma_{V(L)} \wedge \xi_{\{o\}}))\}_{t \geq 0}$, and using Lemma 3.2 we have

$$q_L^{-1} \mu_{L,o}(x \notin \mathfrak{T}_{L,o}) = \frac{\mathbb{P}(o \notin S_x[0, \sigma_{V(L)}])}{\mathbb{P}(o \notin S_e[0, \sigma_{V(L)}])} = \frac{a(x)(\log L)^{-1} + o((\log L)^{-1})}{a(e)(\log L)^{-1} + o((\log L)^{-1})} = \frac{a(x)}{a(e)} + o(1).$$

(ii) Let $S_x^{h,L}$ denote a random walk conditioned on the event $\{\sigma_L < \xi_o\}$, started from x . The path in $\mathfrak{T}_{L,s}$ from x to s , conditional on $x \notin \mathfrak{T}_{L,o}$ has the law of $\mathcal{L}S_x^{h,L}[0, \sigma_L]$. As $L \rightarrow \infty$, $S_x^{h,L}$ converges weakly to a transient process S_x^h (the h -transform of random walk by the potential kernel $a(\cdot)$). This implies that $\mathcal{L}S_x^{h,L}[0, \sigma_L]$ converges weakly to $\mathcal{L}S_x^h[0, \infty)$; see [24, Exercise 11.2]. \square

Proof of Proposition 5.7. Let $A = \{x_1, \dots, x_p\}$, ($p \geq 1$) and $B = \{o, w_1, \dots, w_q\}$, ($q \geq 0$). We use Wilson’s algorithm in the graph $G_{L,o}$. We start with the vertex x_1 , followed by the vertices x_2, \dots, x_p , followed by the vertices in B . For the rest of the vertices we use an ordering such that their Euclidean norms form a non-decreasing sequence. Due to Lemma 5.8(i), the probability that the first walk hits s before o is asymptotic to $q_L a(x_1)/a(e)$, as $L \rightarrow \infty$. Assuming this event happens, let $\pi_{x_1}^L$ denote the loop-erasure of the walk starting at x_1 , and write π_{x_1} for its weak limit under the conditioning, whose existence is guaranteed by Lemma 5.8(ii). The probability that the walks starting in $(A \setminus \{x_1\}) \cup B$ terminate before exiting a ball $B(r)$ of large radius r goes to 1 as $r \rightarrow \infty$, and $L > r$, uniformly in the path $\pi_{x_1}^L$. Since these walks determine the event $\{A \subset \mathfrak{T}_{L,s}, B \subset \mathfrak{T}_{L,o}\}$, statement (i) follows. \square

In the proof of Theorem 5.1 we are going to need the following quantitative bound from [10] on the rate of convergence of ν_L to ν .

Theorem 5.9 (Theorem 4.1 of [10]). Let $d = 2$. There exist constants $0 < \alpha < \beta$ and C such that if E is any cylinder event depending only on the configuration in $B(\ell)$, then

$$|\nu_L(E) - \nu(E)| \leq C \ell^\beta L^{-\alpha}.$$

Proof of Theorem 5.1. (i)–(ii) We showed in Proposition 5.6 that for any fixed $1 \leq k < \infty$, the sequence $R_{\text{out}}^{(k)} = R_{L,\text{out}}^{(k)}$, $L \geq 1$, is tight. Therefore, we have

$$\lim_{\ell \rightarrow \infty} \limsup_{L \rightarrow \infty} \nu_L(N \geq k, \mathcal{W}_{N-i+1} \not\subset B(\ell) \text{ for some } 1 \leq i \leq k) = 0. \quad (5.13)$$

We establish weak convergence of ξ_k^L . Fix $\varepsilon > 0$, and let ℓ and L_0 be such that for all $L \geq L_0$ the probability appearing in (5.13) is $\leq \varepsilon$. Let $\zeta \in \mathcal{R}'_{B(\ell)}$ be a configuration with the following properties:

- (a) $(a'_{o,B(\ell)})^j(\zeta) \in \mathcal{R}'_{B(\ell)} \setminus \mathcal{R}_{B(\ell)}$ for $j = 1, \dots, k - 1$ and $(a'_{o,B(\ell)})^k(\zeta) \in \mathcal{R}_{B(\ell)}$.
- (b) In the stabilization $(a'_{o,B(\ell)})^k(\zeta)$ none of the boundary vertices of $B(\ell)$ topple;

In other words, ζ is an intermediate configuration in $B(\ell)$, corresponding to a k -th last wave such that all of the last k waves stay inside $B(\ell)$. We first show that

$$\lim_{L \rightarrow \infty} \nu_L((\xi_k^L)_{B(\ell)} = \zeta \mid N \geq k)$$

exists for any ζ satisfying (a)–(c).

First observe that the properties of ζ imply that for any $\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L$, if $(\eta_*)_{B(\ell)} = \zeta$ then η_* is an intermediate configuration, in the graph G_L , corresponding to a k -th last wave, and the last k waves all stay in $B(\ell)$. In particular, using that the k -th last wave stays inside $B(\ell)$, property (5.1) of the bijection implies that ζ determines a unique pair (A_0, B_0) , with $A_0 \cup B_0 = B(\ell)$, such that whenever $\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L$ and $\eta_*|_{B(\ell)} = \zeta$, then we have $V(T_o(\eta_*)) = B_0$. Therefore, using Lemma 2.6 we can write

$$\begin{aligned} \nu_L(N \geq k, \xi_k^L|_{B(\ell)} = \zeta) &= \frac{|\{\eta_* \in \mathcal{R}'_L \setminus \mathcal{R}_L : \eta_*|_{B(\ell)} = \zeta\}|}{|\mathcal{R}_L|} \\ &= g_L(o, o) \nu_{L,o}(\eta_* : \eta_*|_{B(\ell)} = \zeta). \end{aligned} \tag{5.14}$$

For any η_* appearing in the right hand side of (5.14), let $\eta = \eta_*|_{V_L \setminus B_0}$. Due to the burning process, the conditional distribution of η , given the event $\{\eta_*|_{B_0} = \zeta|_{B_0}, V(T_o(\eta_*)) = B_0\}$, equals that of a recurrent sandpile in the subgraph $G_L^{B_0}$ of G_L induced by the set of vertices $V(L) \cup \{s\} \setminus B_0$ (i.e. with closed boundary condition at B_0). Hence the last expression in (5.14) equals

$$g_L^{-1} \nu_{L,o}(\eta_* : \eta_*|_{B_0} = \zeta_{B_0}, V(T_o(\eta_*)) = B_0) \nu_{G_L^{B_0}}(\eta : \eta|_{A_0} = \zeta|_{A_0}). \tag{5.15}$$

Since the wired spanning forest in the subgraph of \mathbb{Z}^2 induced by $\mathbb{Z}^2 \setminus B_0$ is one-ended, we can apply [17, Theorem 3] to deduce that the last factor in (5.15) has a limit as $L \rightarrow \infty$. The first factor equals

$$\frac{1}{|\mathcal{T}_{B_0}|} g_L^{-1} \mu_{L,o}(V(\mathfrak{T}_{L,o}) = B_0), \tag{5.16}$$

where $|\mathcal{T}_{B_0}|$ is the number of spanning trees in the graph induced by B_0 . Due to Proposition 5.7, the product of the second and third factors in (5.16) approaches p_{A_0, B_0} , as $L \rightarrow \infty$. This implies the existence of the limit

$$\lim_{L \rightarrow \infty} \nu_L(N \geq k, (\xi_k^L)|_{B(\ell)} = \zeta) =: c_k(\zeta).$$

Summing over all ζ satisfying (a)–(c), and using the choice of ℓ made after (5.13), we get

$$\left| \limsup_{L \rightarrow \infty} \nu_L(N \geq k) - \liminf_{L \rightarrow \infty} \nu_L(N \geq k) \right| \leq \varepsilon.$$

But since the left hand side does not depend on ℓ , we have that the limit $\lim_{L \rightarrow \infty} \nu_L(N \geq k) =: b_k$ exists, proving statement (i). It follows that

$$\lim_{L \rightarrow \infty} \nu_L((\xi_k^L)_{B(\ell)} = \zeta | N \geq k) = \frac{c_k(\zeta)}{b_k} =: \rho_k(\xi_k : (\xi_k)|_{B(\ell)} = \zeta).$$

Statement (ii) follows immediately from this.

(iii) Observe that the proof of parts (i)–(ii) shows that up to a set of measure 0, the support of ρ_k can be partitioned into a countable disjoint union of cylinder sets, such that on each element of the partition, the stabilization $(a'_{o, \mathbb{Z}^2})^k(\xi_k)$ takes place within a finite set $B(\ell)$.

(iv) The countable partition into cylinder sets has the further property that the map $\xi_k \mapsto a'_{o, \mathbb{Z}^2}(\xi_k)$ is measure preserving on each cylinder set of the partition. Hence the claim follows.

(v) Let $\varepsilon > 0$ be fixed. Let $N_{B(\ell)}$ denote the number of times o topples if all topplings in $B(\ell)$ are carried out, but no site in $B(\ell)^c$ is allowed to topple. Due to the ν -a.s.

convergence $N_{B(\ell)} \uparrow N$, there exists $1 \leq \ell < \infty$ such that $\nu(\{N = k\} \Delta \{N_{B(\ell)} = k\}) < \varepsilon$, where Δ denotes symmetric difference. Let

$$F_{\ell,k} = \{N_{B(\ell)} = k \text{ and some boundary vertex of } B(\ell) \text{ topples}\}.$$

Since $\{N = k\} \cap F_{\ell,k}^c$ is a cylinder event, we have

$$\nu(N = k) \geq \nu(N = k, F_{\ell,k}^c) = \lim_{L \rightarrow \infty} \nu_L(N = k, F_{\ell,k}^c) \geq \lim_{L \rightarrow \infty} \nu_L(N = k) - \varepsilon(\ell, k),$$

where $\varepsilon(\ell, k) \rightarrow 0$ as $\ell \rightarrow \infty$, due to (5.13). It follows that $b_k - b_{k+1} = \lim_{L \rightarrow \infty} \nu_L(N = k) \leq \nu(N = k)$.

For an inequality in the other direction, we write:

$$\begin{aligned} \nu(N = k) &\leq \nu(N = k, F_{\ell,k}^c) + \nu(F_{\ell,k}) = \nu(N_{B(\ell)} = k, F_{\ell,k}^c) + \nu(F_{\ell,k}) \\ &= \lim_{L \rightarrow \infty} \nu_L(N_{B(\ell)} = k, F_{\ell,k}^c) + \lim_{L \rightarrow \infty} \nu_L(F_{\ell,k}) \\ &\leq \lim_{L \rightarrow \infty} \nu_L(N = k) + \limsup_{L \rightarrow \infty} \nu_L(N > k, F_{\ell,k}) + \lim_{L \rightarrow \infty} \nu_L(N = k, F_{\ell,k}). \end{aligned}$$

Due to (5.13), the third term on the right hand side is at most $\varepsilon(\ell, k) \rightarrow 0$ as $\ell \rightarrow \infty$. Therefore, it is enough to show that

$$\lim_{\ell \rightarrow \infty} \limsup_{L \rightarrow \infty} \nu_L(N > k, F_{\ell,k}) = 0. \tag{5.17}$$

We fix $0 < \delta < \alpha/\beta$ (where α, β are the constants from Theorem 5.9). Let $r(i) = L^{\delta/\rho^i}$, $i = 0, \dots, k$, where the constant $1 < \rho < \infty$ will be chosen later. Recall we denote by $\eta_1, \dots, \eta_k \in \mathcal{R}'_L \setminus \mathcal{R}_L$ the first k waves corresponding to $\eta \in \mathcal{R}_L$. We define the events

$$H(i) = \left\{ \begin{array}{l} \text{the } i\text{-th wave } \mathcal{W}(\eta_i) \text{ does not topple any vertices} \\ \text{in } B(r(i)) \text{ after it has reached } \partial B(r(i-1)) \end{array} \right\}, \quad i = 1, \dots, k.$$

Recalling property (5.1) of the bijection, an argument similar to the one made in Proposition 5.4 yields

$$\nu_L(H(i)^c) \leq C(\log L) r(i-1)^{-1/4} r(i)^{9/4} \leq C(\log L) L^{2\delta\rho^{-i}} L^{-\delta\rho^{-i}(\rho-1)/4}. \tag{5.18}$$

We choose $\rho > 9$, so that the right hand side of (5.18) goes to 0 as $L \rightarrow \infty$. Therefore, in order to prove (5.17), it is enough to bound

$$\nu_L(N > k, F_{\ell,k}, H(1) \cap \dots \cap H(k)) \leq \nu_L(N > k, N_{B(\ell)} = k, H(1) \cap \dots \cap H(k)). \tag{5.19}$$

Suppose now that we are given a configuration $\eta \in \mathcal{R}_L$. Let us carry out the first wave up to $\partial B(r(0))$, that is, stop the first wave when a vertex of $B(r(0))^c$ would need to be toppled, if any. Then carry out the second wave up to $\partial B(r(1))$, the third wave up to $\partial B(r(2))$, etc. Let F' denote the event that during the k -th ‘‘partial wave’’ defined this way, all neighbours of the origin topple. The event in the right hand side of (5.19) implies the event F' . This is because the event $\{N > k\}$, in the presence of $H(1) \cap \dots \cap H(k)$, implies that the origin can be toppled a $k + 1$ -st time after the first k partial waves.

Observe that F' is measurable with respect to the pile inside $B(r(0))$, and $r(0) = L^\delta$. Hence, using Theorem 5.9, and $\nu(\{N = k\} \Delta \{N_{B(\ell)} = k\}) \leq \varepsilon$, the right hand side of (5.19) is at most

$$\begin{aligned} \nu_L(F', N_{B(\ell)} = k) &\leq \nu(F', N_{B(\ell)} = k) + CL^{\delta\beta} L^{-\alpha} \leq \nu(F', N = k) + \varepsilon + CL^{-\alpha+\beta\delta} \\ &= \varepsilon + CL^{-\alpha+\beta\delta}, \end{aligned}$$

where the last equality follows from the fact that $F' \subset \{N > k\}$. Due to the choice of δ , the second term goes to 0, as $L \rightarrow \infty$. Since ε is arbitrary, we obtain statement (v) of the Theorem. \square

We can now complete the proof of Theorem 1.7.

Proof of Theorem 1.7. (i) This follows immediately from Theorem 5.1(ii)–(iii).

(ii) This follows from Theorem 5.1(i),(v).

(iii) This follows from Theorem 5.1(ii),(v), since on the event $\{N = k\}$ we can approximate by cylinder events on which no vertex topples outside a fixed ball. \square

The following lemma completes the proof of the $d = 2$ case of Proposition 3.11.

Lemma 5.10. *We have*

$$\nu(z \in Av) = \lim_{L \rightarrow \infty} \nu_L(z \in Av).$$

Proof. For a sufficiently large number $k = k(z)$, we have the deterministic implication

$$N > k \Rightarrow z \in Av,$$

both in \mathbb{Z}^2 , and in V_L for L sufficiently large. With such k , we have

$$\begin{aligned} \nu(z \in Av) &= \nu(N \geq k + 1) + \sum_{\ell=1}^k \nu(N = \ell, z \in Av) \\ &= \lim_{L \rightarrow \infty} \nu_L(N \geq k + 1) + \sum_{\ell=1}^k \lim_{L \rightarrow \infty} \nu_L(N = \ell, z \in Av) = \lim_{L \rightarrow \infty} \nu_L(z \in Av). \end{aligned}$$

In the second equality, we applied Theorem 5.1(v) to the first term. In the second term, a.s. finiteness of the last ℓ waves allows us to approximate $\{N = \ell, z \in Av\}$ by a cylinder event, and the equality follows. \square

Proof of Theorem 1.8. (i) The bound follows immediately from the Proposition 5.6.

(ii) Since on the event $\{N \leq k\}$ we have $R \leq \max\{R_1, \dots, R_k\}$, this also follows Proposition 5.6. \square

We now prove Theorem 1.9. The idea of the argument is that, if $f(x) := \mathbb{E}_\nu n(o, x)$ were finite, then by invariance of ν under a_o , f would have to be a bounded harmonic function, hence constant. This is in contradiction with the structure of the avalanche. We first give two short lemmas on which this argument will be based.

Lemma 5.11. *Assuming $\nu(N < \infty) = 1$, we have $\nu(R < \infty) = 1$.*

Proof. This follows easily from Theorem 1.8. \square

Let a_o denote the operation on stable sandpiles on \mathbb{Z}^2 which maps η to $(\eta + 1_o)^\circ$ if a finite stabilization is possible (i.e., if $S < \infty$). Then the preceding lemma implies if $\nu(N < \infty) = 1$, there exists a set Ω with $\nu(\Omega) = 1$ such that a_o is defined on Ω . Given such an Ω , the next lemma shows that, similarly to ν_L , the infinite-volume measure ν is invariant under a_o .

Lemma 5.12. *Assuming $\nu(N < \infty) = 1$, ν is invariant under a_o . That is, for any ν -integrable function f ,*

$$\int f(a_o \eta) \nu(d\eta) = \int f(\eta) \nu(d\eta).$$

Proof. The argument of [15, Prop. 3.14] carries over exactly. There a similar statement is proved in the case $d \geq 3$, but the argument requires only almost sure finiteness of avalanches. \square

Proof of Theorem 1.9. If $\nu(N = \infty) > 0$, there is nothing to prove. We thus assume that $\nu(N < \infty) = 1$. We first note that, under this assumption, the infinite-volume addition operators are well-defined, since $n(o, x) \leq n(o, o)$.

The invariance statement above makes possible a version of the argument underlying Dhar’s formula (Lemma 2.2). Assume for the sake of contradiction that $\mathbb{E}_\nu N < \infty$. In particular, $0 \leq \mathbb{E}n(o, x) \leq \mathbb{E}n(o, o) < \infty$ for all $x \in \mathbb{Z}^2$. Since $N < \infty$, Lemma 5.11 above shows we can (almost surely) write

$$(\eta + \mathbf{1}_o)^\circ = \eta + \mathbf{1}_o - \sum_{z \in \mathbb{Z}^2} n(o, z) \Delta(z, \cdot),$$

where the sum above has finitely many nonzero terms and Δ is the graph Laplacian on \mathbb{Z}^2 . In particular, taking expectations and using the invariance above (which implies $\mathbb{E}_\nu \eta(x) = \mathbb{E}_\nu (\eta + \mathbf{1}_o)^\circ(x)$ for all x) gives

$$\sum_{z \in \mathbb{Z}^2} \mathbb{E}_\nu n(o, z) \Delta(z, x) = \mathbf{1}_o(x) \quad \text{for all } x \in \mathbb{Z}^2. \tag{5.20}$$

In other words, (5.20) says that $f(x) = \mathbb{E}_\nu n(o, x)$ is harmonic away from o and has Laplacian 1 at o . Since f is bounded, recurrence of random walk implies that f is constant. Since $n(o, x) \leq n(o, o)$ for all x , we have $\nu(n(o, x) = n(o, o) \text{ for all } x \in \mathbb{Z}^2) = 1$. However, if all vertices topple, the avalanche is infinite, a contradiction. \square

6 High-dimensional radius bounds

In this section we prove the bounds for the radius of the avalanche for $d \geq 5$. We prove Theorem 1.3 in Section 6.1. We use the results of Section 6.1 in Sections 6.2 and 6.3 to prove the lower and upper bounds on the radius.

6.1 Radius bounds on transitive unimodular graphs

See [26, Chapter 8] for background on unimodularity and mass transport.

Proof of Theorem 1.3. (i) We denote $A_x(a, b) = D_x(b) \setminus D_x(a)$. Consider the following mass transport. When $\text{diam}(\text{past}_x) > r$, let x send unit mass distributed equally among all vertices $y \in \text{past}_x \cap A_x(r, 2r)$. Let us write $\text{sent}(x)$ and $\text{get}(x)$, respectively, for the amount sent and received by x , respectively. Then $\mathbb{E}[\text{sent}(o)] = \mu(\text{diam}(\text{past}_o) > r)$. On the other hand, using Jensen’s inequality, we have

$$\begin{aligned} \mathbb{E}[\text{get}(o)] &= \sum_{x \in A_o(r, 2r)} \mu(o \in \text{past}_x) \mathbb{E} \left[\frac{1}{|\text{past}_x \cap A_x(r, 2r)|} \mid o \in \text{past}_x \right] \\ &\geq \sum_{x \in A_o(r, 2r)} \frac{\mu(o \in \text{past}_x)^2}{\mathbb{E} \left[|\text{past}_x \cap A_x(r, 2r)| \mathbf{1}_{o \in \text{past}_x} \right]}. \end{aligned}$$

Since $\text{past}_x \cap A_x(r, 2r) \subset \mathfrak{T}_o \cap D_o(4r)$, the statement follows.

(ii) When $\text{diam}(\mathfrak{C}_x) > 4r$, let x send unit mass distributed equally among vertices $y \in \mathfrak{C}_x \cap A_x(r, 4r)$. Then $\mathbb{E}[\text{sent}(o)] = \mu(\text{diam}(\mathfrak{C}_o) > 4r)$. On the other hand,

$$\mathbb{E}[\text{get}(o)] = \sum_{x \in A_o(r, 4r)} \mu(o \in \mathfrak{C}_x) \mathbb{E} \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{C}_x; x) > 4r}}{|\mathfrak{C}_x \cap A_x(r, 4r)|} \mid o \in \mathfrak{C}_x \right].$$

and the statement follows.

(iii) Although \mathbf{WSF}_o is not automorphism invariant, essentially the same mass transport can be used as in (ii), due to the fact that the rooted random network (\mathfrak{T}_o, o) is unimodular, in the sense of [7, 1]. Let $T \subset V$ be a finite tree, and $x, y \in T$. Define the function

$$f(T, x, y) = \begin{cases} \mathbf{1}_{y \in T} |T \cap A_x(r, 4r)|^{-1} & \text{when } \text{diam}(T; x) > 4r; \\ 0 & \text{otherwise.} \end{cases}$$

Let x send the following mass to y :

$$F(x, y) = \sum_{T \ni x, y} \mathbf{WSF}_x(\mathfrak{T}_x = T) f(T, x, y).$$

Let $\gamma \in \Gamma$. It is clear that $f(\gamma T, \gamma x, \gamma y) = f(T, x, y)$, and shifting Wilson’s algorithm by γ shows that $\mathbf{WSF}_x(\mathfrak{T}_x = T) = \mathbf{WSF}_{\gamma x}(\mathfrak{T}_{\gamma x} = \gamma T)$. Therefore, F is invariant under the diagonal action of Γ , and hence

$$\begin{aligned} \mathbf{WSF}_o(\text{diam}(\mathfrak{T}_o) > 4r) &= \sum_{x \in V} F(o, x) = \sum_{x \in V} F(x, o) \\ &= \sum_{x \in A_o(r, 4r)} \sum_{\substack{T \ni x, o \\ \text{diam}(T; x) > 4r}} \frac{\mathbf{WSF}_x(\mathfrak{T}_x = T)}{|T \cap A_x(r, 4r)|}. \end{aligned} \tag{6.1}$$

This yields the statement. □

Remark 6.1. Part (iii) of Theorem 1.3 can also be deduced from part (ii) via the following limiting procedure, the details of which we omit. (This construction was our initial approach to the radius upper bound.) Let $\omega \subset V$ be independent site percolation on V with density $0 < p \ll 1$. Conditional on ω , let \mathbf{WSF}_ω denote the measure on spanning forests of (V, E) where each vertex in ω is “wired to infinity”, in analogy with \mathbf{WSF}_o . The measure obtained by averaging \mathbf{WSF}_ω over ω is automorphism invariant. When $x \in \omega$, let x send or not send mass according to the same rule as in part (ii), otherwise let x send no mass. Now let $p \downarrow 0$. Conditional on $o \in \omega$ we have $\mathbf{WSF}_\omega \Rightarrow \mathbf{WSF}_o$, and the mass transport in part (iii) can be recovered in this limit.

6.2 Radius lower bound when $d \geq 5$

We prove the lower bound using the result of the previous section.

Proof of Theorem 1.2(iv), lower bound. We begin with some terminology. As in [25], let the ‘past of a vertex x ’ in a spanning tree T of G_L , be the union of the connected components of $T \setminus \{x\}$, which do not contain s . We denote this object by $\text{past}_x(T)$. Using the characterization of last waves from Section 2.4 and the bijection for intermediate configurations from Lemma 2.12 we have the following observation:

$$|\{\eta \in \mathcal{R}_L : R(\eta) > r\}| \geq |\{(T_o, T_s) \in \mathcal{T}_{L,o} : R(T_o) > r, e \notin T_o \text{ for some } e \sim o\}|. \tag{6.2}$$

Now, consider the map $\Psi : \mathcal{T}_L \rightarrow \mathcal{T}_{L,o}$, which modifies a spanning tree T by deleting the unique edge $\{o, e\}$ such that e is in the future of o and $\{o, e\}$ belongs to T . The following properties of the map Ψ are immediate.

Lemma 6.2. *Let $U \subset \mathcal{T}_L$. Then $|\Psi(U)| \leq |U| \leq 2d|\Psi(U)|$.*

We use the lemma with $U = \{T \in \mathcal{T}_L : R(\text{past}_o(T)) > r\}$, for which we have

$$\Psi(U) \subset \{(T_o, T_s) \in \mathcal{T}_{L,o} : R(T_o) > r, e \notin T_o \text{ for some } e \sim o\}.$$

Therefore equation (6.2) gives that

$$\frac{|\{\eta \in \mathcal{R}_L : R(\eta) > r\}|}{|\mathcal{R}_L|} \geq \frac{|\Psi(U)|}{|\mathcal{R}_L|} \geq \frac{1}{2d} \frac{|U|}{|\mathcal{R}_L|} \geq \frac{1}{2d} \frac{|\{T \in \mathcal{T}_L : R(\text{past}_o(T)) > r\}|}{|\mathcal{T}_L|}.$$

This equation holds for any fixed r , and all L . Using the observation that for fixed r the events at both ends are cylinder events, and taking the limit as $L \nearrow \infty$, we have,

$$\nu(R > r) \geq \frac{1}{2d} \mathbf{WSF}(R(\text{past}_o(\mathfrak{T})) > r). \tag{6.3}$$

The proof is completed by applying Theorem 1.3(i). Using the fact that in dimensions $d \geq 5$ there is probability at least c that two independent simple random walks starting at x do not intersect, we deduce that $\mathbf{WSF}(x \in \text{past}_o) \geq c|x|^{2-d}$. On the other hand, Wilson’s algorithm gives

$$\begin{aligned} \mathbb{E}[|\mathfrak{T}_o \cap B_o(4r)| \mathbf{1}_{o \in \text{past}_x}] &\leq \sum_{y \in B_o(4r)} \sum_{v \in \mathbb{Z}^d} [G(o, v) G(y, v) G(v, x) + G(o, x) G(x, v) G(y, v)] \\ &\leq C r^{6-d}. \end{aligned}$$

Therefore, Theorem 1.3(i) implies $\mathbf{WSF}(R(\text{past}_o) > r) \geq cr^{-2}$. □

6.3 Radius upper bound when $d \geq 5$

In this section we prove the upper bound in Theorem 1.2(iv). Taking d to be ℓ_∞ distance, Theorem 1.3(iii) yields

$$\begin{aligned} &\mathbf{WSF}_o(\text{diam}(\mathfrak{T}_o) > 4r) \\ &= \sum_{x \in \mathbb{Z}^d : r < \|x\|_\infty \leq 4r} \mathbf{WSF}_x(o \in \mathfrak{T}_x) \mathbb{E}_{\mathbf{WSF}_x} \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{T}_x; x) > 4r}}{|\mathfrak{T}_x \cap V_x(4r) \setminus V_x(r)|} \mid x \in \mathfrak{T}_o \right] \\ &\leq C r^{2-d} \sum_{x \in \mathbb{Z}^d : r < \|x\|_\infty \leq 4r} \mathbb{E}_{\mathbf{WSF}_x} \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{T}_x; x) > 4r}}{|\mathfrak{B}_x|} \mid x \in \mathfrak{T}_o \right], \end{aligned} \tag{6.4}$$

where we wrote $\mathfrak{B}_x = \mathfrak{T}_x \cap V_x(4r) \setminus V_x(r)$, and used $\mathbf{WSF}_x(o \in \mathfrak{T}_x) \leq G(o, x) \leq C r^{2-d}$. We show that for $\delta > 0$ there exists $C_1 = C_1(\delta)$ such that the expectation in the right hand side is bounded above by $C_1 (\log r)^{3+\delta} r^{-4}$. This implies the required upper bound on the tail of the diameter. We are going to use the following theorem of [5] on the lower tail of the volume of \mathbf{WSF} components. Given $D \subset \mathbb{Z}^d$, write \mathbf{WSF}_{D^c} for the wired spanning forest measure on the contracted graph \mathbb{Z}^d/D^c .

Theorem 6.3. [5] *Let $x, y \in \mathbb{Z}^d$ be such that $\|y - x\|_\infty = 4r$. Let $D \subset \mathbb{Z}^d$ be such $y \in \partial D$, and $V_x(4r) \subset D$. Let $x \leftrightarrow y$ denote the event that in \mathbf{WSF}_{D^c} the path from x to D^c reaches D^c via an edge incident to y . There exist constants $C, c > 0$ independent of D and r , such that for all $\lambda > 0$ we have*

$$\mathbf{WSF}_{D^c} \left(|\mathfrak{T}_x \cap V_x(2r) \setminus V_x(r)| < \lambda r^4 \mid x \leftrightarrow y \right) \leq C \exp(-c\lambda^{-1/3}).$$

We use Theorem 6.3 to establish the following regularity estimate. Fix a positive constant $\delta > 0$. Let us call \mathfrak{T}_x *bad*, if $\text{diam}(\mathfrak{T}_x) > 4r$, but $|\mathfrak{T}_x \cap V_x(2r) \setminus V_x(r)| < (\log r)^{-3-\delta} r^4$. We show the following lemma.

Lemma 6.4. *Let $x \in V_o(4r) \setminus V_o(r)$. We have*

$$\mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad}) \leq C \exp(-c(\log r)^{1+\delta/3}).$$

Proof. If $\text{diam}(\mathfrak{T}_x) > 4r$, then there exists y with $\|y - x\|_\infty = 4r$ such that $y \in \mathfrak{T}_x$. Therefore,

$$\mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad}) \leq \sum_{y: \|y-x\|_\infty=4r} \mathbf{WSF}_x(y \in \mathfrak{T}_x) \mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad} \mid y \in \mathfrak{T}_x).$$

The term $\mathbf{WSF}_x(y \in \mathfrak{T}_x) \leq C r^{2-d}$, and there are $O(r^{d-1})$ terms. On the other hand, we have $\mathbf{WSF}_x(\mathfrak{T}_x = T \mid y \in \mathfrak{T}_x) = \mathbf{WSF}_y(\mathfrak{T}_x = T \mid x \in \mathfrak{T}_y)$. This follows from the fact that LERW from y to x has the same distribution as LERW from x to y , and hence we can use these walks in the first step of Wilson’s algorithm on the two sides. See [24, Corollary 11.2.2] for “reversibility” of LERW. Hence Theorem 6.3 with $D = \mathbb{Z}^d \setminus \{y\}$ implies that

$$\mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad} \mid y \in \mathfrak{T}_x) = \mathbf{WSF}_y(\mathfrak{T}_x \text{ is bad} \mid x \in \mathfrak{T}_y) \leq C \exp(-c(\log r)^{1+\delta/3}).$$

The statement of the lemma follows. □

Lemma 6.5. *We have*

$$\mathbb{E} \mathbf{WSF}_x \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{T}_x;x) > 4r}}{|\mathfrak{B}_x|} \mid o \in \mathfrak{T}_x \right] \leq C \frac{(\log r)^{3+\delta}}{r^4}.$$

Proof. When \mathfrak{T}_x is bad, we use that $|\mathfrak{B}_x| \geq 1$, and hence due to Lemma 6.4 we have

$$\begin{aligned} \mathbb{E} \mathbf{WSF}_x \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{T}_x;x) > 4r} \mathbf{1}_{\mathfrak{T}_x \text{ is bad}}}{|\mathfrak{B}_x|} \mid o \in \mathfrak{T}_x \right] &\leq \mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad}, o \in \mathfrak{T}_x) C r^{d-2} \\ &\leq \mathbf{WSF}_x(\mathfrak{T}_x \text{ is bad}) C r^{d-2} \\ &\leq C \exp(-c(\log r)^{1+\delta/3}) r^{d-2} = o(r^{-4}). \end{aligned}$$

Therefore it is enough to consider the contribution when \mathfrak{T}_x is not bad. When this occurs, we have $|\mathfrak{B}_x| \geq (\log r)^{-3-\delta} r^4$, and hence

$$\mathbb{E} \mathbf{WSF}_x \left[\frac{\mathbf{1}_{\text{diam}(\mathfrak{T}_x;x) > 4r} \mathbf{1}_{\mathfrak{T}_x \text{ is not bad}}}{|\mathfrak{B}_x|} \mid o \in \mathfrak{T}_x \right] \leq \frac{(\log r)^{3+\delta}}{r^4}.$$

This proves the claim. □

We can now complete the proof of the upper bound in Theorem 1.2(iv). Lemma 6.5 implies that the right hand side of (6.4) is at most

$$C r^d r^{2-d} (\log r)^{3+\delta} r^{-4} = C (\log r)^{3+\delta} r^{-2}.$$

This completes the proof.

7 Bounds on the size

We now prove Theorem 1.5. Sections 7.1, 7.2 and 7.3 consider low dimensions, and Sections 7.4, 7.5 and 7.6 the case $d \geq 5$.

7.1 Upper bounds on the size when $d = 3, 4$

Proof of Theorem 1.5(ii)–(iii), upper bounds. The claimed upper bounds on $\nu(|Av| \geq t)$ follow immediately from Theorem 1.2, via the trivial estimate $\nu(|Av| \geq t) \leq \nu(R > c(d) t^{1/d})$. In order to obtain an upper bound on the tail of S , fix some $k \geq 1$. Recalling that N denotes the number of waves, we have

$$\nu(S > t) = \nu(S > t, N > k) + \nu(S > t, N \leq k). \tag{7.1}$$

Recalling that $d \geq 3$, we can upper bound the first term in the expression above by

$$\nu(N > k) \leq \mathbb{E}_\nu N k^{-1} \leq C k^{-1}.$$

Next, writing S^j for the size of the j -th wave, if $S > t$ and $N \leq k$, then we have $S^j > t/k$ for some $1 \leq j \leq N$. Hence,

$$\begin{aligned} \nu(S > t) &\leq C k^{-1} + \nu(S^j > t/k, \text{ for some } j \leq N) \\ &\leq C k^{-1} + C \nu(R > c(d)(t/k)^{1/d}). \end{aligned}$$

Due to Theorem 1.2(ii)–(iii), we get the bounds:

$$\nu(S > t) \leq \begin{cases} C k^{-1} + C t^{-1/18} k^{1/18} & \text{when } d = 3; \\ C k^{-1} + C t^{-1/16} k^{1/16} & \text{when } d = 4. \end{cases}$$

Optimizing the choice of k yields:

$$\nu(S > t) \leq \begin{cases} C t^{-1/19} & \text{when } d = 3; \\ C t^{-1/17} & \text{when } d = 4. \end{cases}$$

□

7.2 Lower bounds on the size when $2 \leq d \leq 4$

We fix $z = r e_1$, where $r = t^{1/d}$, and write $u = \|z\|/10$. Recall that in the course of the proof of Lemma 3.3 in Section 3, we showed that for $L \geq 100 \|z\|$, and e a neighbour of the origin, and with $\pi = \mathcal{L}S_e[0, \sigma_L]$, we have

$$\mathbb{P} \left(\pi \cap V_z(u) = \emptyset, \xi_o^{S_z} < \xi_\pi^{S_z} \right) \geq \begin{cases} c \mathbb{P}(\Gamma_{z,L}) \log |z| & \text{when } d = 2; \\ (2d)^{-1} \mathbb{P}(\Gamma_{z,L}) & \text{when } d = 3, 4. \end{cases} \quad (7.2)$$

Let us write $F_{z,L}$ for the event in the left hand side of (7.2). Our goal will be to show that conditional on the event $F_{z,L}$ (when z is in the last wave), a large number of vertices in $V_z(u)$ are also in the last wave, with probability bounded away from 0. We will use a second moment argument to prove this, that is based on Proposition 7.1 below.

Let $u' = (1 - \varepsilon)u$, where we are going to choose $0 < \varepsilon < 1/4$ later. Consider the following partial cycle popping in G_L , defined in three stages; see [35] and [26] for background on cycle popping. In the first stage, reveal the LERW π started at e and ending at s . In the second stage, reveal a LERW started at z , ending on hitting $\pi \cup \{o\}$. In the third stage, pop all cycles that are entirely contained in $V_z(u)$ that can be popped. Condition on the event $F_{z,L}$, that is measurable with respect to the result of the first two stages. Let $\pi' = \mathcal{L}S_z[0, \xi_o]$, and let $\pi'(u')$ be the portion of π' from z to the first exit from $V_z(u')$. Let

$$I(x) = \{\text{stage three reveals a path from } x \text{ to } \pi' \cap V_z(u)\}, \quad x \in V_z(u/2).$$

For each $x \in V_z(u/2)$ for which $I(x)$ occurs, let $p(x) \in \pi' \cap V_z(u)$ be the point where the revealed path first meets $\pi' \cap V_z(u)$. For technical reasons (that are only required for our argument when $d = 4$), we also define:

$$J(x) = \left\{ \xi_{\pi'(u')}^{S_{p(x)}} < \sigma_{V_z(u)}^{S_{p(x)}} \right\}.$$

Let

$$Y = Y_{u,\varepsilon} = \sum_{x \in V_z(u/2)} \mathbf{1}_{I(x)} \mathbf{1}_{J(x)}.$$

The following proposition states bounds on the first and second moments of Y .

Proposition 7.1.

There exist $0 < \varepsilon < 1/4$, $c_1 > 0$ and C such that the following hold.

(i) When $d = 2$, we have

$$\mathbb{E}(Y_{u,\varepsilon} | F_{z,L}) \geq c_1 u^2 \quad \text{and} \quad \mathbb{E}(Y_{u,\varepsilon}^2 | F_{z,L}) \leq C u^4.$$

(ii) When $d = 3$, we have

$$\mathbb{E}(Y_{u,\varepsilon} | F_{z,L}) \geq c_1 u^3 \quad \text{and} \quad \mathbb{E}(Y_{u,\varepsilon}^2 | F_{z,L}) \leq C u^6.$$

(iii) When $d = 4$, we have

$$\mathbb{E}(Y_{u,\varepsilon} | F_{z,L}) \geq c_1 u^4 (\log u)^{-1} \quad \text{and} \quad \mathbb{E}(Y_{u,\varepsilon}^2 | F_{z,L}) \leq C u^8 (\log u)^{-2}.$$

Proof of Theorem 1.5(i)–(iii); lower bounds; assuming Proposition 7.1.

Since on the event $F_{z,L}$, we have $\pi' \subset \mathfrak{T}_{L,o}$, we have $|\mathfrak{T}_{L,o}| \geq Y$. Hence for any $t' > 0$ we have

$$\begin{aligned} \nu_L(|Av| \geq t') &\geq (2d)^{-1} \frac{\mu_{L,o}(z \in \mathfrak{T}_{L,o}, e \notin \mathfrak{T}_{L,o}, |\mathfrak{T}_{L,o}| \geq t')}{\mu_{L,o}(e \notin \mathfrak{T}_{L,o})} \\ &\geq (2d)^{-1} \frac{\mathbb{P}(F_{z,L})}{\mathbb{P}_e(\sigma_L < \xi_o)} \mathbb{P}(Y \geq t' | F_{z,L}). \end{aligned} \tag{7.3}$$

Let us set $t' = (1/2) c_1 u^2$ in $d = 2$; $t' = (1/2) c_1 u^3$ in $d = 3$, and $t' = (1/2) c_1 u^4 (\log u)^{-1}$ in $d = 4$. The Paley-Zygmund inequality implies that

$$\mathbb{P}(Y \geq t' | F_{z,L}) \geq \frac{1}{4} \frac{c_1^2}{C}. \tag{7.4}$$

Letting $L \rightarrow \infty$, we obtain the required lower bounds from (7.3), (7.4), (7.2), Lemma 3.9, and the dimension-dependent estimates in Sections 3.3.1–3.3.3. \square

Proof of Proposition 7.1(i),(ii).

The upper bounds are immediate from $Y_{u,\varepsilon} \leq |V_z(u/2)|$.

For the lower bounds, using the strong Markov property of S_x at time $\xi_{\pi'}$, when S_x is at the point $p(x)$, we have

$$\mathbb{P}(I(x) \cap J(x) | F_{z,L}) = \mathbb{P}_x(\xi_{\pi'} \leq \xi_{\pi'(u')} < \sigma_{V_z(u)}) = \mathbb{P}_x(\xi_{\pi'(u')} < \sigma_{V_z(u)}).$$

Let $\pi'' := \mathcal{L}S_z[0, \sigma_{V_z(u)}]$, and let $\pi''(u')$ be the portion of π'' up to its first exit from $V_z(u')$. Due to Lemma 3.4, there exists $c_2 = c_2(\varepsilon)$, such that the distribution of $\pi'(u')$ is bounded below by c_2 times the distribution of $\pi''(u')$. This implies

$$\mathbb{P}_x(\xi_{\pi'(u')} < \sigma_{V_z(u)}) \geq c_2 \mathbb{P}_x(\xi_{\pi''(u')} < \sigma_{V_z(u)}). \tag{7.5}$$

We lower bound the probability in the right hand side of (7.5) separately in $d = 2, 3$.

When $d = 2$, we have

$$\mathbb{P}_x(\xi_{\pi''(u')} < \sigma_{V_z(u)}) \geq \mathbb{P}_x(S_x \text{ completes a loop around } z \text{ before exiting } V_z(u')) \geq c$$

with some constant $c > 0$. This follows from the invariance principle; see for example [24, Exercise 3.4]. Summing over $x \in V_z(u/2)$ yields the required lower bound.

When $d = 3$, we have

$$\mathbb{P}_x(\xi_{\pi''(u')} < \sigma_{V_z(u)}) \geq \mathbb{P}_x(\xi_{\pi''} < \sigma_{V_z(u)}) - \mathbb{P}_x(\sigma_{V_z(u')} \leq \xi_{\pi'' \setminus \pi''(u')} < \sigma_{V_z(u)}). \tag{7.6}$$

A result of Lyons, Peres and Schramm [27, Lemma 1.2] states that for two independent copies of the same transient Markov chain, the probability for one path to intersect the loop-erasure of the other is at least a universal constant $c_3 > 0$ times the probability that the Markov chain paths themselves intersect. Applying this to the random walks $S_z[0, \sigma_{V_z(u)}]$ and $S_x[0, \sigma_{V_z(u)}]$, we have

$$\mathbb{P}_x (\xi_{\pi''} < \sigma_{V_z(u)}) \geq c_3 \mathbb{P} (S_x[0, \sigma_{V_z(u)}] \cap S_z[0, \sigma_{V_z(u)}] \neq \emptyset). \tag{7.7}$$

The right hand side in (7.7) is bounded below by a constant $c_3 > 0$, independent of u . This can be seen by arguments due to Lawler; by adapting the proof of [23, Theorem 3.3.2].

It remains to bound the negative term in (7.6). For this we write

$$\mathbb{P}_x (\sigma_{V_z(u')} \leq \xi_{\pi'' \setminus \pi''(u')} < \sigma_{V_z(u)}) \leq \mathbb{P} (S_x[\sigma_{V_z(u')}, \sigma_{V_z(u)}] \cap S_z[\sigma_{V_z(u')}, \sigma_{V_z(u)}] \neq \emptyset). \tag{7.8}$$

Consider independent Brownian motions, starting at x/u and z/u . Since the paths are continuous, and with probability 1 they exit the cube $(z/u) + [-1, 1]^3$ at different points, the invariance principle implies that the probability in the right hand side of (7.8) goes to 0 uniformly in $u \geq (1/\varepsilon)$, as $\varepsilon \rightarrow 0$. Therefore, we can fix $\varepsilon > 0$ such that the right hand side of (7.8) is at most $c_3/2$, uniformly in u . With such a choice of ε , the first moment is bounded below by $c u^3$. \square

7.3 Proof of the moment bounds in $d = 4$

7.3.1 Proof of the first moment bounds

We begin with the first moment lower bound – that is, the lower bound in Proposition 7.1(iii). The line of reasoning leading to (7.6) and (7.7) above holds also for $d = 4$, and we take these as our starting point. We next establish an appropriate analogue of the constant lower bound given above for (7.6). We restrict to $x \notin V_z(u/4)$ for simplicity.

Lemma 7.2. *There is a $c > 0$ such that, uniformly in z and $x \in V_z(u/2) \setminus V_z(u/4)$,*

$$\mathbb{P} (S_x[0, \sigma_{V_z(u)}] \cap S_z[0, \sigma_{V_z(u)}] \neq \emptyset) \geq c / \log u .$$

Proof. Fix such an x , and consider the number of intersections

$$J_x := \sum_{k=0}^{\sigma_{V_z(u)}} \sum_{\ell=0}^{\sigma_{V_z(u)}} \mathbf{1}_{S_x(k)=S_z(\ell)} . \tag{7.9}$$

Taking expectations and using Theorem 2.13 gives a $c > 0$ such that $\mathbb{E}J_x \geq c$.

On the other hand, $\mathbb{E}J_x^2$ is of order at most $\log u$. By a computation similar to [23, Theorem 3.3.2; lower bounds], we have

$$\mathbb{E}J_x^2 \leq 2 \sum_{y_1, y_2 \in V_z(u)} [G(x, y_1)G(z, y_1)G(y_1, y_2)^2 + G(x, y_1)G(z, y_2)G(y_1, y_2)^2] .$$

Each term above gives a contribution of order $\log u$; we discuss in detail only the first term. By summing first over y_2 and using Theorem 2.13(ii) (taking the $n \rightarrow \infty$ limit in this theorem), we get an upper bound of order

$$\log u \sum_{y_1 \in V_z(u)} G(x, y_1)G(z, y_1) .$$

For each y_1 , either $|x - y_1|$ or $|z - y_1|$ is at least $u/8$ since $|x - z|$ is at least $u/4$. Thus either $G(x, y_1)$ or $G(z, y_1)$ is at most Cu^{-2} . Summing the other factor over y_1 gives a factor of order u^2 , giving the required upper bound.

Using the second moment method and noting that S_x and S_z intersect if $J_x > 0$ completes the proof. \square

It remains to control the negative term of (7.6). This will be accomplished using the following lemma:

Lemma 7.3. *There exists a constant $C_3 > 0$ such that, uniformly in $0 < \varepsilon < 1/2$ and u large (how large depends on ε), and uniformly in $y \in V_z(u) \setminus V_z(u')$,*

$$\mathbb{P} (S_y[0, \sigma_{V_z(u)}] \cap S_z[0, \sigma_{V_z(u)}] \neq \emptyset) \leq C_3 \varepsilon / \log u .$$

Proof of Proposition 7.1(iii), lower bound; assuming Lemma 7.3. Note that

$$\begin{aligned} \mathbb{P}_x (\sigma_{V_z(u')} \leq \xi_{\pi''} < \sigma_{V_z(u)}) &\leq \mathbb{P} (S_x[\sigma_{V_z(u')}, \sigma_{V_z(u)}] \cap S_z[0, \sigma_{V_z(u)}] \neq \emptyset) \quad (7.10) \\ &\leq \sup_{y \in \partial V_z(u')} \mathbb{P} (S_y[0, \sigma_{V_z(u)}] \cap S_z[0, \sigma_{V_z(u)}] \neq \emptyset) . \end{aligned}$$

The above, combined with Lemma 7.3, allows the choice of an appropriately small ε to give a uniform lower bound of $cu^4 / \log u$ for the right-hand side of (7.6), completing the proof of the first moment of Prop. 7.1(iii). \square

We turn to the proof of Lemma 7.3, which is an adaptation of the proof of [23, Theorem 3.3.2; $d = 4$ upper bound]. Let $u'' = (1 + \varepsilon)u$. To avoid complications introduced by intersections near the boundary, consider the extended number of intersections

$$J'_x := \sum_{k=0}^{\sigma_{V_x(u'')}} \sum_{\ell=0}^{\sigma_{V_z(u'')}} \mathbf{1}_{S_x(k)=S_z(\ell)} , \quad x \in V_z(u) .$$

Lemma 7.4. *There is C such that, uniformly in $\varepsilon < 1/2$ and z , and in $x \in V_z(u) \setminus V_z(u')$, we have*

$$\mathbb{E}J'_x \leq C \varepsilon .$$

Proof. We have

$$\begin{aligned} \mathbb{E}J'_x &\leq \sum_{y \in V_z(u'')} G(z, y) G_{V_z(u'')}(x, y) \\ &= \sum_{y \in V_x(u/10)} G(z, y) G_{V_z(u'')}(x, y) + \sum_{y \in V_z(u'') \setminus V_x(u/10)} G(z, y) G_{V_z(u'')}(x, y) . \quad (7.11) \end{aligned}$$

Let $y \in V_z(u'')$, and write $\|x - y\| = r$. A gambler's ruin estimate yields

$$G_{V_z(u'')}(x, y) \leq \mathbb{P}_x (\sigma_{V_x(r/2)} < \sigma_{V_z(u'')}) \sup_{a \in V_x(r/2)} G(a, y) \leq C \min \left\{ \frac{\varepsilon u}{r}, 1 \right\} r^{-2} . \quad (7.12)$$

Consider the first term of (7.11). Using $\|z - y\| \geq u/2$ and (7.12), this term is at most

$$\frac{C}{u^2} \left[\sum_{r=1}^{\varepsilon u} \left(r^3 \cdot \frac{1}{r^2} \right) + \sum_{r=\varepsilon u}^{u/10} \frac{\varepsilon u}{r} \cdot r^3 \cdot \frac{1}{r^2} \right] \leq C \varepsilon ,$$

The second term, using (7.12) again, is bounded by $C (\varepsilon u / u^3) \sum_{y \in V_z(u'')} G(y, z) \leq C \varepsilon$. \square

Recall the definition of J_x from (7.9). We show that, conditional on $\{J_x > 0\}$, the expectation of J'_x is at least $c \log u$. This gives the desired upper bound for $\mathbb{P}(J_x > 0)$.

Lemma 7.5. *We can find $r > 0$ such that, uniformly in $0 < \varepsilon < 1/2$, u such that $\varepsilon u > u^{1/2}$, and $x \in \partial V_z(u')$ such that x is at least distance $u/10$ from all but one face of $V_z(u)$, we have*

$$\mathbb{E}[J'_x \mid J_x > 0] \geq r \log u . \quad (7.13)$$

Proof. We follow a similar argument to the proofs of [24, Proposition 10.1.1] and [23, Theorem 3.3.2; $d = 4$ upper bound]. On $\{J_x > 0\}$, there is a lexicographically first intersection in $V_z(u)$. Specifically, we can define $\ell_1 := \inf\{j : S_x(j) \in S_z[0, \sigma_{V_z(u)}] \cap V_z(u)\}$ and $\ell_2 := \inf\{j : S_z(j) = S_x(\ell_1)\}$ and note that each ℓ_i is smaller than $\sigma_{V_z(u)}$. Using the strong Markov property of S_x at time ℓ_1 , conditionally on $S_z[0, \sigma_{V_z(u)}]$ and $S_x[0, \ell_1]$ the expected value of J'_x is bounded below by

$$\begin{aligned} \mathbb{E} [J'_x \mid S_z[0, \sigma_{V_z(u)}], S_x[0, \ell_1]] &\geq \sum_{i=\ell_2}^{\sigma_{V_z(u'')}} G_{V_z(u'')}(S_z(\ell_2), S_z(i)) \\ &\geq c \sum_{i=0}^{\sigma_{V_z(\sqrt{u})}} G(z, S_z(i)) \stackrel{d}{=} c \sum_{i=0}^{\sigma_{V(\sqrt{u})}} G(o, S_o(i)). \end{aligned}$$

Here we have used the fact that $G_{V_z(u'')}(a, S_z(\ell_2)) \geq cG(a, S_z(\ell_2))$ for $a \in V_{S_z(\ell_2)}(u^{1/2})$ along with translation invariance, and $\stackrel{d}{=}$ denotes equality in distribution.

The conclusion of Lemma 7.5 follows immediately from the above, using the following proposition (along with an a priori power law lower bound for $\mathbb{P}(J_x > 0)$): \square

Proposition 7.6 ([24, Lemma 10.1.2]). *For every $\alpha > 0$, there exist c, r such that for all n sufficiently large,*

$$\mathbb{P} \left(\sum_{j=0}^{\sigma_n-1} G(o, S_o(j)) \leq r \log n \right) \leq cn^{-\alpha}.$$

Proof of Lemma 7.3. Comparing Lemma 7.5, and Lemma 7.4, the claim nearly follows, except that $y \in \partial V_z(u')$ in (7.10) may be closer than distance $u/10$ to more than one face of $V_z(u)$. However, for such y we can replace $V_z(u)$ by a larger box $V' \supset V_z(u)$, whose diameter is still of order u , in such a way that y is at distance εu from the boundary of V' , and y is bounded away from the corners of V' . Since in V' the intersection probability is larger than in $V_z(u)$, the claim follows. \square

The above completes the proof of the lower bound in Proposition 7.1(iii).

7.3.2 Proof of the second moment bounds

Here we prove the second moment bound in Proposition 7.1 (iii). We first introduce some notation. We will work with dyadic cubes $\prod_{i=1}^4 [a_i 2^k, (a_i + 1) 2^k] \cap \mathbb{Z}^4$, where $a_1, a_2, a_3, a_4 \in \mathbb{Z}$, and $k \geq 1$. We say that such a cube is of scale k . If Q is dyadic cube, we denote by Q' , respectively, Q'' , the cubes that are concentric with Q and have twice, respectively, four times, the side-length. Given $v \in \mathbb{Z}^4$, we denote by $Q(v; k)$ the unique dyadic cube of scale k containing v .

The following inequality recasts the statement of Lemma 7.3. Let Q be a dyadic cube of scale k , $p \in Q$, $q \in Q''$, such that $\text{dist}_\infty(q, \partial Q'') = \varepsilon 2^k$. There is a constant C such that

$$\mathbb{P}(S_q[0, \sigma_{Q''}] \cap S_p[0, \infty) \neq \emptyset) \leq \frac{C \varepsilon}{k}. \tag{7.14}$$

Fix $x, y \in B_z(u/2)$ and assume that $I(x) \cap J(x) \cap I(y) \cap J(y)$ occurs. We distinguish the following two cases:

- (I) the paths from x and y to π' do not meet;
- (II) the paths from x and y to π' meet at a vertex $q(x, y) \notin \pi'$.

In Case (I), using that the events $J(x)$ and $J(y)$ occur, let $v, w \in \pi'(u')$, respectively, be the points where $S_{p(x)}$ and $S_{p(y)}$, respectively, first hit $\pi'(u')$. (Note that v and w may coincide.) It follows that

$$\mathbb{P}(\text{Case (I)}) \leq \mathbb{P}\left(\exists v, w \in \pi'(u') : \xi_v^{S_x} < \sigma_{V_z(u)}^{S_x}, \xi_w^{S_y} < \sigma_{V_z(u)}^{S_y}\right).$$

Due to Lemma 3.4, there exists $C_1 = C_1(\varepsilon)$, such that the distribution of $\pi'(u')$ is bounded above by C_1 times the distribution of $\pi''(u')$. Therefore, we have

$$\mathbb{P}(\text{Case (I)}) \leq C_1 \mathbb{P}\left(\exists v, w \in \pi''(u') : \xi_v^{S_x} < \sigma_{V_z(u)}^{S_x}, \xi_w^{S_y} < \sigma_{V_z(u)}^{S_y}\right).$$

In Case (II), we let $v = q(x, y)$, and noting $p(x) = p(y)$, let w be the point where $S_{p(x)}$ first hits $\pi'(u')$. Again bounding above by $\pi''(u')$, it follows that

$$\mathbb{P}(\text{Case (II)}) \leq C_1 \mathbb{P}\left(\exists w \in \pi''(u'), v \in V_z(u) : \xi_v^{S_x} < \xi_w^{S_x} < \sigma_{V_z(u)}^{S_x}, \xi_v^{S_y} < \sigma_{V_z(u)}^{S_y}\right).$$

We bound the probabilities of Cases (I) and (II) separately. The idea of the bound is *not* to sum over v and w , but rather, sum over the choice of suitable dyadic cubes that v and w fall into, and use the bound (7.14) for the probability of random walk intersections. Throughout, we write K for the integer such that $2^{K-1} < u \leq 2^K$.

Case (I). We may assume without loss of generality that the walk S_z generating π'' hits v before w , as the other case follows by a symmetric argument. For convenience, we assume that $\|z - v\|, \|v - w\|, \|x - v\|, \|y - w\|$ are all at least 32. At the end of the proof we comment on how to handle the remaining configurations of points. We define the following dyadic scales and cubes:

$$\begin{aligned} k_v &:= \max \{k \geq 1 : 2^{k+4} \leq \min\{\|v - z\|_\infty, \|v - w\|_\infty, \|v - x\|_\infty\}\} \\ k_w &:= \max \{k \geq 1 : 2^{k+4} \leq \min\{\|w - v\|_\infty, \|w - y\|_\infty\}\} \\ k_{vw} &:= \max \{k \geq 1 : 2^{k+4} \leq \|v - w\|_\infty\} \\ Q(v) &:= Q(v; k_v) \quad Q(w) = Q(w; k_w). \end{aligned}$$

We also let

$$\begin{aligned} k_z &:= \max \{k \geq 1 : 2^{k+4} \leq \|z - v\|_\infty\} & k_x &:= \max \{k \geq 1 : 2^{k+4} \leq \|x - v\|_\infty\} \\ k_y &:= \max \{k \geq 1 : 2^{k+4} \leq \|y - w\|_\infty\}. \end{aligned} \tag{7.15}$$

A sketch of the argument is as follows: the walks S_z and S_x both have to hit $Q'(v)$, and then they intersect at a point of $Q(v)$. Following the intersection, the walk S_z has to hit $Q'(w)$, and so does the walk S_y . These two walks then intersect at a point of $Q(w)$. Breaking up the paths into pieces, the various hitting and intersection events will give us the estimate:

$$C \frac{(2^{k_v})^2}{(2^{k_z})^2} \frac{(2^{k_v})^2}{(2^{k_x})^2} \frac{1}{\log 2^{k_v}} \frac{(2^{k_w})^2}{\|w - v\|_\infty^2} \frac{(2^{k_w})^2}{(2^{k_y})^2} \frac{1}{\log 2^{k_w}}. \tag{7.16}$$

We need to sum this estimate over the choices of x and y , and the choices of the boxes $Q(v)$ and $Q(w)$. In the summation we will need to distinguish a number of sub-cases according to the relative sizes of the scales k_v, k_w, k_z, k_x, k_y .

We first establish the bound in (7.16). This is provided by the following lemma.

Lemma 7.7. (Probability bound for Case (I)) *Let R_1 and R_2 be dyadic boxes of scales k_1 and k_2 , and let $x, y \in B_z(u/2)$ be points such that:*

- (i) R_1'' and R_2'' are disjoint;

(ii) $\text{dist}(z, R_1''), \text{dist}(x, R_1'') \geq 2^{k_1}$;

(iii) $\text{dist}(y, R_2'') \geq 2^{k_2}$.

Define k'_x, k'_y, k'_z by the formulas (7.15) where $Q(v)$ and $Q(w)$ are replaced by R_1 and R_2 , respectively. Then

$$\begin{aligned} & \mathbb{P}[\exists v \in R_1, \exists w \in R_2 \text{ s.t. Case (I)}] \\ & \leq C \frac{(2^{k_1})^2}{(2^{k'_z})^2} \frac{(2^{k_1})^2}{(2^{k'_x})^2} \frac{1}{k_1} \frac{(2^{k_2})^2}{\text{dist}_\infty(R_1, R_2)^2} \frac{(2^{k_2})^2}{(2^{k'_y})^2} \frac{1}{k_2}. \end{aligned} \tag{7.17}$$

Proof. We need to be careful about the event when the walk S_z first hits R_1' , leaves R_1' and returns, before intersecting the path of S_x . The following definitions take care of this possibility by introducing the variables ℓ_1 and ℓ_2 that count crossings from $\partial R_1''$ to $\partial R_1'$ and from $\partial R_2''$ to $\partial R_2'$, respectively. The definitions are somewhat tedious to write down; however, estimating the resulting probabilities is then straightforward using the strong Markov property. Given $\ell_1, \ell_2 \geq 0$, let

$$\begin{aligned} T_{\ell_1} &= \inf\{n \geq \xi_{R_1'}^{S_z} : S_z[\xi_{R_1'}, n] \text{ has made at least } \ell_1 \text{ crossings from } \partial R_1'' \text{ to } R_1'\} \\ \sigma_{\ell_1, R_1'} &= \inf\{n \geq T_{\ell_1} : S_z(n) \notin R_1'\} \\ \xi_{\ell_1, R_2'} &= \inf\{n \geq \sigma_{\ell_1, R_1'} : S_z \in R_2'\} \\ T_{\ell_1, \ell_2} &= \inf\{n \geq \xi_{\ell_1, R_2'} : S_z[\xi_{\ell_1, R_2'}, n] \text{ has made at least } \ell_2 \text{ crossings from } \partial R_2'' \text{ to } R_2'\} \\ \sigma_{\ell_1, \ell_2, R_2'} &= \inf\{n \geq T_{\ell_1, \ell_2} : S_z(n) \notin R_2'\}. \end{aligned}$$

On the event in the left hand side of (7.17), the following events occur for some integers $\ell_1, \ell_2 \geq 0$:

- | | |
|--|--|
| (i) $\xi_{R_1'}^{S_z} < \infty$ | (v) $\xi_{\ell_1, R_2'}^{S_z} < \infty$ |
| (ii) $\xi_{R_1}^{S_x} < \infty$ | (vi) $\xi_{R_2}^{S_y} < \infty$ |
| (iii) $T_{\ell_1} < \infty$ | (vii) $T_{\ell_1, \ell_2} < \infty$ |
| (iv) $S_z[T_{\ell_1}, \sigma_{\ell_1, R_1'}] \cap S_x[\xi_{R_1}, \infty) \neq \emptyset$ | (viii) $S_z[T_{\ell_1, \ell_2}, \sigma_{\ell_1, \ell_2, R_2'}] \cap S_y[\xi_{R_2}, \infty) \neq \emptyset$ |

We bound the probability that (i)–(viii) occur, with each estimate conditional on the previous ones. The probability of (i)–(ii) is bounded by $C(2^{k_1}/2^{k'_z})^2(2^{k_1}/2^{k'_x})^2$, since $d = 4$. Using the strong Markov property of S_z at times $\xi_{R_1'}, T_1, \dots, T_{\ell_1-1}$, we have that (iii) occurs with conditional probability $\leq c_1^{\ell_1}$ with some $0 < c_1 < 1$. Since $S_z(T_{\ell_1}) \in \partial R_1'$ and $S_x(\xi_{R_1}) \in \partial R_1'$ are at distance of order 2^{k_1} from each other, the conditional probability of (iv) is bounded by $C/(\log 2^{k_1}) = C'/k_1$. The probability of (v)–(vi) is bounded by $C(2^{k_2}/\text{dist}_\infty(R_1, R_2))^2(2^{k_2}/2^{k'_y})^2$. The probability of (vii) is bounded by $c_1^{\ell_2}$. Finally, the probability of (viii) is bounded by C/k_2 , again due to (7.14). Multiplying the bounds and summing over $1 \leq \ell_1, \ell_2 < \infty$ yields the lemma. \square

We continue with the bound for Case (I). We break up Case (I) into the following sub-cases (that partially overlap, but together cover all possibilities):

- | | |
|--|--|
| (I-1) $k_v < k_z, k_x$ and $k_w < k_y$; | (I-4) $k_v = k_z \leq k_x$ and $k_w = k_y$; |
| (I-2) $k_v < k_z, k_x$ and $k_w = k_y$; | (I-5) $k_v = k_x \leq k_z$ and $k_w < k_y$; |
| (I-3) $k_v = k_z \leq k_x$ and $k_w < k_y$; | (I-6) $k_v = k_x \leq k_z$ and $k_w = k_y$. |

Fixing the scales k_v, k_w, k_z, k_x, k_y , we bound the number of choices of x, y and the dyadic boxes containing v and w in each case separately, and apply Lemma 7.7. Then we sum over the scales allowed in each sub-case. A depiction of case (I-1) may be found in Figure 2.

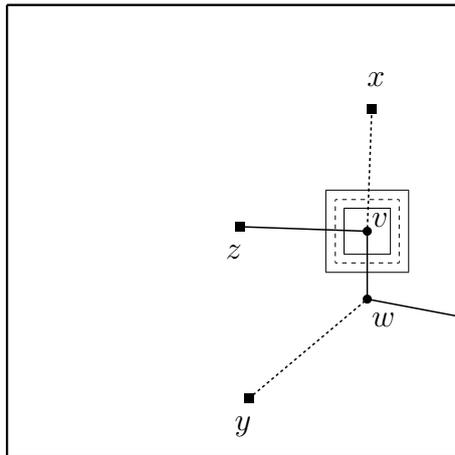


Figure 2: Depiction of sub-case (I-1). The walk S_z is depicted as a solid line; walks from x and y are dashed. S_x and S_z intersect at v , which is surrounded by boxes $Q(v)$, $Q'(v)$, and $Q''(v)$. Walks S_z and S_y intersect at w (boxes not shown). Note that in this case, the factor limiting the size of $Q(v)$ is the proximity of the vertex w . After intersection, S_z terminates upon intersecting the sink s (i.e., the boundary of the large square).

Sub-case (I-1). The number of choices for $Q(v)$ is of order $2^{4k_z}/2^{4k_v}$. The number of choices for x is of order 2^{4k_x} . Given $Q(v)$, the number of choices for $Q(w)$ is $O(1)$ (note that $k_w = k_v$), and the number of choices for y is of order 2^{4k_y} . Multiplying these bound together, and applying Lemma 7.7 to Sub-case (I-1), we get the estimate:

$$\frac{2^{4k_z}}{2^{4k_v}} 2^{4k_x} 2^{4k_y} \frac{2^{2k_v}}{2^{2k_z}} \frac{2^{2k_v}}{2^{2k_x}} \frac{1}{k_v} \frac{2^{2k_v}}{2^{2k_v}} \frac{2^{2k_v}}{2^{2k_y}} \frac{1}{k_v} = 2^{2k_z} 2^{2k_x} 2^{2k_y} \frac{2^{2k_v}}{k_v^2}.$$

Summing this bound for fixed k_v over k_x, k_y, k_z such that $k_v < k_x, k_y, k_z \leq K$, and then over $1 \leq k_v \leq K$, we get

$$\sum_{k_v=1}^K (2^{2K})^3 \frac{2^{2k_v}}{k_v^2} \leq C \frac{(2^K)^8}{K^2} = C \frac{u^8}{(\log u)^2}.$$

Sub-case (I-2). The number of choices for $Q(v)$ is of order $2^{4k_z}/2^{4k_v}$. The number of choices for x is of order 2^{4k_x} . Given $Q(v)$, the number of choices for $Q(w)$ is of order $2^{4k_v}/2^{4k_w}$, and the number of choices for y is of order 2^{4k_w} . Lemma 7.7 now gives:

$$\frac{2^{4k_z}}{2^{4k_v}} 2^{4k_x} \frac{2^{4k_v}}{2^{4k_w}} 2^{4k_w} \frac{2^{2k_v}}{2^{2k_z}} \frac{2^{2k_v}}{2^{2k_x}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_v}} \frac{1}{k_w} = 2^{2k_z} 2^{2k_x} \frac{2^{2k_v}}{k_v} \frac{2^{2k_w}}{k_w}.$$

We sum over $k_v < k_x, k_z \leq K$ for fixed $1 \leq k_w \leq k_v$, then over $1 \leq k_w \leq k_v \leq K$. This yields $Cu^8/(\log u)^2$ and completes the estimate in Sub-case (I-2).

The other four sub-cases are handled similarly, and we only state the number of choices, the probability estimate, and the range of summations over the scales.

Inequalities for critical exponents in sandpiles

	Choices	Probability	Summed over
I-3	$2^{4k_x} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_x}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_y}} \frac{1}{k_w}$	k_x such that $k_x \geq k_v$; k_y such that $k_y > k_w$; and k_v, k_w such that $1 \leq k_v \leq k_w \leq K$
I-4	$2^{4k_x} \frac{2^{4k_{vw}}}{2^{4k_v}} 2^{4k_w}$	$\frac{2^{2k_v}}{2^{2k_x}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_{vw}}} \frac{1}{k_w}$	k_x such that $k_x \geq k_v$; k_{vw} such that $k_{vw} \geq k_w$; and k_v, k_w such that $1 \leq k_v, k_w \leq K$
I-5	$\frac{2^{4k_z}}{2^{4k_v}} 2^{4k_v} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_z}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_y}} \frac{1}{k_w}$	k_z such that $k_z \geq k_v$; k_y such that $k_y > k_w$; and k_v, k_w such that $1 \leq k_v \leq k_w \leq K$
I-6	$\frac{2^{4k_z}}{2^{4k_v}} 2^{4k_v} \frac{2^{4k_{vw}}}{2^{4k_w}} 2^{4k_w}$	$\frac{2^{2k_v}}{2^{2k_z}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_{vw}}} \frac{1}{k_w}$	k_z such that $k_z \geq k_v$; k_{vw} such that $k_{vw} \geq k_w$; and k_v, k_w such that $1 \leq k_v, k_w \leq K$

Case (II). We will use notation similar to Case (I), but with somewhat different meaning. Let

$$\begin{aligned}
 k_v &:= \max \{k \geq 1 : 2^{k+4} \leq \min\{\|v - x\|_\infty, \|v - y\|_\infty, \|v - w\|_\infty\}\} \\
 k_w &:= \max \{k \geq 1 : 2^{k+4} \leq \min\{\|w - v\|_\infty, \|w - z\|_\infty\}\} \\
 k_{vw} &:= \max \{k \geq 1 : 2^{k+4} \leq \|v - w\|_\infty\} \\
 Q(v) &:= Q(v; k_v) \quad Q(w) = Q(w; k_w).
 \end{aligned}$$

We also let

$$\begin{aligned}
 k_z &:= \max \{k \geq 1 : 2^{k+4} \leq \|z - w\|_\infty\} \\
 k_x &:= \max \{k \geq 1 : 2^{k+4} \leq \|x - v\|_\infty\} \\
 k_y &:= \max \{k \geq 1 : 2^{k+4} \leq \|y - v\|_\infty\}.
 \end{aligned} \tag{7.18}$$

The following lemma provides the probability bound in Case (II), and is proved similarly to Lemma 7.7.

Lemma 7.8. (Probability bound for Case (II)) Let R_1 and R_2 be dyadic boxes of scales k_1 and k_2 , and let $x, y \in B_z(u/2)$ be points such that:

- (i) R_1'' and R_2'' are disjoint;
- (ii) $\text{dist}(x, R_1''), \text{dist}(y, R_1'') \geq 2^{k_1}$;
- (iii) $\text{dist}(z, R_2'') \geq 2^{k_2}$.

Define k'_x, k'_y, k'_z by the formulas (7.18) where $Q(v)$ and $Q(w)$ are replaced by R_1 and R_2 , respectively. Then

$$\mathbb{P}[\exists v \in R_1, \exists w \in R_2 \text{ s.t. Case (II)}] \leq C \frac{(2^{k_1})^2}{(2^{k'_x})^2} \frac{(2^{k_1})^2}{(2^{k'_y})^2} \frac{1}{k_1} \frac{(2^{k_2})^2}{\text{dist}_\infty(R_1, R_2)^2} \frac{(2^{k_2})^2}{(2^{k'_z})^2} \frac{1}{k_2}. \tag{7.19}$$

We now list the sub-cases to be considered. These are:

- (II-1) $k_v < k_x, k_y$ and $k_w < k_z$;
- (II-2) $k_v < k_x, k_y$ and $k_w = k_z$;
- (II-3) $k_v = k_x \leq k_y$ and $k_w < k_z$;
- (II-4) $k_v = k_x \leq k_y$ and $k_w = k_z$.

Interchanging the roles of x and y in (II-3) and (II-4) yields the remaining configurations not covered.

	Choices	Probability	Summed over
II-1	$\frac{2^{4k_z}}{2^{4k_v}} 2^{4k_x} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_x}} \frac{2^{2k_v}}{2^{2k_y}} \frac{1}{k_v} \frac{2^{2k_v}}{2^{2k_z}} \frac{1}{k_v}$	$k_x, k_y, k_z > k_v$ for fixed k_v ; and then over $1 \leq k_v \leq K$
II-2	$2^{4k_x} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_x}} \frac{2^{2k_v}}{2^{2k_y}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_v}} \frac{1}{k_w}$	$k_x, k_y > k_v$ for fixed k_v, k_w ; and then over $1 \leq k_w \leq k_v \leq K$
II-3	$\frac{2^{4k_z}}{2^{4k_w}} \frac{2^{4k_w}}{2^{4k_v}} 2^{4k_x} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_y}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_z}} \frac{1}{k_w}$	$k_y > k_v$ and $k_z > k_w$ for fixed k_v, k_w ; and then over $1 \leq k_v \leq k_w \leq K$
II-4	$\frac{2^{4k_{vw}}}{2^{4k_v}} 2^{4k_x} 2^{4k_y}$	$\frac{2^{2k_v}}{2^{2k_y}} \frac{1}{k_v} \frac{2^{2k_w}}{2^{2k_{vw}}} \frac{1}{k_w}$	$k_y > k_v$ and $k_{vw} > k_w$ for fixed k_v, k_w ; and then over $1 \leq k_v, k_w \leq K$

This completes the analysis of Case (II).

It remains to comment on configurations where one of the ℓ_∞ distances is < 32 . In these cases, we can replace the box $Q(v)$ and/or $Q(w)$ by the point v and/or w itself, and omit the random variables T_{ℓ_1} , etc. The combinatorial bounds, as well as the probability bounds still hold with $k_v = 1$ and/or $k_w = 1$, and this completes the proof of the upper bound in Proposition 7.1 (iii).

7.4 The size of the past in invariant forests

Proof of Theorem 1.6. Recall that $A_x(a, b) = D_x(b) \setminus D_x(a)$. When $|\text{past}_x \cap D_x(2r)| > t$, let x send mass 1 distributed equally among the vertices $z \in \text{past}_x \cap A_x(r, (3/2)r)$. Then $\mathbb{E}[\text{sent}(o)] \leq \mu(|\text{past}_o| > t)$. On the other hand, using Jensen’s inequality, we have

$$\begin{aligned}
 \mathbb{E}[\text{get}(o)] &\geq \sum_{x \in A_o(r, (3/2)r)} \mathbb{E} \left[\frac{\mathbf{1}_{\{o \in \text{past}_x\}} \mathbf{1}_{|\text{past}_x \cap D_x(2r)| > t}}{|\text{past}_x \cap A_x(r, (3/2)r)|} \right] \\
 &\geq \sum_{x \in A_o(r, (3/2)r)} \mathbb{E} \left[\frac{\mathbf{1}_{\{o \in \text{past}_x\}} \mathbf{1}_{|\tilde{\mathfrak{F}}_o(r/2)| > t}}{|\mathfrak{F}_o(4r)|} \right] \\
 &\geq \sum_{x \in A_o(r, (3/2)r)} \frac{\mu(o \in \text{past}_x, |\tilde{\mathfrak{F}}_o(r/2)| > t)}{\mathbb{E} \left[|\mathfrak{F}_o(4r)| \mid o \in \text{past}_x, |\tilde{\mathfrak{F}}_o(r/2)| > t \right]} \\
 &\geq \sum_{x \in A_o(r, (3/2)r)} \frac{\mu(o \in \text{past}_x, |\tilde{\mathfrak{F}}_o(r/2)| > t)^2}{\mathbb{E} \left[|\mathfrak{F}_o(4r)| \mathbf{1}_{\{o \in \text{past}_x\}} \right]}.
 \end{aligned} \tag{7.20}$$

This completes the proof. □

7.5 Lower bounds: $d \geq 5$

We now apply Theorem 1.6 to the measure **WSF**.

Proof of Theorem 1.5(iv); lower bound. Let $t = \delta r^4$, where $\delta > 0$ will be chosen in course of the proof. Wilson’s algorithm yields:

$$\begin{aligned}
 \mathbb{E} \left[|\mathfrak{F}_o(4r)| \mathbf{1}_{\{o \in \text{past}_x\}} \right] &\leq \sum_{y \in B_o(4r)} \sum_{u \in \mathbb{Z}^d} \left[\text{WSF}(o \in \text{past}_u, u \in \text{past}_x, y \in \text{past}_u) \right. \\
 &\quad \left. + \text{WSF}(o \in \text{past}_x, x \in \text{past}_u, y \in \text{past}_u) \right]
 \end{aligned}$$

$$\begin{aligned} &\leq \sum_{y \in B_o(4r)} \sum_{u \in \mathbb{Z}^d} [G(o, u) G(u, x) G(y, u) + G(o, x) G(x, u) G(y, u)] \\ &\leq Cr^{6-d}. \end{aligned} \tag{7.21}$$

Write $\text{future}_x = \{y \in \mathbb{Z}^d : x \in \text{past}_y\}$. Using Cauchy-Schwarz we have

$$\begin{aligned} &\sum_{x \in A_o(r, (3/2)r)} \mathbf{WSF}(o \in \text{past}_x, |\tilde{\mathfrak{X}}_o(r/2)| > t)^2 \\ &\geq cr^{-d} \left(\sum_{x \in A_o(r, (3/2)r)} \mathbf{WSF}(o \in \text{past}_x, |\tilde{\mathfrak{X}}_o(r/2)| > t) \right)^2 \\ &= cr^{-d} \mathbb{E} \left[|\text{future}_o \cap A_o(r, (3/2)r)| \mathbf{1}_{|\tilde{\mathfrak{X}}_o(r/2)| > t} \right]^2. \end{aligned} \tag{7.22}$$

We have $\mathbb{E}[|\tilde{\mathfrak{X}}_o(r/2)|] \geq cr^4$, due to a result of Pemantle [31, Lemma 3.1], and using Wilson’s algorithm, we have

$$\mathbb{E}[|\tilde{\mathfrak{X}}_o(r/2)|^2] \leq \sum_{u, y, w \in B_o(r/2)} G(o, w) G(w, u) G(w, y) \leq Cr^8.$$

This yields $\mathbf{WSF}(|\tilde{\mathfrak{X}}_o(r/2)| > t) \geq c_1 > 0$ for some $c_1 > 0$ and sufficiently small δ . Fix such δ . We get a lower bound on $|\text{future}_o \cap A_o(r, (3/2)r)|$ by considering the number of loop-free points of the random walk S_o generating the path from o to ∞ . A result of Lawler says that with probability 1, the fraction of loop-free points is asymptotically a positive constant in $d \geq 5$; see [23, Section 7.7]. Therefore, we can find $\varepsilon > 0$ small enough, such that $\mathbf{WSF}(|\text{future}_o \cap A_o(r, (3/2)r)| \geq \varepsilon r^2) \geq 1 - c_1/2$. With these choices of δ and ε , the right hand side of (7.22) is at least $r^{-d} ((c_1/2) \varepsilon r^2)^2 = c_2 r^{4-d}$. Substituting this and (7.21) into the bound given by Theorem 1.6 we obtain $\mathbf{WSF}(|\text{past}_o| \geq t) \geq cr^{-2} = c't^{-1/2}$. As in Section 6.2, we have

$$\nu(|Av| > t) \geq \frac{1}{2d} \mathbf{WSF}(|\text{past}_o| > t),$$

and the claim of the theorem follows. □

7.6 Upper bounds: $d \geq 5$

Unlike the case of the radius, an upper bound on the size of waves does not yield directly an upper bound on the size of the avalanche. However, we can still get a power law upper bound, that we prove in this section.

Proof of theorem 1.5(iv); upper bound. Fix some $k \geq 1$. Recalling that N denotes the number of waves, we have

$$\nu(S > t) = \nu(S > t, N > k) + \nu(S > t, N \leq k). \tag{7.23}$$

From Lemma 2.3 we have $N \leq R$, and we can upper bound the first term in the expression above by $\nu(R > k)$. Next, writing S^j for the size of the j -th wave, if $S > t$ and $N \leq k$, then we have $S^j > t/k$ for some $1 \leq j \leq N$. Hence,

$$\begin{aligned} \nu(S > t) &\leq \nu(R > k) + \nu(S^j > t/k, \text{ for some } j \leq N) \\ &\leq C \mathbf{WSF}_o(\text{diam}(\mathfrak{X}_o) > k) + C \mathbf{WSF}_o(|\mathfrak{X}_o| > t/k) \\ &\leq k^{-2+o(1)} + C \mathbf{WSF}_o(|\mathfrak{X}_o| > t/k), \end{aligned} \tag{7.24}$$

where we used Theorem 1.2(iv) for the first term. We now control the second term on the right in equation (7.24).

Lemma 7.9. $\mathbf{WSF}_o(|\mathfrak{T}_o| > t) \leq t^{-1/2+o(1)}$.

Proof. We note that an application of Theorem 1.2(iv) yields

$$\begin{aligned} \mathbf{WSF}_o(|\mathfrak{T}_o| > t) &\leq \mathbf{WSF}_o(\text{diam}(\mathfrak{T}_o) > t^{1/4}) + \mathbf{WSF}_o(|B(t^{1/4}) \cap \mathfrak{T}_o| > t) \\ &\leq (t^{1/4})^{-2+o(1)} + \mathbf{WSF}_o(|B(t^{1/4}) \cap \mathfrak{T}_o| > t). \end{aligned} \quad (7.25)$$

Due to Wilson's algorithm, we have $\mathbf{WSF}_o(x \in \mathfrak{T}_o) \leq C|x|^{2-d}$, and hence we have

$$\mathbf{WSF}_o(|B(t^{1/4}) \cap \mathfrak{T}_o| > t) \leq \frac{\mathbf{E}\mathbf{WSF}_o|B(t^{1/4}) \cap \mathfrak{T}_o|}{t} \leq \frac{C(t^{1/4})^2}{t} = Ct^{-1/2}.$$

□

Using the above lemma and optimizing the number of waves k in (7.24), we get $\nu(S > t) \leq t^{-2/5+o(1)}$, thereby completing the proof of the upper bound in Theorem 1.5(iv). □

References

- [1] D. J. Aldous and R. Lyons (2007). Processes on unimodular random networks. *Electron. J. Probab.* **12** no. 54, 1454–1508 (electronic). MR-2354165
- [2] S. R. Athreya and A. A. Járai (2004). Infinite volume limit for the stationary distribution of Abelian sandpile models. *Commun. Math. Phys.* **249** 197–213. MR-2077255
- [3] P. Bak, C. Tang and K. Wiesenfeld (1988). Self-organized criticality, *Phys. Rev. A* **38** 364–374. MR-0949160
- [4] M. T. Barlow and D. Karli (2015). Some remarks on uniform boundary Harnack Principles. *Preprint*. arXiv:1507.04115
- [5] M. T. Barlow and A. A. Járai (2016). Geometry of uniform spanning forest components in high dimensions. *Preprint*. arXiv:1602.01505
- [6] I. Benjamini, R. Lyons, Y. Peres and O. Schramm (2001). Uniform spanning forests. *Ann. Probab.* **29**, 1–65. MR-1825141
- [7] I. Benjamini and O. Schramm (2001). Recurrence of distributional limits of finite planar graphs. *Electron. J. Probab.* **6**, no. 23, 13 pp. (electronic). MR-1873300
- [8] D. Dhar (1990). Self-organized critical state of sandpile automaton models. *Phys. Rev. Lett.* **64**, 1613–1616. MR-1044086
- [9] D. Dhar (2006). Theoretical studies of self-organized criticality. *Phys. A:* **369**, 29–70. MR-2246566
- [10] S. L. Gamlin and A. A. Járai (2014). Anchored burning bijections on finite and infinite graphs, *Electron. J. Probab.* **19**, no. 117, 23 pp. MR-3296533
- [11] A. Fey-den Boer and F. Redig (2008). Limiting shapes for deterministic centrally seeded growth models. *J. Stat. Phys.* **130**, 579–597. MR-2378168
- [12] A. E. Holroyd, L. Levine, K. Mészáros, Y. Peres, J. Propp and D. B. Wilson (2008). Chip-firing and rotor-routing on directed graphs. *In and out of equilibrium. 2*, *Progr. Probab.* **60**, Birkhäuser, Basel, 331–364. MR-2477390
- [13] E. V. Ivashkevich, D. V. Ktitarev and V. B. Priezzhev (1994). Waves of topplings in an Abelian sandpile. *Physica A: Statistical Mechanics and its Applications*, **209**, 347–360.
- [14] A. A. Járai (2014). Sandpile models. *Preprint*. arXiv:1401.0354 MR-2952022
- [15] A. A. Járai and F. Redig (2008). Infinite volume limit of the Abelian sandpile model in dimensions $d \geq 3$. *Probab. Theory Related Fields* **141**, 181–212. MR-2372969
- [16] A. A. Járai, F. Redig and E. Saada (2015). Approaching criticality via the zero-dissipation limit in the abelian avalanche model. *J. Stat. Phys.* **159**, 1369–1407. MR-3350375
- [17] A. A. Járai and N. Werning (2014). Minimal configurations and sandpile measures. *J. Theoret. Probab.* **27**, 153–167. MR-3174221

- [18] R. Kenyon (2000). The asymptotic determinant of the discrete Laplacian. *Acta Math.* **185**, 239–286. MR-1819995
- [19] H. Kesten (1987). Hitting probabilities of random walks on \mathbf{Z}^d . *Stochastic Process. Appl.* **25**, 165–184. MR-0915132
- [20] H. Kesten (1987). Scaling relations for 2D-percolation. *Comm. Math. Phys.* **109**, 109–156. MR-0879034
- [21] G. F. Lawler (1995). The logarithmic correction for loop-erased walk in four dimensions. *Proceedings of the Conference in Honor of Jean-Pierre Kahane (Orsay, 1993)*, J. Fourier Anal. Appl. Special Issue, 347–361. MR-1364896
- [22] G. F. Lawler (1996). Cut times for simple random walk. *Electron. J. Probab.* **1**, Paper no. 13. MR-1423466
- [23] G. F. Lawler (1996). *Intersections of Random Walks*. Birkhäuser, Boston. MR-1117680
- [24] G. F. Lawler and V. Limic (2010). *Random walk: a modern introduction*. Cambridge studies in advanced mathematics, Vol. **123**, Cambridge University Press, Cambridge. MR-2677157
- [25] R. Lyons, B. J. Morris and O. Schramm (2008). Ends in uniform spanning forests. *Electron. J. Probab.* **13**, Paper no. 58. MR-2448128
- [26] R. Lyons and Y. Peres (2016). *Probability on trees and networks*, Cambridge Series in Statistical and Probabilistic Mathematics, **42**, Cambridge University Press, New York. MR-3616205
- [27] R. Lyons, Y. Peres and O. Schramm (2003). Markov chain intersections and the loop-erased walk. *Ann. Inst. H. Poincaré Probab. Statist.* **39**, 779–791. MR-1997212
- [28] S.N. Majumdar and D. Dhar (1992). Equivalence between the Abelian sandpile model and the $q \rightarrow 0$ limit of the Potts model. *Physica A: Statistical Mechanics and its Applications* **185**, 129–145.
- [29] R. Masson (2009). The growth exponent for planar loop-erased random walk. *Electron. J. Probab.* **14**, Paper no. 36. MR-2506124
- [30] B. Morris (2003). The components of the wired spanning forest are recurrent. *Probab. Theory Related Fields* **125**, no. 2, 259–265. MR-1961344
- [31] R. Pemantle (1991). Choosing a spanning tree for the integer lattice uniformly. *Ann. Probab.* **19**, 1559–1574. MR-1127715
- [32] V. B. Priezzhev (2000). The upper critical dimension of the Abelian sandpile model. *J. Statist. Phys.* **98**, 667–684. MR-1749227
- [33] F. Redig (2006). Mathematical aspects of the abelian sandpile model, in *Mathematical statistical physics*, 657–729, Elsevier B. V., Amsterdam. MR-2581895
- [34] D. Shiraishi (2016). Growth exponent for loop-erased random walk in three dimensions. *Preprint*. arXiv:1310.1682.
- [35] D. B. Wilson (1996). Generating random spanning trees more quickly than the cover time, in *Proceedings of the Twenty-eighth Annual ACM Symposium on the Theory of Computing (Philadelphia, PA, 1996)*, 296–303, ACM, New York. MR-1427525
- [36] W. Woess (2000). *Random walks on infinite graphs and groups*. Cambridge Tracts in Mathematics **138**, Cambridge University Press, Cambridge. MR-1743100

Acknowledgments. The authors thank Russ Lyons for helpful discussions. J.H. and A.A.J. also thank the organizers of the 2014 Bath Summer School, where some of this work was initiated.

J.H. thanks Michael Damron for postdoctoral advising and encouragement.

S.B. thanks Russ Lyons for support, encouragement and advice.

Electronic Journal of Probability

Electronic Communications in Probability

Advantages of publishing in EJP-ECP

- Very high standards
- Free for authors, free for readers
- Quick publication (no backlog)
- Secure publication (LOCKSS¹)
- Easy interface (EJMS²)

Economical model of EJP-ECP

- Non profit, sponsored by IMS³, BS⁴, ProjectEuclid⁵
- Purely electronic

Help keep the journal free and vigorous

- Donate to the IMS open access fund⁶ (click here to donate!)
- Submit your best articles to EJP-ECP
- Choose EJP-ECP over for-profit journals

¹LOCKSS: Lots of Copies Keep Stuff Safe <http://www.lockss.org/>

²EJMS: Electronic Journal Management System <http://www.vtex.lt/en/ejms.html>

³IMS: Institute of Mathematical Statistics <http://www.imstat.org/>

⁴BS: Bernoulli Society <http://www.bernoulli-society.org/>

⁵Project Euclid: <https://projecteuclid.org/>

⁶IMS Open Access Fund: <http://www.imstat.org/publications/open.htm>