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Nonequilibrium Phase Transitions in a Driven Sandpile Model

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Abstract

We construct a driven sandpile slope model and study it by numerical simulations in one dimension. The model is specified by a threshold slope σ_c , a parameter α , governing the local current-slope relation (beyond threshold), and j_{in} , the mean input current of sand. A nonequilibrium phase diagram is obtained in the $\alpha - j_{\text{in}}$ plane. We find an infinity of phases, characterized by different mean slopes and separated by continuous or first-order boundaries, some of which we obtain analytically. Extensions to two dimensions are discussed.

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The statistical mechanics of nonequilibrium steady states is a subject of growing general interest. Phase transitions between such states are by no means as well understood as their equilibrium counterparts. Some insight has been gained into this problem by the study of simple driven lattice models [1,2]. Sandpile models, which introduced the notion of *self-organized criticality* (SOC) [3] as a general explanation for the wide occurrence of power laws in nature, are a natural setting in which to study phase transitions far from equilibrium. Surprisingly, to our knowledge, this has not been attempted. We construct a driven sandpile model and show that it exhibits, in spite of its simplicity, a rich phase diagram, thus making it a good laboratory for the study of nonequilibrium phenomena. In particular, our model exhibits continuous transitions from pinned or threshold-dominated states to unpinned states; these are reminiscent of dynamical phase transitions in more complex systems such as sliding charge-density waves (CDW's) [4] and pinned flux-lattices [5].

Earlier studies of sandpile models have concentrated on SOC, either in steadily flowing sandpiles [6,7] or at the angle of repose [8]. Much of the work has been on *critical-height* models, in which the update rule depends only on the *height* at each site; *critical-slope* models (CSM) [8] have been studied to a lesser extent. In this Letter, we present a comprehensive study of a simple CSM in which the current-slope relation for slopes exceeding a threshold is controlled by a parameter α . We monitor the steady states of our model as a function of α and the mean input current j_{in} and find the rich nonequilibrium phase diagram (Fig. 1): It shows that there are many phases characterized by the average slope σ_{av} of the sandpile. In the *repose* phase, $\sigma_{\text{av}} = \sigma_c$, the slope at the angle of repose. As j_{in} is increased at fixed α , the repose phase undergoes a continuous transition (solid line) to an unpinned super-repose phase at which $\sigma_{\text{av}} - \sigma_c$ rises continuously from zero with an exponent $\beta \simeq 0.5$ (Fig. 2). This continuous transition is followed by a series (which we argue is infinite) of similar continuous transitions. The repose phase lies between two first-order boundaries: one at low α to a pinned super-repose ($\sigma_{\text{av}} > \sigma_c$) region, the other at large α , to a sub-repose phase ($\sigma_{\text{av}} < \sigma_c$). The lower one of these first-order lines meets the continuous line at $j_{\text{in}} = 0.5$ at a

multicritical point. The pinned super-repose region contains an infinity of phases, separated by first-order lines parallel to the j_{in} axis (Fig. 1); σ_{av} jumps at these boundaries. We also monitor height profiles, the equal-time height correlation function and the associated correlation length, and current autocorrelations and the associated correlation time. Both the correlation length and time (Fig. 3) diverge at the continuous transitions in Fig. 1. We show that our main results can be understood on the basis of a mean-field theory and a mode-softening argument.

In our model, integers h_i specify the heights of columns of sand at the sites i of a one-dimensional chain ($1 \leq i \leq N$). The stability of the column at a site is determined by a threshold condition which mimics the angle of repose for a real sandpile: When the height difference between a site i and its right neighbor ($i + 1$) (i.e., the local slope $\sigma_i = h_i - h_{i+1}$) exceeds σ_c , some sand topples to the right neighbor, and h_i is updated via

$$h_i \rightarrow h_i - j_i, \quad h_{i+1} \rightarrow h_{i+1} + j_i, \quad (1)$$

$$j_i = \mathcal{N}(\alpha \times \sigma_i) \Theta(\sigma_i - \sigma_c). \quad (2)$$

$\mathcal{N}(x)$ is the integer nearest to x , Θ the step function, and α a real number. This part of our update rule conserves the number of particles locally and yields a local current which increases with the local slope, thus preventing an unbounded buildup of particles in the pile [9]. The parameter α controls the current-slope relation for slopes exceeding σ_c . We restrict our study to $\alpha > 0$, since $\alpha \leq 0$ yields unphysical runaway behavior; the upper bound for α is chosen to limit the region we explore. The mean input current of sand particles j_{in} is another control parameter. At each time step, we add m particles to a randomly chosen site with probability p , so $j_{\text{in}} = p \times m$. We set $m = 10$ for specificity (our results do not depend on this choice) and cover the range $0.001 \leq p \leq 0.15$, so $0.01 \leq j_{\text{in}} \leq 1.5$. This addition of particles violates local particle conservation; and, for such an addition rate, the mean input current and the noise amplitude *per site* vanish as $N \rightarrow \infty$. Particles are allowed to leave our system through the right, but not the left ($i = 1$), boundary: any particle that

reaches the N^{th} site is removed immediately. Our boundary conditions and update rules (1-2) clearly pick a direction for the current (from left to right).

We use initial conditions in which $h_i = \sigma_c(N - i) + \delta_i$, where δ_i is an integer that assumes the values $0, \pm 1$ with equal probability. We update all sites simultaneously and allow the system to evolve till it attains a steady state, i.e., when $j_{\text{in}} = j_{\text{out}}$, the average output current (the average number of particles dropping out from the open boundary per unit time). In practice, we say that the steady state has been achieved when these two currents are within 1% of each other. Once the steady state is obtained [10], we accumulate data for $10^5 - 10^6$ updates per site (UPS). Data for averages are stored after every 50 UPS. We also average our data over 10 – 50 different initial conditions. To minimize boundary effects we ignore a few sites (three or more if necessary) near each boundary while computing all averages.

The nonequilibrium steady state of our model can be characterized by the mean slope σ_{av} (the local slope σ_i averaged over i and many time steps), which is the order parameter for our model. We also monitor the mean current j_{av} , the output current j_{out} , the equal-time height correlation function $C_{hh}(r) = \langle \langle [h_i^t - \langle h_i \rangle_t][h_{i+r}^t - \langle h_{i+r} \rangle_t] \rangle_i \rangle_t$, and the output current autocorrelation function $C_{jj}(\tau) = \langle [j_{\text{out}}^t - \langle j_{\text{out}} \rangle_t][j_{\text{out}}^{t+\tau} - \langle j_{\text{out}} \rangle_t] \rangle_t$, where $\langle \dots \rangle_i$ and $\langle \dots \rangle_t$ denote, respectively, averages over i and time t .

In Fig. 2 the asymptotes (dashed lines) $\sigma_{\text{av}} = \sigma_c$ and $\sigma_{\text{av}} = j_{\text{in}}/(2\alpha)$ indicate the behavior of σ_{av} at very low (< 0.5) and very large ($\gg 0.5$) j_{in} , respectively. The full curve (for $\alpha = 0.1$ and $N = 128$ in Fig. 2) shows that the approach to these limits is nontrivial: As j_{in} increases there are successive onsets, indicated by arrows. There might well be an infinity of such onsets (see below), but they become hard to resolve numerically at large j_{in} . The inset shows a finite-size scaling plot at the first of these onsets. For $j_{\text{in}} \lesssim 0.5$, $\sigma_{\text{av}} = \sigma_c = 10$, i.e., we have threshold-dominated behavior at low j_{in} . The asymptotic behaviors can be understood via the mean-field theory presented below.

The evolution equation for h_i^t is

$$h_i^{t+1} - h_i^t = -j_i^t + j_{i-1}^t + \eta_i^t, \quad (3)$$

where the noise η_i^t has a mean j_{in}/N that accounts for the addition of particles. If we average over this noise, then, in the steady state, we get $j_i - j_{i-1} = j_{\text{in}}/N$. If we impose the boundary condition $j_{i=N} = j_{\text{in}}$, we get $j_i = ij_{\text{in}}/N$ and thence a mean current $j_{\text{av}} = \langle \langle j_i \rangle_i \rangle_t = \frac{N+1}{2N} j_{\text{in}} \approx j_{\text{in}}/2$ for $N \rightarrow \infty$. We use these exact results to check our simulation. Our mean-field theory assumes that

$$\langle \langle \mathcal{N}(\alpha\sigma_i)\Theta(\sigma_i - \sigma_c) \rangle_i \rangle_t \simeq \langle \langle \mathcal{N}(\alpha\sigma_i) \rangle_i \rangle_t \langle \langle \Theta(\sigma_i - \sigma_c) \rangle_i \rangle_t. \quad (4)$$

For large j_{in} , most $\sigma_i > \sigma_c$, so we further assume $\langle \langle \Theta(\sigma_i - \sigma_c) \rangle_i \rangle_t \simeq 1$. If its argument is large, the discontinuities of $\mathcal{N}(\alpha\sigma_i)$ are small relative to $\alpha\sigma_i$, so, on averaging Eq. (2), we make the approximation $j_{\text{av}} = \langle \langle \mathcal{N}(\alpha\sigma_i) \rangle_i \rangle_t = \alpha\sigma_{\text{av}}$, whence $\sigma_{\text{av}} \approx j_{\text{in}}/(2\alpha)$, the large- j_{in} asymptote of Fig. 2. Given these approximations, Eq.(3) yields a discrete diffusion equation for h_i^t with a spatially uniform source j_{in}/N . If we solve this with our boundary conditions we get parabolic height profiles for large j_{in} , which we also find in our simulations [11].

The the current-slope relation Eq. (2), shows that (Fig. 4), for a uniform profile with slope σ , no current flows if $\sigma \leq \sigma_c$. For $\sigma > \sigma_c$, the current grows with the slope in discrete steps, which reflect the \mathcal{N} function in Eq. (2). As we increase α , the width of the step at $j_i = 1$ shrinks till it becomes a single point at $\alpha = 0.125$, the value at which the continuous transition at $j_{\text{in}} = 0.5$ (Fig. 1) terminates. This can be understood as follows: If we turn Fig. 4 on its side, we see that the slope is pinned at σ_c below some threshold value of j_i . Beyond this threshold, the slope rises sharply before it saturates at another value of σ . The onsets in Fig. 2 are just the sharp steps of Fig. 4 rounded by our spatiotemporal average.

The vanishing of a step in Fig. 4 can also be linked with the termination of the continuous line via a mode-softening argument. Consider, e.g., the step at $j_i = 1$ for which the $11 \leq \sigma_i \leq 14$ (Fig. 4 with $\alpha = 0.1$). All slopes σ_i in this interval are equivalent in the sense that the sandpile dispenses the same amount of local current for all of them. Thus the restoring force, in response to a change of an onsite slope from σ to $\sigma + 1$, vanishes and leads to divergent correlation lengths and relaxation times (Fig. 3). Clearly the infinity of steps in Fig. 4 imply an infinity of onsets in Fig. 2, though the large- j_{in} onsets are hard to resolve

numerically.

The above arguments do not yield the value of j_{in} at onset since our spatiotemporal average shifts the values of the thresholds in Fig. 4. The actual value of j_{in} at the onset depends on the distribution of slopes in the interior of the pile. We have done some numerical and analytic calculations [11] on a ‘single-step model’ [9], which justify the occurrence of the transition at $j_{\text{in}} = 0.5$.

For the first onset we calculate the mean slope near $j_{\text{in}} \simeq 0.5$ for $\alpha = 0.1$ and $N = 32, 64, 128, 256$ and 512 and from a finite-size-scaling analysis (Fig. 2) obtain the exponents $\beta = 0.5$ and $\nu = 4.0$, where $J \equiv |j_{\text{in}} - j_c|/j_c$ with $j_c = 0.5$, and we use the scaling form $\delta\sigma = N^{-\beta/\nu} \mathcal{F}(J^\nu N)$, with $\delta\sigma \equiv \sigma_{\text{av}} - \sigma_c$. The exponent ν , which we obtain in this way from finite-size scaling, might not be the same as the actual correlation length exponent as has been pointed out in the context of sliding CDW’s [4].

In the low- j_{in} regime, as α increases from 0 to 0.05, σ_{av} decreases in steps of one. The values of α at which these steps occur yield the first-order boundaries of Fig. 1. All these first-order boundaries end at critical points in the range $1 \lesssim j_{\text{in}} \lesssim 1.2$. The overall envelope of the steps in σ_{av} can be fit approximately to a form $\sigma_{\text{av}} \sim 1/(2\alpha)$. This behavior follows from the update rule (Eq. 2): $\mathcal{N}(\alpha\sigma_i) = 0$ for $\alpha\sigma_i < \frac{1}{2}$, so, if $\alpha < 1/(2\sigma_c)$ ($=0.05$ for $\sigma_c = 10$), even a slope $\sigma_i > \sigma_c$ may become stable; e.g., with $\sigma_c = 10$ and $\alpha = 0.04$, $\mathcal{N}(\alpha\sigma_i) = 0$ for $\sigma_i = \sigma_c + 1$ and $\sigma_c + 2$, so the slope at the effective angle of repose turns out to be $\sigma_{\text{eff}} = \sigma_c + 2 = 12$. As α decreases, σ_{eff} also increases in discrete steps, leading to the first-order lines at low α .

We fit the equal-time height correlation function $C_{hh}(r)$ to an exponential form [11] which then yields a correlation length. Figure 3 shows this correlation length ξ_{hh} versus j_{in} for $\alpha = 0.1$ at different values of N . Clearly ξ_{hh} diverges at $j_{\text{in}} \simeq 0.5$ and also at subsequent continuous transitions. We have not characterized the divergences by an exponent because ξ_{hh} exceeds our system size somewhat before the transition (so we do not show data near $j_{\text{in}} = 0.5$).

The output current autocorrelation function $C_{jj}(\tau)$ does not fit an exponential form

very well. However, we have extracted correlation times τ_{jj} from the area under $C_{jj}(\tau)$ (normalized so that $C_{jj}(\tau = 0) = 1$). Figure 3 shows plots of τ_{jj} versus j_{in} for $\alpha = 0.1$, which sharpen with increasing N . This sharpening shows up clearly near $j_{\text{in}} = 0.5$ and leads to an increasing trend in τ_{jj} near the next onset ($j_{\text{in}} \simeq 1.1$).

It is generally believed that nonequilibrium phase transitions cannot occur in stochastic, one-dimensional models with short-ranged interactions [12]. We believe the transitions in our model occur because the noise amplitude per site vanishes as $N \rightarrow \infty$. This is why our one-dimensional model with short-range interactions exhibits phase transitions. We have checked that these phase transitions get rounded if there is a finite noise amplitude *per site* as $N \rightarrow \infty$. We have also performed simulations on two-dimensional versions of our model. Our preliminary results indicate that, with low noise and a variety of boundary conditions, such two-dimensional models show the same transitions as the one-dimensional model discussed here. In the steady state, the two-dimensional system behaves like uncoupled one-dimensional systems, so no transitions occur in the high-noise limit. The details of this study will be published elsewhere [11].

We have shown that our driven sandpile model displays a variety of steady states and many transitions between them. This richness, coupled with its simplicity, makes our model a promising one for the study of nonequilibrium phenomena in driven systems. As noted above, it displays transitions similar to those in other driven systems, e.g., our onset transitions (Fig. 2) are like unpinning transitions in sliding CDW's [4] or in pinned flux-lattice systems [5]. It would be interesting to study whether this similarity is merely superficial. There are some obvious ways in which the CDW models are different from ours: (1) They exhibit pinning because of quenched randomness but have no external noise; and (2) no current flows in their pinned states. The importance of these differences needs to be elucidated. In this general context it is interesting to study the zero-current limit of our model. We find that it *does not* show conventional SOC when the pile is allowed to relax, after each input of sand, to a completely quiescent state in which no further transfers are possible (à la Bak, Tang, and Wiesenfeld [3]). The precise forms of the distribution of avalanche sizes, etc., will

be reported elsewhere [11].

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FIGURES

FIG. 1. The nonequilibrium phase diagram of our model in the $\alpha - j_{\text{in}}$ plane. Full (dashed) lines indicate continuous (first-order) transitions. At low α there is an infinity of first-order boundaries; only the first five are shown. The corners in the phase boundaries and multicritical points are schematic; our data are too noisy to resolve them.

FIG. 2. A plot (full line) of σ_{av} versus j_{in} . Dashed lines are the asymptotes for large and small j_{in} . Arrows mark successive onsets (see text). The inset shows a finite-size scaling plot for $\delta\sigma N^{\beta/\nu}$ versus NJ^ν at $\alpha = 0.1$ and $j_{\text{in}} \gtrsim 0.5$ for $N = 64(+), 128(\diamond), 256(*)$ and $512(\triangle)$ with $\beta = 0.5$ and $\nu = 4$.

FIG. 3. Equal-time height correlation length ξ_{hh} and output current autocorrelation time τ_{jj} versus j_{in} for $N = 32(\circ), 64(+), 128(\diamond), 256(*)$ and $512(\triangle)$. For ξ_{hh} we do not plot points in regions where $\xi_{hh} > N$.

FIG. 4. A plot of j_i versus σ_i (Eq. (2)) for $\alpha = 0.1$ and 0.125 and $\sigma_c = 10$.