THE APPROACH TO CRITICALITY IN SANDPILES

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ABSTRACT. A popular theory of self-organized criticality relates the critical behavior of driven dissipative systems to that of systems with conservation. In particular, this theory predicts that the stationary density of the abelian sandpile model should be equal to the threshold density of the corresponding fixed-energy sandpile. This "density conjecture" has been proved for the underlying graph $\mathbb{Z}$. We show (by simulation or by proof) that the density conjecture is *false* when the underlying graph is any of $\mathbb{Z}^2$, the complete graph K_n, the Cayley tree, the ladder graph, the bracelet graph, or the flower graph. Driven dissipative sandpiles continue to evolve even after a constant fraction of the sand has been lost at the sink. These results cast doubt on the validity of using fixed-energy sandpiles to explore the critical behavior of the abelian sandpile model at stationarity.

1. INTRODUCTION

In a widely cited series of papers [DVZ98, VDMZ98, DMVZ00, VDMZ00, MDPS$^+$01], Dickman, Muñoz, Vespignani and Zapperi (DMVZ) developed a theory of self-organized criticality as a relationship between driven dissipative systems and systems with conservation. This theory predicts a specific relationship between the classical abelian sandpile model of Bak, Tang, and Wiesenfeld [BTW87], a driven system in which particles added at random dissipate across the boundary, and the corresponding "fixed-energy sandpile," a closed system in which the total number of particles is conserved.

In this introduction, we briefly define these two models and explain the conjectured relationship between them in the DMVZ paradigm of self-organized criticality. In particular, we focus on the prediction that the stationary density of the driven dissipative model equals the threshold density of the fixed-energy sandpile model. In section 2, we present data from large-scale simulations which strongly indicate that this conjecture is false on the two-dimensional square lattice $\mathbb{Z}^2$. In the subsequent sections we expand on the results announced in [FLW10] by examining the conjecture on some simpler families of graphs in which we can provably refute it.

The difference between the stationary and threshold densities on most of these graphs is fairly small — typically on the order of 0.01% to 0.2% — which explains why many previous simulations did not uncover them. The exception is the early experiments by Grassberger

and Manna [GM90], who clearly identified this discrepancy, at least in dimensions 4 and higher. Later studies focused on dimension 2 and missed the discrepancy.

In some more recent papers such as [BM08], the DMVZ paradigm is explicitly restricted to stochastic models. In other recent papers [VD05, dCVdSD09] it is claimed to apply both to stochastic and deterministic sandpiles, although these papers focus on stochastic sandpiles, for the reason that deterministic sandpiles are said to belong to a different universality class.

Despite our contrary findings, we believe that the central idea of the DMVZ paradigm is a good one: even in the deterministic case, the dynamics of a driven dissipative system should in some way reflect the dynamics of the corresponding conservative system. Our results point to a somewhat different relationship than that posited in the DMVZ series of papers: the driven dissipative model exhibits a second-order phase transition at the threshold density of the conservative model. We explain this transition in section 3.

Bak, Tang, and Wiesenfeld [BTW87] introduced the abelian sandpile as a model of self-organized criticality; for mathematical background, see [Red05]. The model begins with a collection of particles on the vertices of a finite graph. A vertex having at least as many particles as its degree *topples* by sending one particle along each incident edge. A subset of the vertices are distinguished as sinks: they absorb particles but never topple. A single time step consists of adding one particle at a random site, and then performing topplings until each non-sink vertex has fewer particles than its degree. The order of topplings does not affect the outcome [Dha90]. The set of topplings that occur before the system stabilizes is called an avalanche.

Avalanches can be decomposed into a sequence of "waves" so that each site topples at most once during each wave. Over time, sandpiles evolve toward a stationary state in which the waves exhibit power-law statistics [KLGP00] (though the full avalanches seem to exhibit multifractal behavior [DMST98, KMS05]). Power law statistics are a hallmark of criticality, and since the stationary state is reached apparently without tuning of a parameter, the model is said to be *self-organized critical.*

To explain how the sandpile model self-organizes to reach the critical state, Dickman *et al.* [DVZ98, DMVZ00] introduced an argument which soon became widely accepted: see, for example, [Sor04, Ch. 15.4.5] and [MQ05, FdBR05, RS09]. Despite the apparent lack of a free parameter, they argued, the dynamics implicitly involve the tuning of a parameter to a value where a phase transition takes place. The phase transition is between an active state, where topplings take place, and a quiescent "absorbing" state where topplings have died out. The parameter is the *density*, the average number of particles per site. When the system is quiescent, addition of new particles increases the density. When the system is active, particles are lost to the sinks via toppling, decreasing the density. The dynamical rule "add a particle when all activity has died out" ensures that these two density changing mechanisms balance one another out, driving the system to the threshold of instability.

To explore this idea, DMVZ introduced the *fixed-energy sandpile* model (FES), which involves an explicit free parameter ζ, the density of particles. On a graph with N vertices, the system starts with ζN particles at vertices chosen independently and uniformly at random. Unlike the driven dissipative sandpile described above, there are no sinks and no addition

of particles, so the total number of particles is conserved. Subsequently the system evolves through toppling of unstable sites. Usually the parallel toppling order is chosen: at each time step, all unstable sites topple simultaneously. In the mathematical literature, this system goes by the name of *parallel chip-firing* [BG92, BLS91]. Toppling may persist forever, or it may stop after some finite time. In the latter case, we say that the system *stabilizes*; in the terminology of DMVZ, it reaches an "absorbing state."

A common choice of underlying graph is the $n \times n$ square grid with periodic boundary conditions. It is believed, and supported by simulations [BCFV03], that there is a *threshold density* ζ_c, such that for $\zeta < \zeta_c$, the system stabilizes with probability tending to 1 as $n \to \infty$; and for $\zeta > \zeta_c$, with probability tending to 1 the system does not stabilize.

For the driven dissipative sandpile on the $n \times n$ square grid, as $n \to \infty$ the stationary measure has an infinite-volume limit [AJ04], which is a measure on sandpiles on $\mathbb{Z}^2$. It turns out that one gets the same limiting measure whether the grid has periodic or open boundary conditions, and whether there is one sink vertex or the whole boundary serves as a sink [AJ04] (see also [Pem91] for the corresponding result on random spanning trees). The statistical properties of this limiting measure have been much studied [Pri94a, JPR06]. Of particular interest is the *stationary density* ζ_s of $\mathbb{Z}^2$, defined as the expected number of particles at a fixed site. Grassberger conjectured that ζ_s is exactly 17/8, and it is now known that $\zeta_s = 17/8 \pm 10^{-12}$ [JPR06].

DMVZ believed that the combination of driving and dissipation in the classical abelian sandpile model should push it toward the critical density ζ_c of the fixed-energy sandpile. This leads to a specific testable prediction, which we call the Density Conjecture.

Conjecture 1.1 (Density Conjecture, [VDMZ00])**.** *On the square grid, $\zeta_c = 17/8$.*

One can also formulate a density conjecture for more general graphs, where it takes the form $\zeta_c = \zeta_s$. We give precise definitions of the densities ζ_c and ζ_s in section 2.

Vespignani *et al.* [VDMZ00] write of the fixed-energy sandpile on the square grid, "the system turns out to be critical only for a particular value of the energy density equal to that of the stationary, slowly driven sandpile." They add that the threshold density ζ_c of the fixed-energy sandpile is "the only possible stationary value for the energy density" of the driven dissipative model. In simulations they find $\zeta_c = 2.1250(5)$, adding in a footnote "It is likely that, in fact, 17/8 is the exact result."

Muñoz *et al.* [MDPS$^+$01] have also expressed this view, asserting that "FES are found to be critical only for a particular value $\zeta = \zeta_c$ (which as we will show turns out to be identical to the stationary energy density of its driven dissipative counterpart)."

Our goal in the present paper is to demonstrate that the density conjecture is more problematic than it first appears. Table 1 presents data from large-scale simulations which strongly suggest that ζ_c is close to but not exactly equal to 17/8 (see also Table 3). We further consider several other families of graphs, including some for which we can determine the exact values ζ_c and ζ_s analytically. We find that they are close, but not equal.

All now known information on the threshold density ζ_c and stationary density ζ_s is summarized in Table 2. The only graph on which the two densities are known to be equal is $\mathbb{Z}$

n	#trials	estimate of $\zeta_c(\mathbb{Z}_n^2)$
64	2^{28}	2.1249561 ± 0.0000004
128	2^{26}	2.1251851 ± 0.0000004
256	2^{24}	2.1252572 ± 0.0000004
512	2^{22}	2.1252786 ± 0.0000004
1024	2^{20}	2.1252853 ± 0.0000004
2048	2^{18}	2.1252876 ± 0.0000004
4096	2^{16}	2.1252877 ± 0.0000004
8192	2^{14}	2.1252880 ± 0.0000004
16384	2^{12}	2.1252877 ± 0.0000004

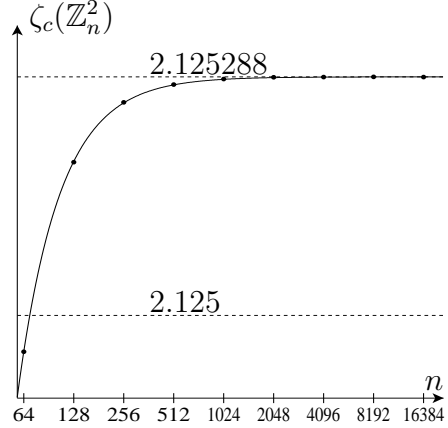

TABLE 1. Summary of our fixed-energy sandpile simulations on $n \times n$ tori $\mathbb{Z}_n^2$, giving our empirical estimate of the threshold density $\zeta_c(\mathbb{Z}_n^2)$. The standard deviation in each of our estimates of $\zeta_c(\mathbb{Z}_n^2)$ is 4×10^{-7}. The data are well approximated by $\zeta_c(\mathbb{Z}_n^2) = 2.1252881 \pm 3 \times 10^{-7} - (0.390 \pm 0.001)n^{-1.7}$, as shown in the graph. (The error bars are too small to be visible, so the data are shown as points.) We conclude that the asymptotic threshold density $\zeta_c(\mathbb{Z}^2)$ is 2.125288 to six decimal places. In contrast, the stationary density $\zeta_s(\mathbb{Z}^2)$ is 2.125000000000 to twelve decimal places.

graph	ζ_s	ζ_c
$\mathbb{Z}$	**1**	**1**
$\mathbb{Z}^2$	$\mathbf{17/8 = 2.125}$	$2.125288\ldots$
bracelet	$\mathbf{5/2 = 2.5}$	$\mathbf{2.496608\ldots}$
flower graph	$\mathbf{5/3 = 1.666667\ldots}$	$\mathbf{1.668898\ldots}$
ladder graph	$\mathbf{\frac{7}{4} - \frac{\sqrt{3}}{12} = 1.605662\ldots}$	$1.6082\ldots$
complete graph	$n/2 + O(\sqrt{n})$	$n - O(\sqrt{n \log n})$
3-regular tree	$\mathbf{3/2}$	$1.50000\ldots$
4-regular tree	**2**	$2.00041\ldots$
5-regular tree	$\mathbf{5/2}$	$2.51167\ldots$

TABLE 2. Stationary and threshold densities for different graphs. Exact values are in bold.

[MQ05, FdBR05, FdBMR09]. On all other graphs we examined, with the possible exception of the 3-regular tree, it appears that $\zeta_c \neq \zeta_s$.

Taken together, these results show that the conclusions of [MDPS$^+$01] that "FES are shown to exhibit an absorbing state transition with critical properties coinciding with those of the corresponding sandpile model" deserve to be re-evaluated. One hope of the DMVZ paradigm was that critical features of the driven dissipative model, such as the exponents governing the distribution of avalanche sizes and decay of correlations, might be more easily

studied in FES by examining the scaling behavior of these observables as $\zeta \uparrow \zeta_c$. However, the failure of the density conjecture suggests that the two models may not share the same critical features.

As Grassberger and Manna observed [GM90], the value of the FES threshold density depends on the choice of initial condition. One might consider a more general version of FES, namely adding $(\zeta - \zeta_0)N$ particles at random to a "background" configuration τ of density ζ_0 already present on the grid. For example, taking τ to be the deterministic configuration on $\mathbb{Z}^d$ of $2d-2$ particles everywhere, by [FLP10, Prop 1.4] we obtain a threshold density of $\zeta_c = 2d - 2$. Many interesting questions present themselves: for instance, for which background does ζ_c take the smallest value, and for which backgrounds do we obtain $\zeta_c = \zeta_s$? It would also be interesting to replicate the phase transition for driven sandpiles (see section 3) for different background configurations.

2. Sandpiles on the square grid $\mathbb{Z}^2$

In this section we give precise definitions of the stationary and threshold densities, and present the results of large-scale simulations on $\mathbb{Z}^2$. The definitions in this section apply to general graphs, but we defer the discussion of results about other graphs to subsequent sections.

2.1. The driven dissipative sandpile and the stationary density ζ_s.

Let $\hat{G} = (V, E)$ be a finite graph, which may have loops and multiple edges. Let $S \subset V$ be a nonempty set of vertices, which we will call *sinks*. The presence of sinks distinguishes the driven dissipative sandpile from its fixed-energy counterpart. To highlight this distinction, throughout the paper, graphs denoted with a "hat" as in $\hat{G}$ have sinks, and those without a hat as in G do not.

For vertices $v, w \in V$, write $a_{v,w} = a_{w,v}$ for the number of edges connecting v and w, and

$$d_v = \sum_{w \in V} a_{v,w}.$$

for the number of edges incident to v. A *sandpile* (or "configuration") η on $\hat{G}$ is a map
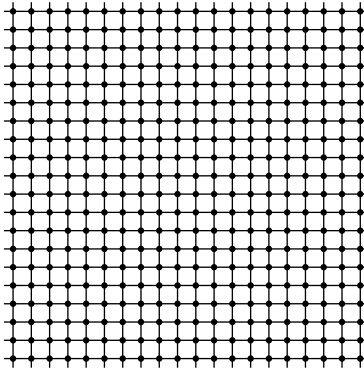
$$\eta : V \to \mathbb{Z}_{\geq 0}.$$

We interpret $\eta(v)$ as the number of sand particles at the vertex v; we will sometimes call this number the *height* of v in η.

A vertex $v \notin S$ is called *unstable* if $\eta(v) \geq d_v$. An unstable vertex can *topple* by sending one particle along each edge incident to v. Thus, toppling v results in a new sandpile η' given by

$$\eta' = \eta + \Delta_v$$

where

$$\Delta_v(w) = \begin{cases} a_{v,w}, & v \neq w \\ a_{v,v} - d_v, & v = w. \end{cases}$$

FIGURE 1. The square grid $\mathbb{Z}^2$.

Sinks by definition are always stable, and never topple. If all vertices are stable, we say that η is stable.

Note that toppling a vertex may cause some of its neighbors to become unstable. The *stabilization* η° of η is a sandpile resulting from toppling unstable vertices in sequence, until all vertices are stable. By the *abelian property* [Dha90], the stabilization is unique: it does not depend on the toppling sequence. Moreover, the number of times a given vertex topples does not depend on the toppling sequence.

The most commonly studied example is the $n \times n$ square grid graph, with the boundary sites serving as sinks (Figure 1).

The driven dissipative sandpile model is a continuous time Markov chain $(\eta_t)_{t \geq 0}$ whose state space is the set of stable sandpiles on $\hat{G}$. Let $V' = V \setminus S$ be the set of vertices that are not sinks. At each site $v \in V'$, particles are added at rate 1. When a particle is added, topplings occur instantaneously to stabilize the sandpile. Writing $\sigma_t(v)$ for the total number of particles added at v before time t, we have by the abelian property

$$\eta_t = (\sigma_t)^\circ.$$

Note that for fixed t, the random variables $\sigma_t(v)$ for $v \in V'$ are independent and have the Poisson distribution with mean t.

The model just described is most commonly known as the abelian sandpile model (ASM), but we prefer the term "driven dissipative" to distinguish it from the fixed-energy sandpile described below, which is also a form of ASM. "Driven" refers to the addition of particles, and "dissipative" to the loss of particles absorbed by the sinks.

Dhar [Dha90] developed the *burning algorithm* to characterize the recurrent sandpile states, that is, those sandpiles η for which, regardless of the initial state,

$$\Pr(\eta_t = \eta \text{ for some } t) = 1.$$

Lemma 2.1 (Burning Algorithm [Dha90]). *A sandpile η is recurrent if and only if every non-sink vertex topples exactly once during the stabilization of $\eta + \sum_{s \in S} \Delta_s$, where the sum is over sink vertices S.*

The recurrent states form an abelian group under the operation of addition followed by stabilization. In particular, the stationary distribution of the Markov chain η_t is uniform on the set of recurrent states.

The combination of driving and dissipation organizes the system into a critical state. To measure the density of particles in this state, we define the *stationary density* $\zeta_s(\hat{G})$ as

$$\zeta_s(\hat{G}) = \mathbb{E}_\mu \left[\frac{1}{\#V'} \sum_{v \in V'} \eta(v) \right]$$

where $V' = V \setminus S$, and μ is the uniform measure on recurrent sandpiles on $\hat{G}$. The stationary density has another expression in terms of the Tutte polynomial of the graph obtained from $\hat{G}$ by collapsing the set S of sinks to a single vertex; see section 4.

Most of the graphs we will study arise naturally as finite subsets of a locally finite graph Γ, i.e., Γ is a countably infinite graph in which every vertex has finite degree. (We also study the complete graph and the flower graph, which do not arise in this way.) Let $\hat{G}_n$ for $n \geq 1$ be a nested family of finite induced subgraphs with $\bigcup \hat{G}_n = \Gamma$. As sinks in $\hat{G}_n$ we take the set of boundary vertices

$$S_n = \hat{G}_n - \hat{G}_{n-1}.$$

In cases where the free and wired limits are different, such as on regular trees, we will choose a sequence $\hat{G}_n$ corresponding to the wired limit. We denote by μ_n the uniform measure on recurrent configurations on $\hat{G}_n$.

We are interested in the *stationary density*

$$\zeta_s(\Gamma) := \lim_{n \to \infty} \zeta_s(\hat{G}_n).$$

When $\Gamma = \mathbb{Z}^d$, it is known that the infinite-volume limit of measures $\mu = \lim_{n \to \infty} \mu_n$ exists and is translation-invariant [AJ04]. In this case it follows that the limit defining $\zeta_s(\Gamma)$ exists and equals

$$\zeta_s = \mathbb{E}_\mu[\eta(0)].$$

For other families of graphs we consider, we will show that the limit defining $\zeta_s(\Gamma)$ exists.

Much is known about the limiting measure μ in the case $\Gamma = \mathbb{Z}^2$. The following expressions have been obtained for ζ_s and the single site height probabilities. The symbol $\overset{?}{=}$ denotes expressions that are rigorous up to a conjecture [JPR06] that a certain integral, numerically evaluated as 0.5 ± 10^{-12}, is exactly $1/2$.

$$\zeta_s(\mathbb{Z}^2) \overset{?}{=} 17/8 \quad \text{[JPR06]},$$
$$\mu\{\eta(x) = 0\} = \frac{2}{\pi^2} - \frac{4}{\pi^3} \quad \text{[MD91]},$$
$$\mu\{\eta(x) = 1\} \overset{?}{=} \frac{1}{4} - \frac{1}{2\pi} - \frac{2}{\pi^2} + \frac{12}{\pi^3} \quad \text{[Pri94a, JPR06]},$$
$$\mu\{\eta(x) = 2\} \overset{?}{=} \frac{3}{8} + \frac{1}{\pi} - \frac{12}{\pi^3} \quad \text{[JPR06], and}$$
$$\mu\{\eta(x) = 3\} \overset{?}{=} \frac{3}{8} - \frac{1}{2\pi} + \frac{1}{\pi^2} + \frac{4}{\pi^3} \quad \text{[JPR06]}.$$

2.2. **The fixed-energy sandpile model and the threshold density ζ_c.** Next we describe
the fixed-energy sandpile model, in which the driving and dissipation are absent, and the
total number of particles is conserved. As before, let G be a finite graph, possibly with loops
and multiple edges. Unlike the driven dissipative model, we do not single out any vertices
as sinks. The fixed-energy sandpile evolves in discrete time: at each time step, all unstable
vertices topple once in parallel. Thus the configuration η_{j+1} at time $j + 1$ is given by

$$\eta_{j+1} = \eta_j + \sum_{v \in U_j} \Delta_v$$

where

$$U_j = \{v \in V \,:\, \eta_j(v) \geq d_v\}$$

is the set of vertices that are unstable at time j. We say that η_0 *stabilizes* if toppling
eventually stops, i.e. $U_j = \varnothing$ for all sufficiently large j.

If η_0 stabilizes, then there is some site that never topples [Tar88] (see also [FdBMR09,
Theorem 2.8, item 4] and [FLP10, Lemma 2.2]). Otherwise, for each site x, let $j(x)$ be
the last time x topples. Choose a site x minimizing $j(x)$. Then each neighbor y of x has
$j(y) \geq j(x)$, so y topples at least once at or after time $j(x)$. Thus x receives at least d_x
additional particles and must topple again after time $j(x)$, a contradiction. This criterion is
very useful in simulations: as soon as every site has toppled at least once, we know that the
system will not stabilize.

Let $(\sigma_\lambda(v))_{\lambda \geq 0}$ be a collection of independent Poisson point processes of intensity 1, indexed
by the vertices of G. So each $\sigma_\lambda(v)$ has the Poisson distribution with mean λ. We define the
threshold density of G as

$$\zeta_c(G) = \mathbb{E}\Lambda_c,$$

where

$$\Lambda_c = \sup\{\lambda \,:\, \sigma_\lambda \text{ stabilizes}\}.$$

We expect that Λ_c is tightly concentrated around its mean when G is large. Indeed, if Γ is an
infinite vertex-transitive graph, then the event that σ_λ stabilizes on Γ is translation-invariant.
By the ergodicity of the Poisson product measure, this event has probability 0 or 1. Since
this probability is monotone in λ, there is a (deterministic) threshold density $\zeta_c(\Gamma)$, such
that

$$\Pr[\sigma_\lambda \text{ stabilizes on } \Gamma] = \begin{cases} 1, & \lambda < \zeta_c(\Gamma) \\ 0, & \lambda > \zeta_c(\Gamma). \end{cases}$$

We expect the threshold densities on natural families of finite graphs to satisfy a law of large
numbers such as the following.

Conjecture 2.2. *With probability 1,*

$$\Lambda_c(\mathbb{Z}_n^2) \to \zeta_c(\mathbb{Z}^2) \quad \text{as } n \to \infty.$$

Previous simulations ($n = 160$ [DVZ98]; $n = 1280$ [VDMZ98]) to estimate the threshold
density $\zeta_c(\mathbb{Z}^2)$ found a value of 2.125, in agreement with the stationary density $\zeta_s(\mathbb{Z}^2) \overset{?}{=} 17/8$.
By performing larger-scale simulations, however, we find that ζ_c exceeds ζ_s.

grid size (n^2)	#samples	$\zeta_c(\mathbb{Z}_n^2)$	distribution of height h of sand				#topplings $\div n^3$
			$\Pr[h=0]$	$\Pr[h=1]$	$\Pr[h=2]$	$\Pr[h=3]$	
64^2	268435456	2.124956	0.073555	0.173966	0.306447	0.446032	0.197110
128^2	67108864	2.125185	0.073505	0.173866	0.306567	0.446062	0.197808
256^2	16777216	2.125257	0.073488	0.173835	0.306609	0.446068	0.198789
512^2	4194304	2.125279	0.073481	0.173826	0.306626	0.446067	0.200162
1024^2	1048576	2.125285	0.073479	0.173822	0.306633	0.446066	0.201745
2048^2	262144	2.125288	0.073478	0.173821	0.306635	0.446065	0.203378
4096^2	65536	2.125288	0.073477	0.173821	0.306637	0.446064	0.205323
8192^2	16384	2.125288	0.073477	0.173821	0.306638	0.446064	0.206475
16384^2	4096	2.125288	0.073478	0.173821	0.306638	0.446064	0.208079
$\mathbb{Z}^2$ (stationary)		2.125000	0.073636	0.173900	0.306291	0.446172	

TABLE 3. Fixed-energy sandpile simulations on $n \times n$ tori $\mathbb{Z}_n^2$. The third column gives our empirical estimate of the threshold density $\zeta_c(\mathbb{Z}_n^2)$. The next four columns give the empirical distribution of the height of a fixed vertex in the stabilization $(\sigma_\lambda)^\circ$, for λ just below Λ_c. Each estimate of the expectation $\zeta_c(\mathbb{Z}_n^2)$ and of the marginals $\Pr[h=i]$ has standard deviation less than $4 \cdot 10^{-7}$. The total number of topplings needed to stabilize σ_λ appears to scale as n^3.

Table 3 summarizes the results of our simulations, which indicate that $\zeta_c(\mathbb{Z}^2)$ equals 2.125288 to six decimal places. In each random trial, we add particles one at a time at uniformly random sites of the $n \times n$ torus. After each addition, we perform topplings until either all sites are stable, or every site has toppled at least once since the last addition. For deterministic sandpiles on a connected graph, if every site topples at least once, the system will never stabilize [FdBMR09, FLP10]. We record m/n^2 as an empirical estimate of the threshold density, where m is the maximum number of particles for which the configuration stabilizes. We then average these empirical estimates were over many independent trials. The one-site marginals we report are obtained from the stable configuration just before the $(m+1)^{\text{st}}$ particle was added, and the number of topplings reported is the total number of topplings required to stabilize the first m particles.

We used a random number generator based on the Advanced Encryption Standard (AES-256), which has been found to exhibit good randomness properties. Our simulations were conducted on a High Performance Computing (HPC) cluster of computers.

3. Sandpiles on the bracelet

Next we examine a family of graphs for which we can determine ζ_c and ζ_s exactly and prove that they are not equal. Despite this inequality, we show that an interesting connection remains between the driven dissipative and conservative dynamics: the threshold density of the conservative model is the point at which the driven dissipative model begins to lose a macroscopic amount of sand to the sink.

The *bracelet graph* B_n (Figure 2) is a multigraph with vertex set $\mathbb{Z}_n$ (the n-cycle) with the usual edge set $\{(i, i+1 \bmod n) : 0 \le i < n\}$ doubled. Thus all vertices have degree 4. The

graph $\hat{B}_n$ is the same, except that vertex 0 is distinguished as a sink from which particles disappear from the system. We denote by B_∞ the infinite path $\mathbb{Z}$ with doubled edges.

For $\lambda > 0$, let σ_λ be the configuration with Poisson(λ) particles independently on each site of $\hat{B}_n$. Let $\eta_\lambda = (\sigma_\lambda)^\circ$ be the stabilization of σ_λ, and let

$$\rho_n(\lambda) = \frac{1}{n-1} \sum_{x=1}^{n-1} \eta_\lambda(x)$$

be the final density. The following theorem gives the threshold and stationary densities of the infinite bracelet graph B_∞, and identifies the $n \to \infty$ limit of the final density $\rho_n(\lambda)$ as a function of the initial density λ.

Theorem 3.1. *For the bracelet graph,*

(1) *The threshold density $\zeta_c(B_\infty)$ is the unique positive root of $\zeta = \frac{5}{2} - \frac{1}{2}e^{-2\zeta}$ (numerically, $\zeta_c = 2.496608$).*
(2) *The stationary density $\zeta_s(B_\infty)$ is $5/2$.*
(3) *$\rho_n(\lambda) \to \rho(\lambda)$ in probability as $n \to \infty$, where*

$$\rho(\lambda) = \min\left(\lambda, \frac{5 - e^{-2\lambda}}{2}\right) = \begin{cases} \lambda, & \lambda \le \zeta_c \\ \frac{5 - e^{-2\lambda}}{2}, & \lambda > \zeta_c. \end{cases}$$

Part 3 of this theorem shows that the final density undergoes a second-order phase transition at ζ_c: the derivative of $\rho(\lambda)$ is discontinuous at $\lambda = \zeta_c$ (Figure 3). Thus in spite of the fact that $\zeta_s \ne \zeta_c$, there remains a connection between the conservative dynamics used to define ζ_c and the driven-dissipative dynamics used to define ζ_s. For $\lambda < \zeta_c$, very little dissipation takes place, so the final density equals the initial density λ; for $\lambda > \zeta_c$ a substantial amount of dissipation takes place, many particles are lost to the sink, and the final density is strictly less than the initial density. The sandpile continues to evolve as λ increases beyond ζ_c; in particular its density keeps changing.

We believe that this phenomenon is widespread. As evidence, in section 5 we introduce the "flower graph," which looks very different from the bracelet, and prove (in Theorem 5.5) that a similar phase transition takes place there.

For the proof of Theorem 3.1, we compare the dynamics of pairs of particles on the bracelet graph to single particles on $\mathbb{Z}$. At each vertex x of the bracelet, we group the particles starting
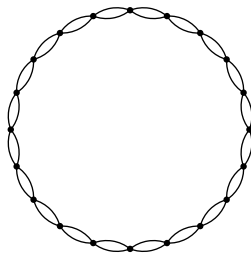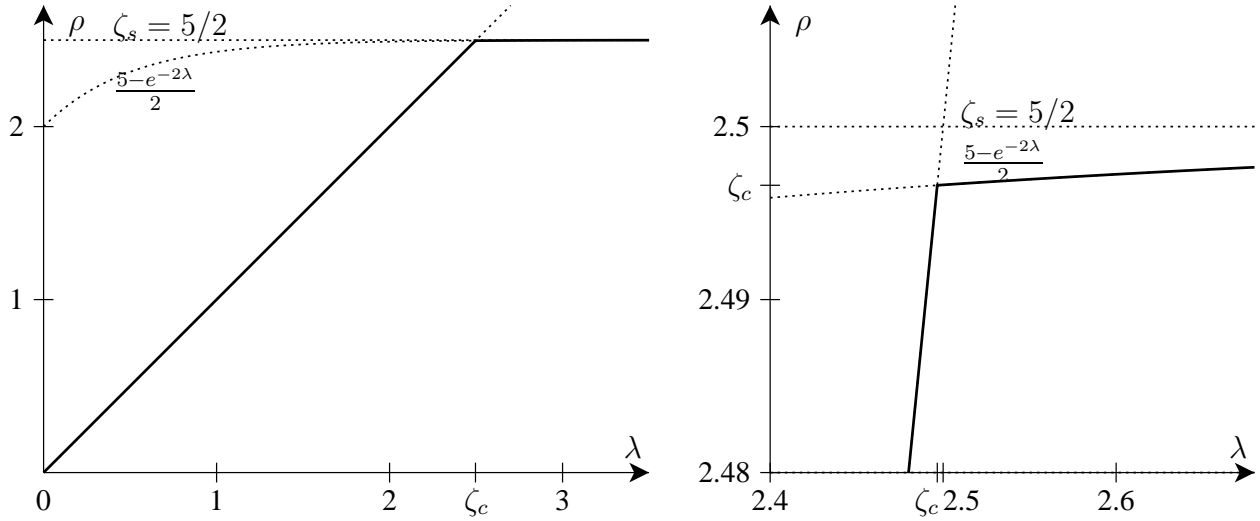

FIGURE 2. The bracelet graph B_{20}.

FIGURE 3. Density $\rho(\lambda)$ of the final stable configuration as a function of initial density λ, for the driven sandpile on the bracelet graph $\hat{B}_n$ as $n \to \infty$. A second-order phase transition occurs at $\lambda = \zeta_c$. Beyond this transition, the density of the driven sandpile continues to increase, approaching the stationary density ζ_s from below.

at x into pairs, with one "passive" particle left over if $\sigma_\lambda(x)$ is odd. Since all edges in the bracelet are doubled, we can ensure that in each toppling the two particles comprising a pair always move to the same neighbor, and that the passive particles never move. The toppling dynamics of the pairs are equivalent to the usual abelian sandpile dynamics on $\mathbb{Z}$.

We recall the relevant facts about one-dimensional sandpile dynamics:

- In any recurrent configuration on a finite interval of $\mathbb{Z}$, every site has height 1, except for at most one site of height 0. Therefore, $\zeta_s = 1$ [Red05].
- On $\mathbb{Z}$, an initial configuration distributed according to a nontrivial product measure with mean λ stabilizes almost surely (every site topples only finitely many times) if $\lambda < 1$, while it almost surely does not stabilize (every site topples infinitely often) if $\lambda \geq 1$ [FdBMR09]. Thus, $\zeta_c = 1$.

Proof of Theorem 3.1 parts 1 and 2. Given $\lambda > 0$, let λ^* be the pair density $\mathbb{E}\lfloor \sigma_\lambda(x)/2 \rfloor$, and let

$$p_{\text{odd}}(\lambda) = e^{-\lambda} \sum_{m \geq 0} \frac{\lambda^{2m+1}}{(2m+1)!} = \frac{1}{2}(1 - e^{-2\lambda}).$$

be the probability that a Poisson(λ) random variable is odd. Then λ and λ^* are related by

$$\lambda = 2\lambda^* + p_{\text{odd}}(\lambda). \tag{1}$$

The configuration σ_λ stabilizes on B_∞ if and only if the pair configuration σ_λ^* stabilizes on $\mathbb{Z}$. Thus $\zeta_c(B_\infty)^* = \zeta_c(\mathbb{Z})$. Setting $\lambda = \zeta_c(B_\infty)$ in (1), using the fact that $\zeta_c(\mathbb{Z}) = 1$, and that λ^*

is an increasing function of $\lambda > 0$, we conclude that $\zeta_c(B_\infty)$ is the unique positive root of

$$\zeta = 2 + p_{\mathrm{odd}}(\zeta),$$

or $\zeta = \frac{5}{2} - \frac{1}{2}e^{-2\zeta}$. This proves part 1.

For part 2, by the burning algorithm, a configuration σ on $\hat{B}_n$ is recurrent if and only if it has at most one site with fewer than two particles. Thus, in the uniform measure on recurrent configurations on $\hat{B}_n$,

$$\Pr(\sigma(x) = 2) = \Pr(\sigma(x) = 3) = \frac{1}{2} - \frac{1}{2n}, \qquad \Pr(\sigma(x) = 0) = \Pr(\sigma(x) = 1) = \frac{1}{2n}.$$

We conclude that $\zeta_s(\hat{B}_n) = \mathbb{E}\sigma(x) = \frac{5}{2} - \frac{2}{n} \to \frac{5}{2}$ as $n \to \infty$. □

To prove part 3 of Theorem 3.1, we use the following lemma, whose proof is deferred to the end of this section. Let $\hat{\mathbb{Z}}_n$ be the n-cycle with vertex 0 distinguished as a sink. Let σ'_λ be a sandpile on $\hat{\mathbb{Z}}_n$ distributed according to a product measure (not necessarily Poisson) of mean λ. Let η'_λ be the stabilization of σ'_λ, and let $\rho'_n(\lambda) = \frac{1}{n-1}\sum_{x=1}^{n-1} \eta'_\lambda(x)$ be the final density after stabilization.

Lemma 3.2. *On $\hat{\mathbb{Z}}_n$, we have $\rho'_n(\lambda) \to \min(\lambda, 1)$ in probability.*

Proof of Theorem 3.1, part 3. Let η_λ be the stabilization of σ_λ on $\hat{B}_n$, and let η^*_λ be the stabilization of $\sigma^*_\lambda = \lfloor \sigma_\lambda/2 \rfloor$ on $\hat{\mathbb{Z}}_n$. Then

$$\eta_\lambda(x) = 2\eta^*_\lambda(x) + \omega_\lambda(x) \tag{2}$$

where $\omega_\lambda(x) = \sigma_\lambda(x) - 2\sigma^*_\lambda(x)$ is 1 or 0 accordingly as $\sigma_\lambda(x)$ is odd or even. Let

$$\rho^*_n(\lambda) = \frac{1}{n-1}\sum_{x=1}^{n-1} \eta^*_\lambda(x)$$

be the final density after stabilization of σ^*_λ on $\hat{\mathbb{Z}}_n$. Then

$$\rho_n(\lambda) = 2\rho^*_n(\lambda) + \frac{1}{n-1}\sum_{x=1}^{n-1} \omega_\lambda(x).$$

By the weak law of large numbers, $\frac{1}{n-1}\sum_{x=1}^{n-1} \omega_\lambda(x) \to p_{\mathrm{odd}}(\lambda)$ in probability as $n \to \infty$. If $\lambda < \zeta_c$, then $\lambda^* < 1$, so by Lemma 3.2, $\rho^*_n(\lambda) \to \lambda^*$ in probability, and hence

$$\rho_n(\lambda) \to 2\lambda^* + p_{\mathrm{odd}}(\lambda) = \lambda$$

in probability. If $\lambda \geq \zeta_c$, then $\lambda^* \geq 1$, so by Lemma 3.2, $\rho^*_n(\lambda) \to 1$ in probability, hence

$$\rho_n(\lambda) \to 2 + p_{\mathrm{odd}}(\lambda) = \frac{5 - e^{-2\lambda}}{2}$$

in probability. This proves part 3. □

Proof of Lemma 3.2. We may view $\hat{\mathbb{Z}}_n$ as the path in $\mathbb{Z}$ from $a_n = -\lfloor n/2 \rfloor$ to $b_n = \lceil n/2 \rceil$, with both endpoints serving as sinks. For $x \in \hat{\mathbb{Z}}_n$, let $u_n(x)$ be the number of times that x topples during stabilization of the configuration σ'_λ on $\hat{\mathbb{Z}}_n$. Let $u_\infty(x)$ be the number of times x topples during stabilization of σ'_λ on $\mathbb{Z}$. The procedure of "toppling in nested volumes" [FdBMR09] shows that $u_n(x) \uparrow u_\infty(x)$ as $n \to \infty$.

We consider first $\lambda < 1$. In this case $u_\infty(x)$ is finite almost surely (a.s.). The total number of particles lost to the sinks on $\hat{\mathbb{Z}}_n$ is $u_n(a_n + 1) + u_n(b_n - 1)$, so the final density is given by

$$\rho'_n(\lambda) = \frac{1}{n-1} \left[\sum_{x=a_n+1}^{b_n-1} \sigma_\lambda(x) - u_n(a_n + 1) - u_n(b_n - 1) \right].$$

By the law of large numbers, $\frac{1}{n-1} \sum \sigma_\lambda(x) \to \lambda$ in probability as $n \to \infty$. Since $u_\infty(x)$ is a.s. finite, we have $\frac{u_n(a_n+1)+u_n(b_n-1)}{n-1} \to 0$ in probability, so $\rho'_n(\lambda) \to \lambda$ in probability.

Next we consider $\lambda \geq 1$. In this case we have $u_n(x) \uparrow u_\infty(x) = \infty$, a.s. Let $p(n,x) = \Pr(u_n(x) = 0)$ be the probability that $x \in \hat{\mathbb{Z}}_n$ does not topple. By the abelian property, adding sinks can not increase the number of topplings, so

$$p(n, x) \leq p(m, 0)$$

where $m = \min(x - a_n, b_n - x)$. Let

$$Y_n = \sum_{x=a_n+1}^{b_n-1} 1_{\{u_n(x)=0\}}$$

be the number of sites in $\hat{\mathbb{Z}}_n$ that do not topple. Since $u_n(0) \uparrow \infty$ a.s., we have $p(n,0) \downarrow 0$, hence

$$\mathbb{E}\frac{Y_n}{n} = \frac{1}{n} \sum_{x=a_n+1}^{b_n-1} p(n,x) \leq \frac{2}{n} \sum_{m=1}^{n/2} p(m,0) \to 0$$

as $n \to \infty$. Since $Y_n \geq 0$ it follows that $Y_n/n \to 0$ in probability.

In an interval where every site toppled, there can be at most one empty site. We have $Y_n + 1$ such intervals. Therefore, the number of empty sites is at most $2Y_n + 1$. Hence

$$\frac{n - 2Y_n - 2}{n-1} \leq \rho'_n(\lambda) \leq 1.$$

The left side tends to 1 in probability, which completes the proof. $\square$

4. Sandpiles on the complete graph

Let K_n be the complete graph on n vertices: every pair of distinct vertices is connected by an edge. In $\hat{K}_n$, one vertex is distinguished as the sink. The maximal stable configuration on $\hat{K}_n$ has density $n-2$, while the minimal recurrent configurations have exactly one vertex of each height $0, 1, \ldots, n-2$, hence density $\frac{n-2}{2}$. The following result shows that the stationary and threshold densities are quite far apart: ζ_s is close to the minimal recurrent density, while ζ_c is close to the maximal stable density.

Theorem 4.1.

$$\zeta_s(\hat{K}_n) = \frac{n}{2} + O(\sqrt{n})$$

$$\zeta_c(K_n) \geq n - O(\sqrt{n \log n}).$$

The proof uses an expression for the stationary density ζ_s in terms of the Tutte polynomial, due to C. Merino López [ML97]. Our application will be to the complete graph, but we state Merino López' theorem in full generality. Let $G = (V, E)$ be a connected undirected graph with n vertices and m edges. Let v be any vertex of G, and write $\hat{G}$ for the graph G with v distinguished as a sink. Let d be the degree of v.

Recall that the Tutte polynomial $T_G(x, y)$ is defined by

$$T_G(x, y) = \sum_{A \subseteq E} (x - 1)^{c(A) - c(E)} (y - 1)^{c(A) + |A| - |V|}$$

where $c(A)$ denotes the number of connected components of the spanning subgraph (V, A).

Theorem 4.2 ([ML97]). *The Tutte polynomial $T_G(x, y)$ evaluated at $x = 1$ is given by*

$$T_G(1, y) = y^{d-m} \sum_\sigma y^{|\sigma|}$$

where the sum is over all recurrent sandpile configurations σ on $\hat{G}$, and $|\sigma|$ denotes the number of particles in σ.

Differentiating and evaluating at $y = 1$, we obtain

$$\frac{d}{dy} T_G(1, y) \bigg|_{y=1} = \sum_\sigma (d - m + |\sigma|). \tag{3}$$

Referring to the definition of the Tutte polynomial, we see that $T_G(1, 1)$ is the number of spanning trees of G, and that the left side of (3) is the number of spanning unicyclic subgraphs of G. (In evaluating T_G at $x = y = 1$, we interpret 0^0 as 1.) The number of recurrent configurations equals the number of spanning trees of G, so the stationary density ζ_s may be expressed as

$$\zeta_s(\hat{G}) = \frac{1}{n T_G(1, 1)} \sum_\sigma |\sigma|.$$

Combining these expressions yields the following:

Corollary 4.3.

$$\zeta_s(\hat{G}) = \frac{1}{n} \left(m - d + \frac{u(G)}{\kappa(G)} \right)$$

where $\kappa(G)$ is the number of spanning trees of G, and $u(G)$ is the number of spanning unicyclic subgraphs of G.

Note that $m - d$ is the minimum number of particles in a recurrent configuration, so the ratio $u(G)/\kappa(G)$ can be interpreted as the average number of excess particles in a recurrent configuration.

Everything so far applies to general connected graphs G. The following is specific for the complete graph.

Theorem 4.4 (Wright [Wri77]). *The number of spanning unicyclic subgraphs of K_n is*

$$u(K_n) = \left(\sqrt{\frac{\pi}{8}} + o(1) \right) n^{n-\frac{1}{2}}.$$

Proof of Theorem 4.1. For $\hat{K}_n$ we have

$$m - d = \frac{n(n-1)}{2} - (n-1) = \frac{(n-2)(n-1)}{2}.$$

From Corollary 4.3, Theorem 4.4, and Cayley's formula $\kappa(K_n) = n^{n-2}$, we obtain

$$\zeta_s(\hat{K}_n) = \frac{1}{n} \left(\frac{(n-2)(n-1)}{2} + \frac{u(K_n)}{\kappa(K_n)} \right)$$

$$= \frac{n}{2} + \left(\sqrt{\frac{\pi}{8}} + o(1) \right) \sqrt{n}.$$

On the other hand, if we let

$$\lambda = n - 2\sqrt{n \log n}$$

and start with $\sigma(v) \sim \text{Poisson}(\lambda)$ particles at each vertex v of K_n, then for all v

$$\Pr[\sigma(v) \geq n] < \frac{1}{n^2}.$$

So

$$\Pr[\sigma(v) \geq n \text{ for some } v] < \frac{1}{n};$$

in other words, with high probability no topplings occur at all. Thus

$$\Pr\left(\Lambda_c(K_n) \geq n - 2\sqrt{n \log n} \right) > 1 - \frac{1}{n}$$

which completes the proof. $\square$

One might guess that the large gap between ζ_s and ζ_c is related to the small diameter of $\hat{K}_n$: since the sink is adjacent to every vertex, its effect is felt with each and every toppling. This intuition is misleading, however, as shown by the lollipop graph $\hat{L}_n$ consisting of K_n connected to a path of length n, with the sink at the far end of the path. Since L_n has the same number of spanning trees and unicyclic subgraphs as K_n, we have by Corollary 4.3

$$\zeta_s(\hat{L}_n) = \frac{1}{2n} \left(\overbrace{\frac{n(n-1)}{2}}^{m} + n - 1 + \frac{u(L_n)}{\kappa(L_n)} \right) = \frac{n}{4} + O(\sqrt{n}).$$

On the other hand, by first stabilizing the vertices on the path, close to half of which end up in the sink without reaching the K_n, it is easy to see that with high probability

$$\Lambda_c(L_n) \geq \frac{2n}{3} - O(\sqrt{n \log n}).$$

5. Sandpiles on the flower graph

An interesting feature of parallel chip-firing is that further phase transitions appear above the threshold density ζ_c. On a finite graph $G = (V, E)$, since the time evolution is deterministic, the system will eventually reach a periodic orbit: for some positive integer m, we have $\eta_{t+m} = \eta_t$ for all sufficiently large t. The *activity density*, ρ_a, measures the proportion of vertices that topple in an average time step:

$$\rho_a(\lambda) = \mathbb{E}_\lambda \lim_{t \to \infty} \frac{1}{t} \sum_{s=0}^{t-1} \frac{1}{\#V} \sum_{x \in V} \mathbf{1}_{\{\eta_s(x) \geq d_x\}}.$$

The expectation $\mathbb{E}_\lambda$ refers to the initial state η_0, which we take to be distributed according to the Poisson product measure with mean λ. Note that the limit in the definition of ρ_a can also be expressed as a finite average, due to the eventual periodicity of the dynamics.

Bagnoli et al. [BCFV03] observed that ρ_a tends to increase with λ in a sequence of flat steps punctuated by sudden jumps. This "devil's staircase" phenomenon is so far explained only on the complete graph [Lev08]: The number of flat stairs increases with n, and in the $n \to \infty$ limit there is a stair at each rational number height $\rho_a = p/q$.

On the cycle $\mathbb{Z}_n$ [Dal06] there are just two jumps: at $\lambda = 1$, the activity density jumps from 0 to $1/2$, and at $\lambda = 2$, from $1/2$ to 1. For the $n \times n$ torus, simulations [BCFV03] indicate a devil's staircase, which is still not completely understood despite much effort [CDVV06].

In this section we study the "flower" graph, which was designed with parallel chip-firing in mind: the idea is that a graph with only short cycles should give rise to short period orbits under the parallel chip-firing dynamics. We find that there are four activity density jumps (Theorem 5.4). In addition, we determine the stationary and threshold densities of the flower graph, and find a second-order phase transition at ζ_c (Theorem 5.5).

The flower graph F_n consists of a central site together with $n \geq 1$ petals (Figure 4). Each petal consists of two sites connected by an edge, each connected to the central site by an edge. Thus the central site has degree $2n$, and all other sites have degree 2. The number of sites is $2n + 1$. The graph $\hat{F}_n$ is the same, except one petal serves as sink.

Recall that we defined the density of a configuration as the total number of particles, divided by the total number of sites. Since the flower graph is not regular, the central site has a different expected number of particles than the petal sites.

Proposition 5.1. *For parallel chip-firing on the flower graph F_n, every configuration has eventual period at most 3.*

The proof uses the following two lemmas.

Lemma 5.2 ([Pri94b] Lemma 2.5). *If the eventual period is not 1, then after some finite time, every site x has height at most $2d_x - 1$.*
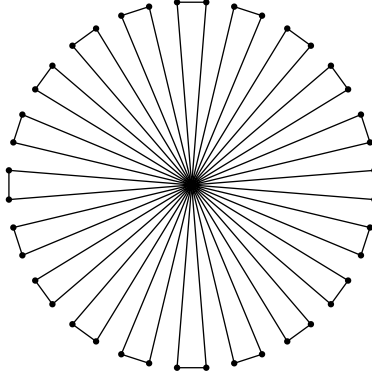
FIGURE 4. The flower graph F_{20}.

We also use an observation from [BCFV03]; it was stated and proved there for $\mathbb{Z}^2$, but the same proof works for general graphs.

Lemma 5.3 ([BCFV03]). *Let two height configurations η and ξ be "mirror images" of each other, that is, $\eta(x) = 2d_x - 1 - \xi(x)$ for all x. Then after performing for each a parallel chip-firing time step, the two configurations are again mirror images of each other.*

Proof of Proposition 5.1. Suppose at time t the model has settled into periodic orbit, and the period is not 1. Then by Lemma 5.2, at time t every petal site has height at most 3. Say that a petal is in state ij if it has i particles at one site and j particles at the other site. *A priori* there are 10 possible petal states, listed below, where each one has two possible successor states, depending on whether or not the central site is stable. If a petal is in state ij, then by S(ij) we denote the state that it is in after one time step in which the central site does not topple, and likewise by U(ij) after one time step in which the central site topples.

state	S(state)	U(state)
00	00	11
01	01	12
02	01	12
03	11	22
11	11	22
12	02	13
13	12	23
22	11	22
23	12	23
33	22	33

From this we see that a petal will be in state 00 only if the central site is always stable, and consequently each site is always stable, in which case the period is 1. Similarly, petal state 33 only occurs if the central site is unstable each step, in which case each site must be unstable each step, and the period is again 1. State 03 is not a successor of any state of

these states, so it will not be a periodic petal state either. Thus the set of allowed periodic petal states is $\{01, 02, 11, 12, 13, 22, 23\}$.

If the central site is stable every other time step, then the possible petal states are $12 \rightarrow 02 \rightarrow 12$, $22 \rightarrow 11 \rightarrow 22$, and $13 \rightarrow 12 \rightarrow 13$, each of which has period 2. Then the period of the entire configuration is 2.

Thus if the period is larger than 2, the central site must be stable for at least two consecutive time steps, or else unstable for at least two consecutive time steps. We will label a time step S if the central site is stable in that time step, otherwise we label it U. So, if the period is larger than 2 we will see SS or UU in the time evolution. In the latter case, we can study the mirror image, which will have the same period, and for which we will see SS.

Eventually the central site must be unstable again, since otherwise the period would be 1. Therefore, we can examine three time steps labeled SSU. Examining the evolution of the central site together with the petals, we see

	S	S	U
01, 02	01	01	12
12, 23	02	01	12
11, 22	11	11	22
23	12	02	12

Whenever we have SSU, during the second and third time steps each petal contributes at most two particles to the central site, while the central site topples, so the central site must again be stable. Thus SSUU cannot occur, and we see SSUS.

There are two cases for what the central site does next. Let us first consider SSUSU.

	S	S	U	S	U
01, 02	01	01	12	02	12
12, 23	02	01	12	02	12
11, 22	11	11	22	11	22
23	12	02	12	02	12

During the last two time steps, each petal contributes exactly 2 particles to the central site, and the central site topples once. Thus after two time steps not only the petals, but also the central site is in the same state. Therefore, the period becomes 2.

Next we consider SSUSS. At this stage each petal is in state 01 or 11, so if there were yet another S, the sandpile would be periodic with period 1. So we see SSUSSU, and because SSUU is forbidden, we conclude that we see SSUSSUS.

	S	S	U	S	S	U	S
01, 02	01	01	12	02	01	12	
12, 23	02	01	12	02	01	12	
11, 22	11	11	22	11	11	22	
23	12	02	12	02	01	12	

At the time of the third S, each petal is in state 12 or 22. Between the third S and the fifth S, each petal contributes exactly two particles to the central site and returns to the same

state, while the central site topples once. Thus the configuration is periodic with period 3. □

We conclude from the above case analysis that the activity ρ_a is always one of 0, 1/3, 1/2, 2/3, or 1. Table 4 summarizes the behavior of the periodic sandpile states for different values of ρ_a.

		periodic sandpile states							
activity ρ_a	0	1/3			1/2		2/3			1
central site	S	S	S	U	S	U	S	U	U	U
petals	01	12	02	01	12	02	23	12	13	≥ 22
	11	22	11	11	22	11	22	11	22	
	00				13	12				

TABLE 4. Behavior of the central site and petals as a function of the activity ρ_a.

The following theorem shows that parallel chip-firing on the flower graph exhibits four distinct phase transitions where the activity ρ_a jumps in value: For each $\alpha \in \{0, \frac{1}{3}, \frac{1}{2}, \frac{2}{3}, 1\}$, there is a nonvanishing interval of initial densities λ where $\rho_a = \alpha$ asymptotically almost surely.

Theorem 5.4. *Let ζ_c be the unique root of $\frac{5}{3} + \frac{1}{3}e^{-3\zeta} = \zeta$, and let ζ_c' be the unique positive root of $\frac{10}{3} - \frac{1}{3}e^{-3\zeta} = \zeta$. (Numerically, $\zeta_c = 1.6688976\ldots$ and $\zeta_c' = 3.3333182\ldots$.) With probability tending to 1 as $n \to \infty$, the activity density ρ_a of parallel chip-firing on the flower graph F_n is given by*

$$\rho_a = \begin{cases} 0, & \text{if } 0 \leq \lambda < \zeta_c \\ 1/3, & \text{if } \zeta_c < \lambda < 2 \\ 1/2, & \text{if } 2 < \lambda < 3 \\ 2/3, & \text{if } 3 < \lambda < \zeta_c' \\ 1 & \text{if } \zeta_c' < \lambda. \end{cases}$$

Proof. In a given petal, let X denote the difference modulo 3 of the number of particles on the two sites of the petal. Observe that X is unaffected by toppling. Let Z denote the number of petals for which $X = 0$, and R denote the total number of particles, in a given initial configuration. Using Table 4, we can relate Z, R, and the activity ρ_a.

When $\rho_a = 0$, we have less than $2n$ particles at the central site, at most two particles for the Z petals of type $X = 0$, and exactly one particle for the other $n - Z$ petals, so $0 \leq R < 2n + 2Z + (n - Z) = 3n + Z$.

When $\rho_a = 1/3$, by considering the U time step, we have $R \geq 2n + 2Z + (n - Z) = 3n + Z$. By considering the preceding S time step, we have $R < 2n + 2Z + 2(n - Z) = 4n$.

When $\rho_a = 1/2$, by considering the U time step, we have $R \geq 4n$, and by considering the S time step, we get $R < 2n + 4Z + 4(n - Z) = 6n$.

When $\rho_a = 2/3$, by considering the second U step, we have $R \geq 2n + 4Z + 4(n - Z) = 6n$. By considering the S time step, we have $R < 2n + 4Z + 5(n - Z) = 7n - Z$.

When $\rho_a = 1$, we have $R \geq 2n + 4Z + 5(n - Z) = 7n - Z$.

Since for given n and Z, these intervals on the values of R are disjoint, we see that the converse statements hold as well: the values of R and Z determine the activity ρ_a. We summarize these bounds:

$$\rho_a = \begin{cases} 0 & \text{if and only if } 0 \leq R < 3n + Z \\ 1/3 & \text{if and only if } 3n + Z \leq R < 4n \\ 1/2 & \text{if and only if } 4n \leq R < 6n \\ 2/3 & \text{if and only if } 6n \leq R < 7n - Z \\ 1 & \text{if and only if } 7n - Z \leq R. \end{cases}$$

Everything so far holds deterministically; next we use probability to estimate R and Z. By the weak law of large numbers, $R/n \to 2\lambda$ and $Z/n \to \Pr(X = 0)$ in probability. Thus, to complete the proof it suffices to show

$$\Pr(X = 0) = \frac{1}{3}(1 + 2e^{-3\lambda}). \tag{4}$$

We can think of building the initial configuration σ_λ by starting with the empty configuration and adding particles in continuous time. Then the value of X for a single petal as particles are added is a continuous time Markov chain on the state space $\{0, \pm 1\}$ with transitions $0 \to \pm 1$ at rate 2, and $\pm 1 \to 0$ and $\pm 1 \to \pm 1$ each at rate 1. Starting in state 0, after running this chain for time λ we obtain

$$\begin{bmatrix} \Pr(X = 0) \\ \Pr(X \neq 0) \end{bmatrix} = \exp\{2\lambda (P - I)\} \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

where $P = \begin{bmatrix} 0 & 1/2 \\ 1 & 1/2 \end{bmatrix}$, and I is the 2×2 identity matrix. The eigenvalues of $P - I$ are 0 and $-\frac{3}{2}$, with corresponding eigenvectors $v_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$ and $v_2 = \begin{bmatrix} 1 \\ -1 \end{bmatrix}$. Since $\begin{bmatrix} 1 \\ 0 \end{bmatrix} = \frac{1}{3}v_1 + \frac{2}{3}v_2$, we obtain (4). $\qquad\square$

The following theorem describes a phase transition in the driven sandpile dynamics on the flower graph analogous to Theorem 3.1 for the bracelet graph. We remark on one interesting difference between the two transitions: for $\lambda > \zeta_c$, the final density $\rho(\lambda)$ is increasing in λ for the bracelet, and decreasing in λ for the flower graph.

For $\lambda > 0$, let σ_λ be the configuration with $\text{Poisson}(\lambda)$ particles independently on each site of $\hat{F}_n$. Let $\eta_\lambda = (\sigma_\lambda)^\circ$ be the stabilization of σ_λ, and let

$$\rho_n(\lambda) = \frac{1}{n - 1} \sum_{x=1}^{n-1} \eta_\lambda(x)$$

be the final density.

Theorem 5.5. *For the flower graph with n petals, in the limit $n \to \infty$ we have*
  (1) *The threshold density ζ_c is the unique positive root of $\zeta = \frac{5}{3} + \frac{1}{3}e^{-3\zeta}$.*
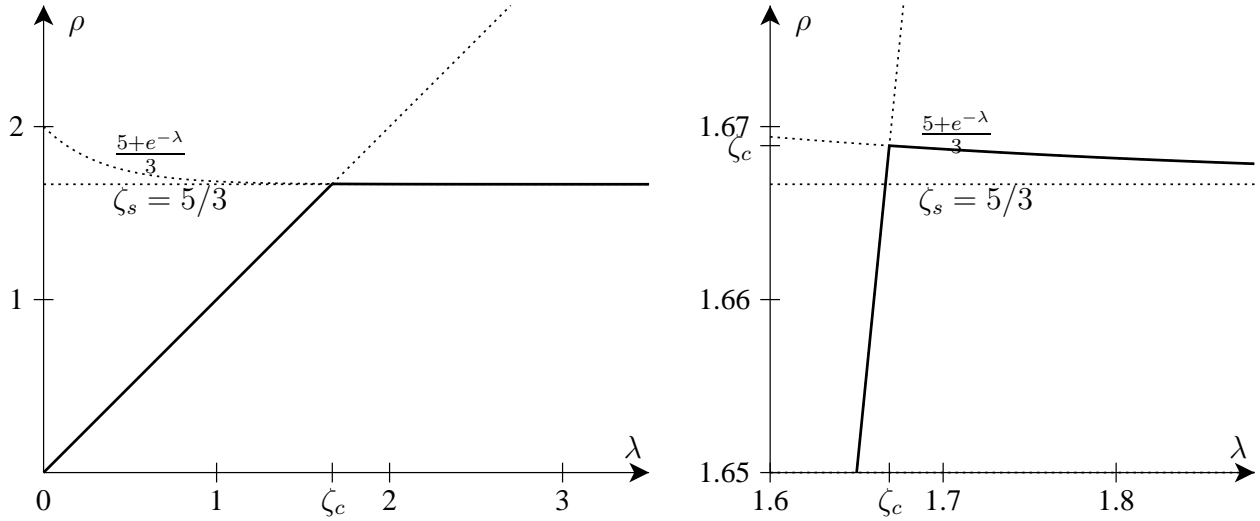 (2) *The stationary density ζ_s is $5/3$.*

FIGURE 5. Density $\rho(\lambda)$ of the final stable configuration as a function of initial density λ on the flower graph $\hat{F}_n$ for large n. A second-order phase transition occurs at $\lambda = \zeta_c$. Beyond this transition, the density of the driven sandpile decreases with λ.

(3) $\rho_n(\lambda) \to \rho(\lambda)$ *in probability, where*

$$\rho(\lambda) = \min\left(\lambda, \ \frac{5}{3} + \frac{1}{3}e^{-\lambda}\right) = \begin{cases} \lambda, & \lambda \le \zeta_c \\ \frac{5}{3} + \frac{1}{3}e^{-3\lambda}, & \lambda > \zeta_c. \end{cases}$$

Proof. Part 1 follows from Theorem 5.4.

For Part 2, we use the burning algorithm. In all recurrent configurations on $\hat{F}_n$, the central site has either $2n-1$ or $2n-2$ particles. All other sites have at most one particle, and in each petal (except the sink) there is at least one particle. For each petal that is not the sink, there are two possible configurations with 1 particle, and one with 2 particles. Each of these occurs with equal probability in the stationary state, so the expected number of particles in the petals is $(n-1)(\frac{2}{3} \cdot 1 + \frac{1}{3} \cdot 2) = \frac{4n}{3} + O(1)$ as $n \to \infty$. Therefore, the total density is $\zeta_s = \lim_{n\to\infty} \frac{2n+4n/3}{2n-1} + o(1) = 5/3$.

For part 3, for the driven dissipative sandpile on $\hat{F}_n$, we first stabilize all the petals, then topple the center site if it is unstable, then stabilize all the petals, and so on. For each toppling of the center site, the sandpile loses $O(1)$ particles to the sink. If the center topples at least once, then each petal will be in one of the states 11, 01, or 10, after which the number of particles at the center site is $R - n - Z + O(1)$. Recall from the proof of Theorem 5.4 that $R/n \to 2\lambda$ and $Z/n \to \frac{1}{3}(1 + 2e^{-3\lambda})$ in probability. Thus if $\lambda \le \zeta_c$, then $\frac{R-n+1-Z}{2n} \to \lambda - \frac{2}{3} + \frac{1}{3}e^{-3\lambda} \le 1$ in probability, so the sandpile does not lose a macroscopic amount of sand, and $\rho_n(\lambda) \to \lambda$ in probability.

If $\lambda > \zeta_c$, then the number of particles that remain after stabilization is $2n+n+Z+O(1)$. In this case, we have $\rho_n(\lambda) = \frac{3n+Z}{2n+1} + o(1) \to \frac{5}{3} + \frac{1}{3}e^{-3\lambda}$ in probability. $\square$

6. Sandpiles on the Cayley tree

Dhar and Majumdar [DM90] studied the abelian sandpile model on the Cayley tree (also called the Bethe lattice) with branching factor q, which has degree $q + 1$. Implicit in their formulation is that they used wired boundary conditions, i.e., where all the vertices of the tree at a certain large distance from a central vertex are glued together and become the sink. (The other common boundary condition is free boundary conditions, where all the vertices at a certain distance from the central vertex become leaves, and one of them becomes the sink. The issue of boundary conditions becomes important for trees, because in any finite subgraph, a constant fraction of vertices are on the boundary. This is in contrast to $\mathbb{Z}^2$, where free and wired boundary conditions lead to the same infinite volume limit. See [LP10].)

The finite regular wired tree $T_{q,n}$ is the ball of radius n in the infinite $(q + 1)$-regular tree, with all leaves collapsed to a single vertex s. In $\hat{T}_{q,n}$ the vertex s serves as the sink. Maes, Redig and Saada [MRS02] show that the stationary measure on recurrent sandpiles on $\hat{T}_{q,n}$ has an infinite-volume limit, which is a measure on sandpiles on the infinite tree. Denoting this measure by Pr_q, if h denotes the number of particles at a single site far from the boundary, then we have [DM90]

$$\mathrm{Pr}_q[h = i] = \frac{1}{(q^2 - 1)q^q} \sum_{m=0}^{i} \binom{q + 1}{m} (q - 1)^{q+1-m}.$$

From this formula we see that the stationary density is

$$\zeta_s = \mathbb{E}_q[h] = \frac{q + 1}{2}.$$

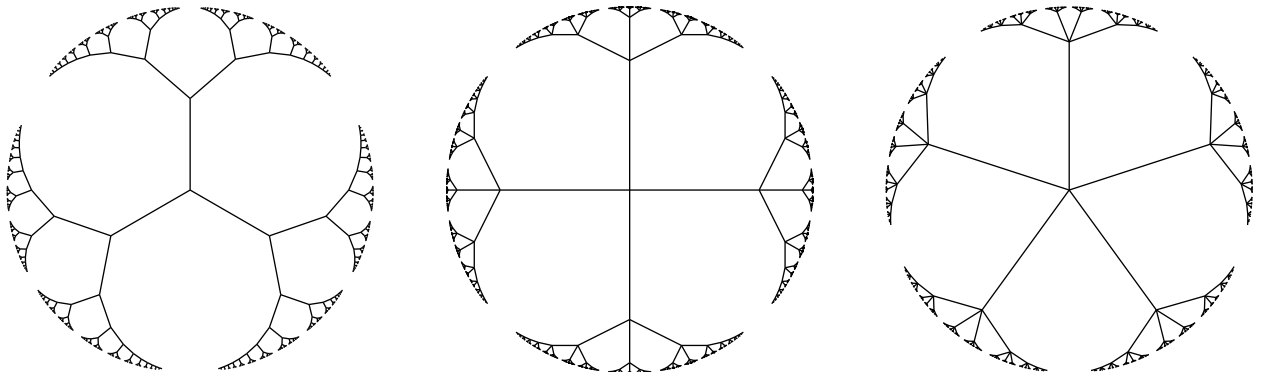For 3-regular, 4-regular, and 5-regular trees, these values are summarized below:

FIGURE 6. The Cayley trees (Bethe lattices) of degree $d = 3, 4, 5$.

tree		$\mathbb{E}_q[h]$	distribution of height h of sand				
q	degree		$\Pr_q[h=0]$	$\Pr_q[h=1]$	$\Pr_q[h=2]$	$\Pr_q[h=3]$	$\Pr_q[h=4]$
2	3	3/2	1/12	4/12	7/12		
3	4	2	2/27	2/9	1/3	10/27	
4	5	5/2	81/1280	27/160	153/640	21/80	341/1280

Large-scale simulations on $T_{q,n}$ are rather impractical because the vast majority of vertices are near the boundary. Consequently, each simulation run produces only a small amount of usable data from vertices near the center.

To experimentally measure ζ_c for the Cayley trees, we generated large random regular graphs $G_{q,n}$, and used these as finite approximations of the infinite Cayley tree. We used the following procedure to generate random connected bipartite multigraphs of degree $q+1$ on n vertices (n even). Let M_0 be the set of edges $(i, i+1)$ for $i = 1, 3, 5, \ldots, n-1$. Then take the union of M_0 with q additional i.i.d. perfect matchings $M_1, \ldots, M_q$ between odd and even vertices. Each M_j is chosen uniformly among all odd-even perfect matchings whose union with M_0 is an n-cycle.

Most vertices of G_n will not be contained in any cycle smaller than $\log_q n + O(1)$ (see e.g., [Bol86]), so these graphs are locally tree-like. For this reason, we believe that as $n \to \infty$ the threshold density $\Lambda_c(G_n)$ will be concentrated at the threshold density of the infinite tree.

Since the choice of multigraph affects the estimate of ζ_c, we generated a new independent random multigraph for each trial. The results for random regular graphs of degree 3, 4 and 5 are summarized in Tables 5, 6, and 7. We find that for the 5-regular tree, the threshold density is about 2.511 rather than 2.5, for the 4-regular tree the threshold density is very close to but decidedly larger than 2, while for the 3-regular tree the threshold density is extremely close to 1.5, with a discrepancy that we were unable to measure. However, for the 3-regular tree there is a measurable discrepancy (about 2×10^{-6}) in the probability that a site has no particles.

n	#samples	$\mathbb{E}[h]$	distribution of height h of sand			#topplings
			$\Pr[h=0]$	$\Pr[h=1]$	$\Pr[h=2]$	$\div n \log^{1/2} n$
1048576	2097152	1.5004315	0.0833326	0.332903	0.583764	1.263145
2097152	1048576	1.5003054	0.0833321	0.333031	0.583637	1.258046
4194304	524288	1.5002161	0.0833314	0.333121	0.583548	1.253092
8388608	262144	1.5001528	0.0833311	0.333185	0.583484	1.247642
16777216	131072	1.5001081	0.0833311	0.333230	0.583439	1.242359
33554432	65536	1.5000765	0.0833307	0.333262	0.583407	1.237317
67108864	32768	1.5000540	0.0833307	0.333285	0.583385	1.232398
134217728	16384	1.5000382	0.0833307	0.333300	0.583369	1.227548
268435456	8192	1.5000269	0.0833308	0.333311	0.583358	1.222371
536870912	4096	1.5000191	0.0833308	0.333319	0.583350	1.214431
1073741824	2048	1.5000136	0.0833307	0.333325	0.583344	1.212751
∞ (stationary)		1.5	0.0833333	0.333333	0.583333	

TABLE 5. Data for the fixed-energy sandpile on a pseudorandom 3-regular graph on n nodes. Each estimate of $\mathbb{E}[h]$ has standard deviation less than $7 \cdot 10^{-8}$, and each estimate of the marginals $\Pr[h=i]$ has standard deviation less than $3 \cdot 10^{-7}$. The data for $\mathbb{E}[h]$ appears to fit $3/2 + \text{const}/\sqrt{n}$ very well, and extrapolating to $n \to \infty$ it appears that $\mathbb{E}[h] \to 1.500000$ to six decimal places. However, apparently $\Pr[h=0] \to 0.083331 < 1/12$.

n	#samples	$\mathbb{E}[h]$	distribution of height h of sand				#topplings
			$\Pr[h=0]$	$\Pr[h=1]$	$\Pr[h=2]$	$\Pr[h=3]$	$\div n \log^{1/2} n$
1048576	2097152	2.001109	0.073884	0.221887	0.333466	0.370763	0.623322
2097152	1048576	2.000853	0.073881	0.221978	0.333547	0.370593	0.618848
4194304	524288	2.000688	0.073880	0.222037	0.333599	0.370484	0.620894
8388608	262144	2.000584	0.073878	0.222075	0.333631	0.370416	0.631324
16777216	131072	2.000518	0.073877	0.222100	0.333651	0.370372	0.649328
33554432	65536	2.000477	0.073877	0.222114	0.333664	0.370345	0.670838
67108864	32768	2.000451	0.073877	0.222123	0.333673	0.370328	0.691040
134217728	16384	2.000434	0.073876	0.222130	0.333678	0.370316	0.699706
268435456	8192	2.000424	0.073876	0.222134	0.333681	0.370310	0.695065
536870912	4096	2.000417	0.073876	0.222136	0.333683	0.370305	0.684507
1073741824	2048	2.000413	0.073876	0.222138	0.333684	0.370303	0.673061
∞ (stationary)		2	0.074074	0.222222	0.333333	0.370370	

TABLE 6. Data for the fixed-energy sandpile on a pseudorandom 4-regular graph on n nodes. Each estimate of $\mathbb{E}[h]$ and of the marginals $\Pr[h=i]$ has standard deviation less than $3 \cdot 10^{-7}$.

n	#samples	$\mathbb{E}[h]$	distribution of height h of sand					#topplings
			$\Pr[h=0]$	$\Pr[h=1]$	$\Pr[h=2]$	$\Pr[h=3]$	$\Pr[h=4]$	$\div n$
1048576	1048576	2.512106	0.062271	0.166547	0.237230	0.264711	0.269242	1.666086
2097152	524288	2.511947	0.062269	0.166579	0.237256	0.264727	0.269169	1.666244
4194304	262144	2.511847	0.062268	0.166599	0.237272	0.264737	0.269123	1.666404
8388608	131072	2.511781	0.062267	0.166613	0.237283	0.264743	0.269093	1.666589
16777216	65536	2.511743	0.062267	0.166621	0.237289	0.264748	0.269075	1.667322
33554432	65536	2.511716	0.062267	0.166627	0.237293	0.264750	0.269063	1.668196
67108864	32768	2.511700	0.062267	0.166630	0.237296	0.264752	0.269056	1.669392
134217728	16384	2.511689	0.062267	0.166632	0.237297	0.264755	0.269050	1.671613
268435456	8192	2.511683	0.062266	0.166634	0.237299	0.264753	0.269048	1.675479
536870912	4096	2.511680	0.062267	0.166633	0.237300	0.264755	0.269045	1.677092
1073741824	2048	2.511677	0.062266	0.166634	0.237300	0.264755	0.269044	1.688093
∞ (stationary)		2.5	0.063281	0.168750	0.239063	0.262500	0.266406	

TABLE 7. Data for the fixed-energy sandpile on a pseudorandom 5-regular graph on n nodes. Each estimate of $\mathbb{E}[h]$ and of the marginals $\Pr[h=i]$ has standard deviation less than $2 \cdot 10^{-6}$.

7. Sandpiles on the ladder graph

The examples in previous sections suggest that the density conjecture can fail for (at least) two distinct reasons: local toppling invariants, and boundary effects. A *toppling invariant* for a graph G is a function f defined on sandpile configurations on G which is unchanged by performing topplings; that is

$$f(\sigma) = f(\sigma + \Delta_x)$$

for any sandpile σ and any column vector Δ_x of the Laplacian of G. Examples we have seen are

$$f(\sigma) = \sigma(x) \bmod 2$$

where x is any vertex of the bracelet graph B_n; and

$$f(\sigma) = \sigma(x_1) - \sigma(x_2) \bmod 3$$

where x_1, x_2 are the two vertices comprising any petal on the flower graph F_n. Both of these toppling invariants are *local* in the sense that they depend only on a bounded number of vertices as $n \to \infty$.

The Cayley tree has no local toppling invariants, but the large number of sinks, comparable to the total number of vertices, produce a large boundary effect. The density conjecture fails even more dramatically on the complete graph (Theorem 4.1). One might guess that this is due to the high degree of interconnectedness, which causes boundary effects from the sink to persist as $n \to \infty$. A good candidate for a graph G satisfying the density conjecture, then, should have

- no local toppling invariants,
- most vertices far from the sink.

The best candidate graphs G should be essentially one-dimensional, so that the sink is well insulated from the bulk of the graph, keeping boundary effects to a minimum. Indeed, the only graph known to satisfy the density conjecture is the infinite path $\mathbb{Z}$.

Járai and Lyons [JL07] study sandpiles on graphs of the form $G \times P_n$, where G is a finite connected graph and P_n is the path of length n, with the endpoints serving as sinks. The simplest such graphs that are not paths are obtained when $G = P_1$ has two vertices and one edge. These graphs are a good candidate for $\zeta_c = \zeta_s$, for the reasons described above. Nevertheless, we find that while ζ_c and ζ_s are very close, they appear to be different.

First we calculate ζ_s. Jarai and Lyons [JL07, section 5] define recurrent configurations as Markov chains on the state space

$$X = \left\{ (3,3), (3,2), (2,3), (3,1), (1,3), \overline{(3,2)}, \overline{(2,3)} \right\}$$

FIGURE 7. The ladder graph.

describing the possible transitions from one rung of the ladder to the next. States (i, j) and $\overline{(i, j)}$ both represent rungs whose left vertex has $i - 1$ particles and whose right vertex has $j - 1$ particles. The distinction between states $(3, 2)$ and $\overline{(3, 2)}$ lies only in which transitions are allowed. The adjacency matrix describing the allowable transitions is given by

$$A = \begin{pmatrix} 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 1 & 1 & 1 & 1 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 1 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix}.$$

Its largest eigenvalue is $2 + \sqrt{3}$, and the corresponding left and right eigenvectors are

$$u = (1 + \sqrt{3}, 1 + \sqrt{3}, 1 + \sqrt{3}, 1, 1, 1, 1)$$
$$v = (3 + \sqrt{3}, 1 + \sqrt{3}, 1 + \sqrt{3}, 1 + \sqrt{3}, 1 + \sqrt{3}, 1, 1)^T$$

By the Parry formula [Par64], the stationary probabilities are given by $p(i) = u_i v_i / Z$, where Z is a normalizing constant. So

$$p(3, 3) = (1 + \sqrt{3})(3 + \sqrt{3})/Z$$
$$p(2, 3) = p(3, 2) = (1 + \sqrt{3})^2/Z$$
$$p(1, 3) = p(3, 1) = (1 + \sqrt{3})/Z$$
$$p(\overline{2, 3}) = p(\overline{3, 2}) = 1/Z$$

where

$$Z = (1 + \sqrt{3})(3 + \sqrt{3}) + 2(1 + \sqrt{3})^2 + 2(1 + \sqrt{3}) + 2.$$

Thus we find that for the ladder graph in stationarity, the number h of particles at a site satisfies

$$\Pr[h = 0] = -\tfrac{1}{2} + \tfrac{\sqrt{3}}{3} = 0.0773503\ldots$$
$$\Pr[h = 1] = \tfrac{5}{4} - \tfrac{7\sqrt{3}}{12} = 0.2396370\ldots$$
$$\Pr[h = 2] = \tfrac{1}{4} + \tfrac{\sqrt{3}}{4} = 0.6830127\ldots$$
$$\zeta_s = \mathbb{E}[h] = \tfrac{7}{4} - \tfrac{\sqrt{3}}{12} = 1.60566243\ldots$$

In contrast, the threshold density for ladders appears to be about 1.6082. Table 8 summarizes simulation data on finite $2 \times n$ ladders.

n	#samples	$\mathbb{E}[h]$	distribution of height h of sand			#topplings $\div n^{5/2}$
			$\Pr[h=0]$	$\Pr[h=1]$	$\Pr[h=2]$	
256	4194304	1.60567	0.07695	0.24043	0.68262	0.094773
512	2097152	1.60693	0.07656	0.23996	0.68349	0.095366
1024	1048576	1.60757	0.07636	0.23970	0.68393	0.095864
2048	524288	1.60788	0.07626	0.23960	0.68414	0.096316
4096	262144	1.60805	0.07621	0.23952	0.68426	0.096545
8192	131072	1.60814	0.07618	0.23950	0.68432	0.096753
16384	65536	1.60816	0.07618	0.23949	0.68434	0.097113
32768	32768	1.60818	0.07617	0.23948	0.68435	0.096944
65536	16384	1.60820	0.07616	0.23948	0.68436	0.097342
131072	8192	1.60820	0.07617	0.23946	0.68437	0.097648
262144	4096	1.60821	0.07615	0.23949	0.68436	0.096158
∞ (stationary)		1.60566	0.07735	0.23964	0.68301	

TABLE 8. Data for the fixed-energy sandpile on $2 \times n$ ladder graphs. Each estimate of $\mathbb{E}[h]$ and of the marginals $\Pr[h = i]$ has a standard deviation smaller than 10^{-5}. To four decimal places, the threshold density ζ_c equals 1.6082, which exceeds the stationary density $\zeta_s = 7/4 - \sqrt{3}/12 = 1.6057$. The total number of topplings appears to scale as $n^{5/2}$.

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