# Universality and Sharpness in Absorbing-State Phase Transitions 

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#### Abstract

We consider the Activated Random Walk model in any dimension with any sleep rate and jump distribution, and show that the stabilization properties depend only on the average density of particles, regardless of how they are initially located on the lattice.


## 1 Introduction

The concept of self-organized criticality was introduced in the late 80's to explain the emergence of critical behavior in steady states without fine tuning of system parameters [2]. Intrinsic relationships between this phenomenon and that of ordinary phase transition started to unfold in the late 90 's with a new paradigm: the self-organized critical behavior of a driven-dissipative system is related to ordinary criticality of a corresponding fixed-energy system that uses the same relaxation mechanism [6]. A central issue is the density conjecture: the typical density $\zeta_{s}$ in the steady state of the driven-dissipative system arguably coincides with the threshold density $\zeta_{c}$ of the fixed-energy system. Ten years later, the density conjecture was shown to be false [8] if the relaxation mechanism is the Abelian Sandpile (ASM) on $\mathbb{Z}^{d}$ for $d \geqslant 2$. This property is attributed to the fact that the ASM dynamics does not wipe out details of the initial condition by the time a configuration becomes explosive [11].

Some properties of the ASM that indicate bad mixing behavior had already appeared in the literature. First of all, $\zeta_{c}$ is defined as the threshold density at which the fixed-energy model becomes unstable, under the implicit assumption of i.i.d. Poisson distributions. Otherwise, the notion of threshold density is not even well-defined. Indeed, it turns out that for every $d<\zeta<2 d-1$ there are some spatially ergodic states with average particle density $\zeta$ that are explosive and some others with same density which are stabilizable [10].

Stochastic Sandpiles (SSM) and Activated Random Walks (ARW) were introduced as alternative nondeterministic relaxation mechanisms. The long-range space-time correlations caused by conservation of particles and the lack of an algebraic structure similar to the ASM make the mathematical analysis of these models very challenging. It took two decades for the first rigorous results regarding stability properties of these systems to appear in the literature [7, 16]. Considerable progress for the ARW has been made in the past three years $[3,4,5,17,19,20,21]$, with the introduction of a number of ad hoc techniques and tools. Some of these tools were sensitive to assumptions about the initial state, requiring independence, light tails, etc.

In this paper we study the ARW and prove that the critical density $\zeta_{c}$ is well-defined and separates two entire families of spatially ergodic states: those whose density $\zeta$ is below $\zeta_{c}$, for which configurations are a.s. stabilizable, and those having density above $\zeta_{c}$, for which configurations are a.s. explosive. One consequence is that all recently-introduced mathematical techniques having special requirements for the initial state are immediately bootstrapped to general results valid for every ergodic distribution. Another consequence is to support the general belief that the ARW has much better mixing properties than the

ASM. The result and technique introduced in this paper do not yield a proof of the density conjecture for the ARW, but may give one step in that direction.

Sharpness and self-organized criticality. Models of avalanches became a standard example of selforganized criticality in the context of non-equilibrium steady states. Unlike usual statistical mechanics systems, these models are not explicitly equipped with a tuning parameter at which a phase transition is observed. Instead, they are expected to spontaneously drive themselves to a critical steady state, featuring characteristics of critical systems such as power law statistics and scale invariance.

Three different relaxation procedures have been used in the study of this phenomenon: in the deterministic sandpile model, sites with at least $2 d$ particles send one to each neighbor; in the stochastic sandpile, sites with at least $N$ particles send $N$ particles to neighbors chosen at random; in the activated random walk model, sites with active particles send one particle to a neighbor chosen at random, and particles can become passive with probability proportional to a parameter $\lambda$ if they are alone. All these systems contain mechanisms that cause both spread of activity and a tendency of this activity to die out, and the system behavior is determined by the balance between these two factors.

Self-organized criticality appears in the corresponding driven-dissipative dynamics: particles are added to the bulk of a large finite box, and absorbed at its boundary during relaxation, following one of the above-mentioned mechanisms. A particle is added only after the system globally stabilizes. In this dynamics, when the average density $\zeta$ inside the box is too small, mass tends to accumulate. When it is too large, there is intense activity and substantial dissipation at the boundary. Within this setting, the system is attracted to a critical state with an average density $\zeta_{s}$.

A new paradigm was introduced in [6], arguing that self-organized criticality in these systems is related to ordinary phase transition. More precisely, the corresponding conservative systems in infinite volume, where the density $\zeta$ is kept constant, exhibit ordinary phase transition and their critical behavior is closely related to properties of the self-organized critical system described above. In particular, there is a threshold density $\zeta_{c}$ such that the infinite-volume dynamics should fixate for $\zeta<\zeta_{c}$ and remain active for $\zeta>\zeta_{c}$, and $\zeta_{c}$ should coincide with the driven-dissipative stationary density $\zeta_{s}$. Since then, a rich literature appeared, exploring this relation and the principles behind it.

Years later, rigorous results and precise large-scale simulation showed that equality $\zeta_{c}=\zeta_{s}$ is false in general, at least for the deterministic sandpile model [8,9], indicating that the relation between drivendissipative and conservative systems is much more subtle. This is attributed to the fact that the ASM is very sensitive to the initial state [11], and indeed a robust definition of threshold density for the ASM is not even possible, as previously shown in [10].

This discovery had two implications. First, it increased interest in the mathematical properties of this system and its intricate behavior, and propelled debate about how to recover the density conjecture for the ASM $[12,14]$. Second, it increased physical interest in other models such as SSM and ARW, which are supposed to have stronger mixing properties, for the study of self-organized criticality and analysis of avalanche statistics. We show that $\zeta_{c}$ is well-defined for the ARW, providing some support for the latter claim. We now state the main result of this paper, postponing precise definitions to $\S 2$.

Theorem 1. Consider the Activated Random Walk model on the usual graph $\mathbb{Z}^{d}$ for fixed $d \geqslant 1$ with given sleep rate $\lambda$, and given jump distribution $p(\cdot)$. There is a number $\zeta_{c}$ such that, for any spatially ergodic distribution $\nu$ supported on active configurations with average density $\zeta$, a configuration sampled from $\nu$ is a.s. stabilizable if $\zeta<\zeta_{c}$ and a.s. explosive if $\zeta>\zeta_{c}$.

In the sequel we situate the above result in the context of existing mathematical literature, and finally discuss the method of proof and organization of the paper.

Mathematical overview. Several non-trivial bounds for $\zeta_{c}$ were proved in the past three years. For $d=1$, it was proved in [16] that $\zeta_{c}>0$ for all $\lambda$ and $\zeta_{c} \rightarrow 1$ as $\lambda \rightarrow \infty$. For $d \geqslant 2$ and $\lambda=\infty$, it was also shown in [18] that $\mu_{c}>0$ and in $[4,5]$ that $\mu_{c} \geqslant 1$. For $d \geqslant 2$ and $\lambda>0$ it was shown in [19] that $\zeta_{c}>0$,
assuming short-range unbiased jump distributions. This was generalized to general jump distributions in [20], where it was also shown that $\zeta_{c} \rightarrow 1$ as $\lambda \rightarrow \infty$. It was proved in [1, 18] that $\zeta_{c} \leqslant 1$ in any dimension for any $\lambda$. For biased jump distributions, it was shown in [21] that, on $d=1, \zeta_{c}<1$ for every $\lambda$ and $\zeta_{c} \rightarrow 0$ as $\lambda \rightarrow 0$, and on $d \geqslant 2$ that $\zeta_{c}<1$ for small $\lambda$. The picture on $d \geqslant 2$ was extended in [17] by showing show that $\zeta_{c}<1$ for every $\lambda$ and $\zeta_{c} \rightarrow 0$ as $\lambda \rightarrow 0$. For unbiased jumps, it was shown that $\zeta_{c} \rightarrow 0$ as $\lambda \rightarrow 0$, in [3] for $d=1$ and [20] for $d \geqslant 3$. See [15] for a detailed account.

While many of the proofs were robust with respect to the initial state, others were sensitive to it. Some results in $[1,4,5,20,21]$ required an i.i.d. field, the proofs in [19] are presented for i.i.d. Poisson (but may be adapted for states with some spatial mixing and light tails), and some of the results in [20] and [21] required the initial state to be i.i.d. Bernoulli. With Theorem 1, all these conditions can be waived, and arguments having special requests for the initial state now produce lower and upper bounds valid for any other ergodic initial state. A good example is the case of biased walks in dimension $d \geqslant 2$. The proof in [21] that $\zeta_{c}<1$ for all $\lambda>0$ and i.i.d. Bernoulli initial state was extended to an arbitrary i.i.d. field in [17], however such an extension now follows directly from this general property.

Comments on the proof and outline of the paper. The idea of the proof is simple, although its implementation has some subtleties. It uses the characterization of stabilizability in terms of the Diaconis-Fulton representation, and consists of two stages: first embedding one infinite configuration into another one with higher density (which is assumed to be fixating) by means of enforced activation, and then using monotonicity to argue that the embedded configuration is also fixating. The first stage relies on ergodic properties and on the mass-transport principle.

In $\S 2$, we recall usual tools for this system and fix the notation. We quickly describe a continuous-time evolution and the event of fixation, the Diaconis-Fulton representation and the notion of stabilization with its main properties, the relation between stabilization and fixation, and finally we state the masstransport principle. In $\S 3$, we state a theorem equivalent to Theorem 1, briefly sketch its proof, and then give the proof in three parts: embedding the initial configuration into another one with higher density, stabilization of the embedded configuration, and finally stabilization of the original configuration.

## 2 Definitions and tools

To be consistent with most of the existing literature, we describe the ARW evolution as a continuous-time stochastic process. The ARW system starts with active particles placed in $\mathbb{Z}^{d}$ according to a distribution $\nu$, and evolves as follows. Active particles perform independent continuous-time random walks on $\mathbb{Z}^{d}$ with translation-invariant jump distribution $p(x, y)=p(y-x)$, and switch to passive state at rate $\lambda>0$ when they are alone on a site. Passive particles do not move, and are reactivated immediately when visited by another particle. The law of this evolution is denoted $\mathbb{P}^{\nu}$. By fixation we mean that the dynamics eventually halts at any finite region, and non-fixation is the opposite event.

In the sequel we introduce notation for this continuous-time evolution, and briefly describe the toppling operators and their properties, referring the reader to [15] for details. One important difference in notation between here and [15] is that here we do not necessarily start from a zeroed odometer. One difference between the terminology of [15] and that of $[16,19]$ is that we do not replace sleep instructions by neutral ones, but instead allow for the toppling of a site containing a sleepy particle.

Notation and continuous-time evolution. Let $\mathbb{N}_{0}=\{0,1,2, \ldots\}$ and $\mathbb{N}_{\mathfrak{s}}=\mathbb{N}_{0} \cup\{\mathfrak{s}\}$ with $0<\mathfrak{s}<$ $1<2<\cdots$. Define $|\mathfrak{s}|=1, \llbracket \mathfrak{s} \rrbracket=0$, and $|n|=\llbracket n \rrbracket=n$ for $n \in \mathbb{N}_{0}$. Also define $\mathfrak{s}+1=2$ and

$$
n \cdot \mathfrak{s}= \begin{cases}n, & n \geqslant 2 \\ \mathfrak{s}, & n=1 \\ \text { undefined, }, & n=0\end{cases}
$$

The state of the ARW at time $t \geqslant 0$ is given by $\eta_{t} \in \Sigma=\left(\mathbb{N}_{\mathfrak{s}}\right)^{\mathbb{Z}^{d}}$, and the process evolves as follows. For each site $x$, a Poisson clock rings at rate $(1+\lambda) \llbracket \eta_{t}(x) \rrbracket$. When this clock rings, the system goes through the transition $\eta \rightarrow \tau_{x s} \eta$ with probability $\frac{\lambda}{1+\lambda}$, otherwise $\eta \rightarrow \tau_{x y} \eta$ with probability $p(y-x) \frac{1}{1+\lambda}$. The transitions are given by

$$
\tau_{x y} \eta(z)=\left\{\begin{array}{ll}
\eta(x)-1, & z=x, \\
\eta(y)+1, & z=y, \\
\eta(z), & \text { otherwise }
\end{array} \quad \tau_{x \mathfrak{s}} \eta(z)= \begin{cases}\eta(x) \cdot \mathfrak{s}, & z=x \\
\eta(z), & \text { otherwise }\end{cases}\right.
$$

We assume that $\eta_{0}(x) \in \mathbb{N}_{0}$ for all $x$ a.s., and use $\mathbb{P}^{\nu}$ to denote the law of $\left(\eta_{t}\right)_{t \geqslant 0}$, where $\nu$ denotes the distribution of $\eta_{0}$. We say that $\left(\eta_{t}\right)_{t \geqslant 0}$ fixates if $\eta_{t}(x)$ is eventually constant for each fixed $x \in \mathbb{Z}^{d}$.

Diaconis-Fulton and stabilization. We now use $\eta$ to denote configurations in $\Sigma$ instead of a continuous-time process. We say that site $x$ is unstable for the configuration $\eta$ if $\eta(x) \geqslant 1$. Otherwise, $x$ is said to be stable. By toppling site $x$ we mean the application of an operator $\tau_{x y}$ or $\tau_{x s}$ to $\eta$. Toppling an unstable site is legal. If $\eta(x)=\mathfrak{s}$, toppling $x$ is not legal but is acceptable (here we are departing from the original dynamics of the model, but this operation is useful in the proofs), and to that end we define $\mathfrak{s}-1=0$ and $\mathfrak{s} \cdot \mathfrak{s}=\mathfrak{s}$. Legal topplings are acceptable. If $\eta(x)=0$, toppling $x$ is not acceptable.

Let $\mathcal{I}=\left(\tau^{x, j}\right)_{x \in \mathbb{Z}^{d}, j \in \mathbb{N}}$ be a fixed field of instructions, that is, for each $x$ and $j, \tau^{x, j}$ equals $\tau_{x \mathfrak{s}}$ or $\tau_{x y}$ for some $y$. Let $h \in\left(\mathbb{N}_{0}\right)^{\mathbb{Z}^{d}}$. The toppling operation at $x$ is defined by $\Phi_{x}(\eta, h)=\left(\tau^{x, h(x)+1} \eta, h+\delta_{x}\right)$. Given a finite sequence $\alpha=\left(x_{1}, \ldots, x_{k}\right)$, define $\Phi_{\alpha}=\Phi_{x_{k}} \circ \Phi_{x_{k-1}} \circ \cdots \circ \Phi_{x_{1}}$. We say that $\alpha$ is a legal/acceptable sequence of topplings for $(\eta, h)$ if, for every $j=1, \ldots, k, \Phi_{x_{j}}$ is legal/acceptable for $\Phi_{x_{1}, \ldots, x_{j-1}}(\eta, h)$. Given $V \subseteq \mathbb{Z}^{d}$, we say that $(\eta, h)$ is stable in $V$ if every $x \in V$ is stable for $\eta$. We write $\alpha \subseteq V$ if $x_{1}, \ldots, x_{k} \in V$. We say that $\alpha$ stabilizes $(\eta, h)$ in $V$ if $\alpha$ is acceptable for $(\eta, h)$ and $\Phi_{\alpha}(\eta, h)$ is stable in $V$. Let $m_{\alpha}$ be given by $m_{\alpha}(x)=\sum_{\ell} \mathbb{1}_{\left\{x_{\ell}=x\right\}}$, for all $x \in \mathbb{Z}^{d}$. We write $m_{\beta} \leqslant m_{\alpha}$ if $m_{\beta}(x) \leqslant m_{\alpha}(x)$ for all $x \in \mathbb{Z}^{d}$, and the same for $\eta \leqslant \tilde{\eta}$.
Lemma 2 (Local abelianness). If $\alpha$ and $\beta$ are acceptable sequences of topplings for the configuration $(\eta, h)$, such that $m_{\alpha}=m_{\beta}$, then $\Phi_{\alpha}(\eta, h)=\Phi_{\beta}(\eta, h)$.

For $V \subseteq \mathbb{Z}^{d}$, let

$$
m_{V, \eta, h}(x)=\sup \left\{m_{\beta}(x): \beta \subseteq V \text { legal for }(\eta, h)\right\}
$$

$x \in \mathbb{Z}^{d}$. Notice that $m_{V, \eta, h}(x)=\sup \left\{m_{V^{\prime}, \eta, h}(x): V^{\prime} \subseteq V\right.$ finite $\}$. Let $m_{\eta, h}=m_{\mathbb{Z}^{d}, \eta, h}$.
Lemma 3 (Least Action Principle). If $\alpha$ is an acceptable sequence of topplings that stabilizes $(\eta, h)$ in $V$, then $m_{V, \eta, h} \leqslant m_{\alpha}$.

Lemma 4 (Global abelianness). If $\alpha$ and $\beta$ are both legal toppling sequences for $(\eta, h)$ that are contained in $V$ and stabilize $(\eta, h)$ in $V$, then $m_{\alpha}=m_{\beta}=m_{V, \eta, h}$. In particular, $\Phi_{\alpha}(\eta, h)=\Phi_{\beta}(\eta, h)$.
Lemma 5 (Monotonicity). If $V \subseteq \tilde{V}$ and $\eta \leqslant \tilde{\eta}$, then $m_{V, \eta, h} \leqslant m_{\tilde{V}, \tilde{\eta}, h}$.
A configuration $\eta$ is said to be stabilizable starting from odometer $h$ if $m_{\eta, h}(x)<\infty$ for every $x \in \mathbb{Z}^{d}$, and it is said to be explosive if $m_{\eta, h}(x)=\infty$ for every $x \in \mathbb{Z}^{d}$. The field of instructions $\mathcal{I}$ can be made explicit in the notation, for example, we say that $\eta$ is $\mathcal{I}$-stabilizable if $m_{\eta, h ; \mathcal{I}}(x)<\infty$ for every $x \in \mathbb{Z}^{d}$. The letter $h$ can be omitted in all the above notation when $h$ is identically zero.

Stabilization and fixation. We finally state the connection between stabilization for the DiaconisFulton representation and fixation for the continuous-time dynamics. Assume that distribution $\nu$ of $\eta_{0}$ on $\mathbb{N}_{0}^{\mathbb{Z}^{d}}$ is spatially ergodic and has finite density $\nu(\eta(o))<\infty$. We further assume that the support of $p(\cdot)$ generates $\mathbb{Z}^{d}$ and not a sublattice. Let $\mathcal{I}$ be independent of $\eta_{0}$, and distributed as follows. For each $x \in \mathbb{Z}^{d}$ and $j \in \mathbb{N}$, choose $\tau^{x, j}$ as $\tau_{x y}$ with probability $\frac{p(y-x)}{1+\lambda}$ or $\tau_{x \mathfrak{s}}$ with probability $\frac{\lambda}{1+\lambda}$, independently over $x$ and $j$. To avoid extra notation, we define the field $\mathcal{I}$ on the same probability space $\mathbb{P}^{\nu}$.

Lemma 6. $\mathbb{P}^{\nu}\left(\right.$ fixation of $\left.\left(\eta_{t}\right)_{t \geqslant 0}\right)=\mathbb{P}^{\nu}\left(m_{\eta_{0}}(o)<\infty\right)=0$ or 1 .

Mass-transport principle. The mass-transport principle will be used a couple of times in the proof, and consists of the following. Let $f(x, y)=f(x, y ; \omega)$ be a non-negative function of two points $x, y \in \mathbb{Z}^{d}$ and the randomness $\omega$. Suppose that, for every translation $\theta$ of $\mathbb{Z}^{d}$, the law of $\theta \omega$ is the same as the law of $\omega$ (whatever $\theta \omega$ means), and moreover $f(\theta x, \theta y ; \theta \omega)=f(x, y ; \omega)$ for every $x, y$ and a.e. $\omega$. Then

$$
\mathbb{E} \sum_{y} f(x, y)=\mathbb{E} \sum_{y} f(y, x), \quad \text { for any } x \in \mathbb{Z}^{d}
$$

This identity says that on average the mass sent by $x$ equals the mass received by $x$, although "mass" can be any non-negative function. See [13] for proof and applications.

## 3 Stabilization

Instead of proving Theorem 1 as stated, we consider the following equivalent formulation.
Theorem 7. Let $d, \lambda$ and $p(\cdot)$ be given. Let $\nu_{1}$ and $\nu_{2}$ be two spatially ergodic distributions for initial states on $\mathbb{Z}^{d}$, with respective densities $\zeta_{1}<\zeta_{2}$. If the $A R W$ system is a.s. fixating with initial state $\nu_{2}$, then it is also a.s. fixating with initial state $\nu_{1}$.

Let us give a brief sketch before moving to the proof.
The proof is algorithmic and has two stages, both stages being infinite. The idea is very simple and is related to what is sometimes called decoupling. Let $\eta_{0}$ and $\xi_{0}$ be independent and distributed as $\nu_{1}$ and $\nu_{2}$. In the first stage, we embed $\eta$ in $\xi_{0}$. In the second stage, we use the same set of instructions to evolve both systems; since $\xi$ fixates a.s., so does $\eta$, concluding the proof. More precisely, in the first stage we force each particle of $\eta$ to move (by waking it up when necessary) until it meets a particle of $\xi_{0}$; once they meet, they are paired and will not be moved until the second stage. Even if it takes infinitely many steps to finish pairing, a.s. every particle in $\eta$ will eventually be paired and the resulting odometer will be a.s. finite at every site (if the odometer were infinite somewhere, by ergodicity it would be infinite everywhere, so every particle in $\xi_{0}$ would be paired, which implies $\zeta_{2} \leqslant \zeta_{1}$ ), yielding a configuration $\eta_{0}^{\prime} \leqslant \xi_{0}$. In the second stage, we simply evolve the system using the remaining instructions. Since they are independent of $\xi_{0}, \eta_{0}$, and of the instructions used in the first stage, by assumption the remaining instructions a.s. stabilize $\xi_{0}$ leaving a locally-finite odometer. By monotonicity of the final odometer with respect to the configuration, the same set of remaining instructions also stabilizes $\eta_{0}^{\prime}$, again with a locally-finite odometer. Adding the odometer of both stages would give the final locally-finite odometer given by stabilization of $\eta_{0}$, except that in the first stage we have not followed the toppling rules correctly. But it still gives an upper bound due to monotonicity of the odometer with respect to waking up particles.

We now turn to the proof. To make the argument precise we will not 'move' particles as in the previous sketch, as the embedding requires an infinite number of topplings. We instead explore the instructions and define a sequence of configurations in terms of $\eta_{0}, \xi_{0}$ and $\mathcal{I}$. We end up concluding that a.s. the result of this exploration implies that $\eta_{0}$ is stabilizable, which in turn implies the statement of the theorem.

Embedding of the smaller configuration. Without loss of generality we can assume that $\nu_{1}$ or $\nu_{2}$ is not only ergodic but also mixing. Let $\eta_{0}, \xi_{0}$ and $\mathcal{I}$ be given, and take $h_{0} \equiv 0$. For $k=1,2,3,4, \ldots$, suppose $\eta_{k-1}$ and $h_{k-1}$ have been defined. Denote by $A_{k}$ the set given by

$$
A_{k}=\left\{x: \eta_{k-1}(x)>\xi_{0}(x)\right\}
$$

and consider an arbitrary enumeration

$$
A_{k}=\left\{x_{1}^{k}, x_{2}^{k}, x_{3}^{k}, \ldots\right\} .
$$

Let

$$
\left(\eta_{k}, h_{k}\right)=\lim _{j} \Phi_{\left(x_{1}^{k}, x_{2}^{k}, \ldots, x_{j}^{k}\right)}\left(\eta_{k-1}, h_{k-1}\right)
$$

in case $A_{k}$ is infinite - in case it is finite, by ergodicity it is a.s. empty in which case we let $\left(\eta_{k}, h_{k}\right)=$ $\left(\eta_{k-1}, h_{k-1}\right)$. Notice that the condition $\eta_{k-1}(x)>\xi_{0}(x)$ is also satisfied when $\eta_{k-1}(x)=\mathfrak{s}$ and $\xi_{0}(x)=0$, so this operation may require waking up particles.

As we go through $j=1,2,3, \ldots$ in the above expression, for each $j$ the field $m$ is increased by one unit at $x_{j}^{k}$, so $h_{k}$ is well-defined and satisfies

$$
h_{k}(x)=h_{k-1}(x)+\mathbb{1}_{A_{k}}(x)
$$

The limit for $\eta$ is also well-defined because, for each site $x$, the sequence decreases for at most one value $j$. In case it decreases, it may send one particle to one other site $z \neq x$. Thus, by a standard use of the mass-transport principle, a.s. the configuration at each site $x$ increases a finite number of times (the expected number of times is less than one as can be seen by taking $f(x, y)$ as the indicator of the event that $x \in A_{k}$ and toppling $x$ sends a particle to $y$ ), so the limit $\eta_{k}$ is a.s. finite.

We now prove that, if $\lim _{k} h_{k}(o)=\infty$ with positive probability then we must have $\zeta_{1} \geqslant \zeta_{2}$.
First, we argue that $\mathbb{P}\left(h_{0}^{\prime}(o)=\infty\right)=0$ or 1 , where

$$
h_{0}^{\prime}(x)=\lim _{k} h_{k}(x) .
$$

By the Local Abelianness, $\left(\eta_{k}, h_{k}\right)$ does not depend on the enumeration of $A_{k}$. In particular, $\eta_{k}$ and $h_{k}$ are determined by $\eta_{0}, \xi_{0}$ and $\mathcal{I}$ in a translation-covariant way, so the random set of sites $x$ for which $h_{0}^{\prime}(x)=\infty$ is ergodic with respect to translations (this is true only because $\mathcal{I}$ is mixing and we assumed that $\nu_{1}$ or $\nu_{2}$ is mixing). Moreover, the event $h_{0}^{\prime}(o)=\infty$ a.s. implies the event that $h_{0}^{\prime}(z)=\infty$ for every $z$ such that $p(z)>0$. These two facts together imply that, either $h_{0}^{\prime}(x)=\infty$ a.s. for every $x$, or $h_{0}^{\prime}(x)<\infty$ a.s. for every $x$, see [16, proof of Lemma 4]. This proves the zero-one law.

Suppose $h_{0}^{\prime}(o)=\infty$ with positive probability. By the zero-one law we have $h_{0}^{\prime}(o)=\infty$ a.s., which means that $\mathbb{P}\left(\lim \sup _{k}\left\{o \in A_{k}\right\}\right)=1$. But if $o \in A_{k_{0}}$ for some $k_{0}$, then necessarily $\eta_{k_{0}-1}(o)>\xi_{0}(o)$ and hence, by definition of $A_{k}, \eta_{k}(o) \geqslant \xi_{0}(o)$ for all $k \geqslant k_{0}$ and therefore $\liminf _{k}\left\|\eta_{k}(o)\right\| \geqslant\left\|\xi_{0}(o)\right\|$. On the other hand, from the mass-transport principle we have $\mathbb{E}\left\|\eta_{k}(o)\right\|=\mathbb{E}\left\|\eta_{k-1}(o)\right\|=\cdots=\mathbb{E}\left\|\eta_{0}(o)\right\|=\zeta_{1}$ (to show the first identity, we let $f_{k}(x, y)$ be the indicator that, on step $k, x$ sends a particle to $y$, and let $f_{k}(x, x)$ be the number of particles that were present at $x$ at the beginning of stage $k$ and stayed at $\left.x\right)$. By Fatou's Lemma, $\zeta_{1} \geqslant \zeta_{2}$.

Since we are assuming $\zeta_{1}<\zeta_{2}$, we must have $h_{0}^{\prime}(o)<\infty$ a.s. Now, as we go through $k=1,2,3, \ldots$, the value of $\eta_{k}(o)$ can decrease only when $o \in A_{k}$, i.e., only when $h_{k}(o)$ increases. Hence, $\left(\eta_{k}(o)\right)_{k}$ is a.s. eventually non-decreasing, so it converges. Its limit $\eta_{0}^{\prime}(o)$ satisfies $\eta_{0}^{\prime}(o) \leqslant \xi_{0}(o)$, otherwise $o$ would be in $A_{k}$ for all large enough $k$ and $h_{0}^{\prime}(o)$ would be infinite. By translation invariance, a.s. $h_{0}^{\prime}(x)<\infty$ and $\eta_{0}^{\prime}(x)=\lim _{k} \eta_{k}(x) \leqslant \xi_{0}(x)$ for every $x$.

Stabilization of the embedded configuration. In the previous stage we obtained a pair $\left(\eta_{0}^{\prime}, h_{0}^{\prime}\right)$ a.s. satisfying $h_{0}^{\prime}(x)<\infty$ and $\eta_{0}^{\prime}(x) \leqslant \xi_{0}(x)$ for every $x \in \mathbb{Z}^{d}$. Let $\tilde{\mathcal{I}}$ be the set of instructions given by

$$
\tilde{\tau}^{x, j}=\tau^{x, h_{0}^{\prime}(x)+j}, \quad x \in \mathbb{Z}^{d}, j \in \mathbb{N}
$$

that is, the field obtained by deleting the instructions used in the embedding stage described above. Since the first $h_{0}^{\prime}(x)$ instructions have been deleted at each site $x$, stabilizing a system with the instructions in $\tilde{\mathcal{I}}$ instead of $\mathcal{I}$ is equivalent to starting with odometer at $h_{0}^{\prime}$ instead of $h_{0} \equiv 0$.

Now note that the collection of instructions $\left(\tau^{x, j}: x \in \mathbb{Z}^{d}, j>h_{0}^{\prime}(x)\right)$ played no role in the construction of $\eta_{0}^{\prime}$ and $h_{0}^{\prime}$, so they are independent of $\xi_{0}$ and $h_{0}^{\prime}$. Hence, $\tilde{\mathcal{I}}$ is an i.i.d. field just like $\mathcal{I}$, and it is also
independent of $\xi_{0}$.
Therefore, $\mathbb{P}\left[\xi_{0}\right.$ is $\tilde{\mathcal{I}}$-stabilizable $]=\mathbb{P}\left[\xi_{0}\right.$ is $\mathcal{I}$-stabilizable $]$, and the latter equals 1 by assumption. Since $\eta_{0}^{\prime} \leqslant \xi_{0}$, we have $\mathbb{P}\left[\eta_{0}^{\prime}\right.$ is $\tilde{\mathcal{I}}$-stabilizable $] \geqslant \mathbb{P}\left[\xi_{0}\right.$ is $\tilde{\mathcal{I}}$-stabilizable $]=1$. This means that a.s. there exists $h_{1}^{\prime}$ such that, for all finite $V \subseteq \mathbb{Z}^{d}$ and all $x \in \mathbb{Z}^{d}, m_{V, \eta_{0}^{\prime} ; \tilde{\mathcal{I}}}(x) \leqslant h_{1}^{\prime}(x)<\infty$.

Stabilization of the smaller configuration. We now recall some properties from these two stages to show that $\eta_{0}$ is a.s. $\mathcal{I}$-stabilizable with

$$
m_{\eta_{0} ; \mathcal{I}}(x) \leqslant h_{0}^{\prime}(x)+h_{1}^{\prime}(x)<\infty, \quad \forall x \in \mathbb{Z}^{d} .
$$

In the first stage, the limits $\eta_{0}^{\prime}$ and $h_{0}^{\prime}$, which are determined by $\eta_{0}, \xi_{0}$ and $\mathcal{I}$, almost-surely exist and satisfy $h_{0}^{\prime}<\infty$ and $\eta_{0}^{\prime} \leqslant \xi_{0}$. Suppose this event occurs, and let $V$ be a fixed finite set.

If we start from $\left(\eta_{0}, h_{0}\right)$ and perform all topplings in $V$ as well as particle additions to $V$ (coming from $V^{c}$ ), following the same order as in the first stage, only a finite number of operations will be performed, and we end up with a state that equals $\left(\eta_{0}^{\prime}, h_{0}^{\prime}\right)$ on $V$.

By the local Abelian property, we can add the particles first and then topple the sites in $V$ as in the first stage, obtaining the same result. This means that there is some $\bar{\eta}_{V} \geqslant \eta_{0}$ and $\alpha_{V}=\left(x_{1}, \ldots, x_{n}\right)$ contained in $V$ such that $m_{\alpha_{V}}=h_{0}^{\prime}$ on $V$ and

$$
\Phi_{\alpha_{V}}\left(\bar{\eta}_{V}, h_{0}\right)=\left(\eta_{0}^{\prime}, h_{0}^{\prime}\right) \text { on } V
$$

Now, in the second stage, we showed that a.s. there exists $h_{1}^{\prime}(x)<\infty$ such that $m_{V^{\prime}, \eta_{0}^{\prime} ; \tilde{\mathcal{I}}}(x) \leqslant h_{1}^{\prime}(x)$ for any finite $V^{\prime}$. Suppose this event occurs.

Notice that $m_{V, \eta_{0}^{\prime}, h_{0}^{\prime} ; \mathcal{I}}(x)=m_{V, \eta_{0}^{\prime} ; \tilde{\mathcal{I}}}(x)$, that is, to stabilize $\eta_{0}^{\prime}$ in $V$ using the shifted field of instructions is the same as stabilize $\eta_{0}^{\prime}$ in $V$ using the original field of instructions and shifted odometer. Therefore, there exists $\beta_{V}=\left(x_{n+1}, \ldots, x_{m}\right)$ contained in $V$ such that $m_{\beta_{V}} \leqslant h_{1}^{\prime}$ on $V$ and $\Phi_{\beta_{V}}\left(\eta_{0}^{\prime}, h_{0}^{\prime}\right)$ is stable in $V$.

By the above identity, $\Phi_{\beta_{V}} \circ \Phi_{\alpha_{V}}\left(\bar{\eta}_{V}, h_{0}\right)=\Phi_{\beta_{V}}\left(\eta_{0}^{\prime}, h_{0}^{\prime}\right)$, on $V$. Since the latter is stable in $V$, by the Least Action Principle we have

$$
m_{V, \bar{\eta}_{V} ; \mathcal{I}}(x) \leqslant m_{\alpha_{V}}(x)+m_{\beta_{V}}(x), \quad \text { for all } x \in \mathbb{Z}^{d} .
$$

Thus, by monotonicity,

$$
m_{V, \eta_{0} ; \mathcal{I}}(x) \leqslant m_{V, \bar{\eta}_{V} ; \mathcal{I}}(x) \leqslant m_{\alpha_{V}}(x)+m_{\beta_{V}}(x) \leqslant h_{0}^{\prime}(x)+h_{1}^{\prime}(x)<\infty .
$$

We now note that the above bound does not depend on $V$, so

$$
m_{\eta_{0} ; \mathcal{I}}(x)=\sup _{V \text { finite }} m_{V, \eta_{0} ; \mathcal{I}}(x) \leqslant h_{0}^{\prime}(x)+h_{1}^{\prime}(x)<\infty, \quad \forall x \in \mathbb{Z}^{d}
$$

which means that $\eta_{0}$ is stabilizable, concluding the proof.

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