Probability distribution of residence times of grains in models of rice piles

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We study the probability distribution of residence time of a grain at a site, and its total residence time inside a pile, in different rice pile models. The tails of these distributions are dominated by the grains that get deeply buried in the pile. We show that, for a pile of size L, the probabilities that the residence time at a site or the total residence time is greater than t, both decay as $1/t(\ln t)^x$ for $L^{\omega} \le t \le \exp(L^{\gamma})$ where γ is an exponent ≥ 1 , and values of x and ω in the two cases are different. In the Oslo rice pile model we find that the probability of the residence time T_i at a site i being greater than or equal to t is a nonmonotonic function of L for a fixed t and does not obey simple scaling. For model in d dimensions, we show that the probability of minimum slope configuration in the steady state, for large L, varies as $\exp(-\kappa L^{d+2})$ where κ is a constant, and hence $\gamma=d+2$.

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I. INTRODUCTION

Granular materials have drawn a lot of attention due to their complex flow behavior under different driving conditions [1]. Slowly driven piles of sand grains serve as a prototype for self-organized criticality (SOC) [2]. Although SOC was not seen in experiments on piles of sand [3], but experiments on piles of long-grained rice have shown evidence of power law distribution of avalanche sizes [4–6]. Historically, perhaps because of their relation to earthquake phenomena, studies of sandpiles [7] have generally focused on the distribution of avalanche sizes. There are only a few theoretical studies of other interesting quantities such as the distribution of total residence times of grains in piles, even though the experimental studies by Frette and co-workers [4,5] using colored tracer grains are now almost a decade old.

In this paper we consider the probability distribution of residence times of grains at a site, and of their total residence times in the pile, in critical slope type slowly driven sandpile (equivalently rice pile) models. In these models, these residence time distributions are qualitatively different from the critical height type models. In critical height models, the distribution decays exponentially with average total residence time equal to average active mass in the pile [8]. In critical slope models, there is a possibility that the grain gets buried very deep in the pile, and then takes a long time to come out. We shall show that this makes the cumulative probability distribution of these residence times to have a characteristic 1/t decay for large times t (modified by a logarithmic multiplicative correction factor), independent of details of the toppling rules, and of the dimensionality of the system.

There have been some numerical and analytical studies of these distributions earlier. Frette has proposed a theoretical model, called the Oslo rice pile model [9], which seems to reproduce the phenomenology of the ricepile experiment well. In the experimental studies of ricepiles, Christensen and co-workers [4,5] estimated that average total residence time of grains in a pile of size L varies as L^{ν} , where they estimated $\nu=1.5\pm0.2$. From numerical simulations of the Oslo model for systems of size $L \leq 1600$, the exponent characterizing the power law decay of the probability density of total residence times at large times was estimated as 2.2 ± 0.1 . Boguna and Corral [10] and Carreras *et al.* [11] have used a continuous-time random walk model of the motion of grains, with long trapping times and a power-law distribution of step sizes, to explain the anomalous diffusion of tracer grains. In earlier papers [8,12], we have studied the total residence time distribution in the critical height type sandpile models with both deterministic and stochastic toppling rules. We reduced the problem to a diffusion problem of a single particle in a medium with space-dependent jump rates and showed that the distribution of the total residence time does not have any power-law tail. We also obtained the nonuniversal scaling form of the distribution which depends on the probability distribution of where grains are added into the pile.

In this paper we study the distribution of total residence times of grains in the pile, and also of residence times at a site, in a class of critical slope type sandpile models. We define the residence time T_i at a site *i* is the time spent by a grain at the site, measured in units of the time interval between successive addition of grains. The total residence time *T* is defined similarly. We show that the probability of the residence time at a site or the total residence time in the pile, being greater than or equal to *t*, decays as $1/t(\ln t)^{\delta}$ for a very wide range of *t*. The upper cutoff in both the distributions scales with system size *L* as $\exp(\kappa L^{\gamma})$ where γ is an exponent ≥ 1 and κ is a positive constant.

For the Oslo rice pile model, we find an unexpected behavior in the cumulative probability that a grain staying at a site *i* at least up to time *t* is not a monotonically increasing function of system size *L*. We argue that this implies the cumulative probability distribution function $\operatorname{Prob}_L(T_1 \ge t)$ cannot have a simple finite size scaling form. We show that $\gamma=d+2$ in *d* dimensions for this model.

The plan of the paper is as follows. In Sec. II we define the four models studied in this paper. In Sec. III we present the simulation results for the residence time T_1 at site 1 for the one-dimensional (1D) Oslo rice pile model and explain the nonmonotonic behavior of the cumulative distribution Prob_L($T_1 \ge t$) with L by relating the residence times of grains at the site 1 to the statistical properties of height fluctuation at that site. We also explain the origin of the multiplicative logarithmic correction factor appearing in the 1/t decay of Prob_L($T_1 \ge t$). In Sec. IV we discuss the 1/t power-law form of Prob_L($T \ge t$), where T is the total residence times, for large t in the 1D Oslo model and show that this also has a multiplicative logarithmic correction. In Sec. V we argue that the probability of minimum slope configuration occurring in the steady state of the 1D Oslo rice pile model, scales with system size L as $\exp(-\kappa L^3)$ where κ is some positive constant. In Sec. VI we present our simulation results for other models and show that in all cases the cumulative distributions is qualitatively similar to the 1D Oslo rice pile model. The last section contains a summary and some concluding remarks.

II. DEFINITION OF THE MODELS

We consider general critical slope type sandpile models where the configurations are specified by integer height variables $h(\vec{x})$, i.e., the number of grains at any site \vec{x} of a finite *d*-dimensional lattice. Whenever the height difference between two adjacent sites is greater than a threshold value, some specified number of grains are transferred to the neighboring sites. Piles are driven by adding grains, one at a time, at a fixed or at a randomly chosen site. Grains are added only when there are no unstable sites left in the system, and can leave the pile from the boundary. We update all unstable sites in parallel. We have studied four different models in both one and two dimensions: the Oslo rice pile model and its 2D generalization, and the local limited model and its variation. We now define the precise rules of these four models.

A. Model A: The Oslo rice pile model

The Oslo rice pile model [5] is defined as follows. We consider a one-dimensional rice pile, which is specified by an integer height variable h_i at each site i of a one-dimensional lattice, with $1 \le i \le L$. The slope at site i is defined to be $h_i - h_{i+1}$. Whenever the slope z_i at any site i is higher than a critical value $z_{c,i}$, the site becomes unstable and one grain from the unstable site goes to the right neighbor, i.e., $h_i \rightarrow h_i - 1$ and $h_{i+1} \rightarrow h_{i+1} + 1$. Whenever there is a toppling at site i, $z_{c,i}$ is randomly, independent of the history, reset to one of the two values 1 and 2, with probabilities q and p, respectively, where p+q=1. Whenever there is a toppling at site i=L (rightmost end), one grain goes out of the system. Grains are added only at site 1.

The 1D Oslo rice pile model has an Abelian property [13]. The final height configuration does not depend on the order in which we topple the unstable sites. After addition of total L(L+1) grains, the pile reaches the critical steady state [13]. Since we have chosen the values of z_c to be 1 or 2, in the steady state the height profile fluctuates between slopes 1 and 2. For number of sites L, the number of possible configurations in the critical states is exponentially large, approximately $[(1+\sqrt{5})/2\sqrt{5}][(3+\sqrt{5})/2]^L$, for large L [14]. The probabilities of various configurations in the steady state dif-

fer from one another by many orders of magnitude, unlike in Bak-Tang-Wiesenfeld model [2].

B. Model B: 2D generalization of the Oslo model

The Oslo model defined above can easily be generalized to two dimensions. We take a triangular region of a square lattice, the sites of which are indexed by (i,j) with $i,j \ge 1$ and $i+j \leq L+1$. The height of the pile at site (i,j) is denoted by h(i, j). Whenever the height difference between site (i, j)and any of its neighboring sites exceeds a critical value $z_c(i,j)$, assigned to the site (i,j), there is a toppling at site (i,j) and one grain is transferred from this site to the lower neighboring site toward the unstable direction. If there are more than one unstable directions, the grain is transferred toward the greatest slope. If the two directions have equal slope values, one grain is transferred randomly toward either one of these two directions. Whenever there is a toppling at a site $(i, j), z_c(i, j)$ is reset randomly, independent of the history, to either 2 or 1 with probability p or q, respectively, where p+q=1. One grain is lost whenever there is a toppling at the boundary sites, i.e., along the i+j=L+1 line. The model defined above in two dimensions is not Abelian because the final stable configuration depends on the order in which we topple the unstable sites. Grains are added only at the corner site (1,1).

C. Model C: The local limited model

The local limited model [15] is a one-dimensional model defined as follows. The slope z_i is defined as, $z_i=h_i-h_{i+1}$. Whenever the value of the slope z_i at any site *i* is higher than a critical value z_c , which we choose to be 2, the site becomes unstable and two grains from the unstable site go to the right neighbor, i.e., $h_i \rightarrow h_i-2$ and $h_{i+1} \rightarrow h_{i+1}+2$. The slope at some sites may be negative in the local limited model. Whenever there is a toppling at site i=L (rightmost end), two grains go out of the system simultaneously.

Grains are added uniformly everywhere. This model is also not Abelian. It is easy to see that in this case the maximum and the minimum slopes are 2 and 1, respectively. The total number of recurrent configurations in the steady state can be determined exactly, and varies as $4^{L}/L^{3/2}$ for large L [16]. It is known that the probabilities of occurrence of various configurations in the steady state are not equal, and may differ from one another by many orders of magnitude.

D. Model D: Model with non-nearest-neighbor transfer of grains

Model D is a variation of model C [15]. Whenever the value of the slope $z_i > 2$ at any site *i*, the site becomes unstable and two grains from the unstable site are transferred to the right, one grain transferred to site *i*+1 and the other one transferred to site *i*+2, i.e., $h_i \rightarrow h_i - 2$, $h_{i+1} \rightarrow h_{i+1} + 1$, and $h_{i+2} \rightarrow h_{i+2} + 1$. If there is a toppling near the right boundary, the grain goes out of the pile. The order in which we relax unstable sites matters. The grains are added uniformly everywhere. The local slope can be negative as in model C. The

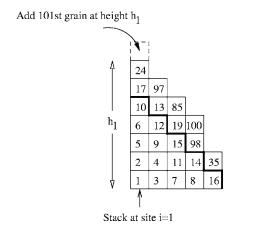


FIG. 1. Rice pile of size L=5 after addition of 100 grains. All grains are numbered whenever added in the pile. Minimum slope is denoted by the thick line.

minimum and maximum slopes in this model are also 1 and 2, respectively.

E. Marking grains and residence times

The pile in all four cases is driven slowly, by adding one grain per unit time, starting with the initial configuration of height zero at all sites. We assume that the time interval between addition of two grains is chosen long enough so that all avalanche activity has died before a new grain is added. The grain added at time n will be labeled by the number n. We think of the grains at a particular site as stacked vertically, one above the other (Fig. 1). Whenever a grain is added at a site, it sits on the top of the stack. When one unstable grain leaves the stack, it is taken from the top of the stack. In model C, when two grains leave a site, we first take out the topmost grain from the site and put it on the top of the right nearest-neighbor stack, then we take the second unstable grain and put it on the top of the first grain at the right nearest stack. In model D, we transfer the unstable grain, second from the top, to the right nearest neighbor and transfer the topmost one to next to the right nearest neighbor.

If a particular grain *n* enters a site *i* at time $t_{in}(i,n)$ and leaves the site at time $t_{out}(i,n)$, its residence time $T_i(n)$ at site *i* is defined as the time spent by the grain at the site *i*, i.e., $T_i(n) = t_{out}(i,n) - t_{in}(i,n)$. The residence time of the *n*th grain, T(n), is the total time spent by the grain inside the pile. For a directed rice pile in one dimension where grains move only in one direction and by one step in each toppling, the residence time T(n) equals $\sum_{i=1}^{L} T_i(n)$ (e.g., in models A and C). We define the function $\operatorname{Prob}_L(T_j \ge t)$ as the probability that a new grain added in the steady state of the pile will have a residence time at site *j* is greater than or equal to *t*, and $\operatorname{Prob}_L(T \ge t)$ as the probability that its total residence time in the pile is greater than or equal to *t*. Clearly, we have $\operatorname{Prob}_L(T_i \ge 0) = \operatorname{Prob}_L(T \ge 0) = 1$.

III. RESIDENCE TIMES AT THE FIRST SITE IN THE OSLO RICE PILE MODEL

The qualitative behavior of distributions $\operatorname{Prob}_L(T_i \ge t)$ for i=1 can be seen in the simulation results shown in Figs. 2

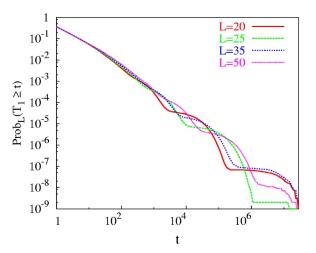


FIG. 2. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ versus time *t* for lattice sizes L=20, 25, 35, and 50 in the 1D Oslo rice pile model. Total 10^9 grains were added.

and 3. We have done our simulations for $p=q=\frac{1}{2}$ and different system sizes L=20, 25, 35, and 50. We averaged the data for a total 10^9 grains added in the pile for each L. Figure 2 shows the plot of $\operatorname{Prob}_{L}(T_1 \ge t)$ versus time t for different values of L. Interestingly, various curves for different L have step like structures. The curves for different values of L cross each other many times. The unusual nonmonotonic behavior is not an artifact of statistical fluctuations. The statistical errors in the data are much smaller than the step sizes except in the tail region (i.e., $t \ge 10^6$). The crossing of the curves for the cumulative probabilities persists for quite large system sizes also. In Fig. 3, we have plotted $\operatorname{Prob}_L(T_1 \ge t)$ versus t for two much bigger system sizes, L=100 and 200. We see that in this case also the probability that a grain remains in the pile of size L=100, for time greater than or equal to 6 $\times 10^5$, is higher by a factor 1.8 than for a pile of two times larger size L=200. Step like structures are not log periodic as the height and width of a step in each curve increases when going down the curve even on a logarithmic scale. The existence of several steps, whose positions and logarithmic widths are different for different L's, implies that simple finite size scaling cannot hold in this case.

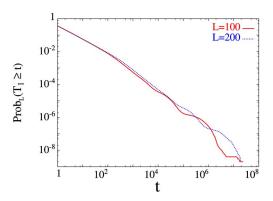


FIG. 3. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ versus time t for lattice sizes L=100 and 200 in the 1D Oslo rice pile model.

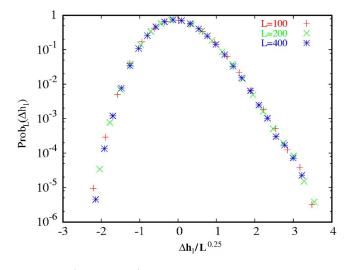


FIG. 4. (Color online) Scaling collapse of various probability distributions $\operatorname{Prob}_L(\Delta h_1)$ where Δh_1 is the deviation of height at site 1 about its average value, for different system sizes L=100, 200, and 400 for the 1D Oslo rice pile model.

A. Relationship between residence times T_1 and height fluctuations.

We can understand the residence time distribution of grains at any site in terms of the fluctuation of height at that site. The $h_i(t)$ is the height of the pile at a site *i* just after the *t*th grain has been added. This is a stochastic process and, in the steady state, it fluctuates in time between an upper bound h_{max} and a lower bound h_{min} . In case of the height fluctuation at site 1, $h_{max}=2L$ and $h_{min}=L$. The height $h_1(t)$ at the site 1 has a stationary probability distribution which is sharply peaked near its average value \bar{h}_1 , and has the width σ_{h_1} , which is the standard deviation of the fluctuation of height h_1 . In the steady state, the average value of h_1 varies as L, and the width σ_{h_1} varies as L^{ω_1} , where the exponent $\omega_1 < 1$. For large L, the probability distribution of h_1 has a scaling form as given below:

$$\operatorname{Prob}_{L}(h_{1}) = L^{-\omega_{1}}g\left(\frac{h_{1}-\bar{h}}{L^{\omega_{1}}}\right).$$
(1)

In Fig. 4 we have shown a scaling collapse of various probability distributions of height at site 1, $\operatorname{Prob}_L(\Delta h_1)$, where $\Delta h_1 = h_1 - \bar{h}_1$, for various values of system sizes L = 100, 200, and 400 in the 1D Oslo rice pile model. We get a good collapse using the scaled variable $\Delta h/L^{\omega_1}$ where $\omega_1 \approx 0.25$. Here the scaling function g(x) is nearly Gaussian for x near zero. But very large deviations of h_1 from the mean value are not well described in the Gaussian approximation. Later we shall argue that in the Oslo model the scaling function g(x) varies as $\exp(-|x|^{1/(1-\omega_1)})$ for $x \ge 1$ and it varies as $\exp(-|x|^{3/(1-\omega_1)})$ for $x \le -1$.

Let us consider variation of height h_1 at the first site with time t shown schematically in Fig. 5. Note that $h_1(t)$ has piecewise constant line segments, with possible jumps at the integer time t. The value of $h_1(t)$ at time t is denoted by the y coordinate of the line segment which is just at the right of

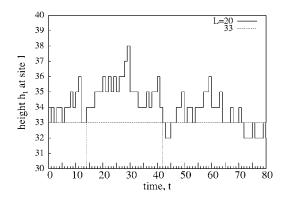


FIG. 5. Time fluctuation of height at first site is plotted as a function of time in the 1D Oslo rice pile model for L=20. The horizontal line is at $h_1=33$ which is the maximum probable height. The first and second vertical lines are at t=14, and 42, respectively.

the coordinate *t*, e.g., $h_1(0)=33$, $h_1(1)=34$, etc. A grain added at time *t* when the height at the first site is $h_1(t-1)$ leaves the site at time *t'*; we must have $h_1(t') \le h_1(t)$, and $h_1(t'') > h_1(t)$, for all *t''* satisfying t < t'' < t'. As an example, for the time series of $h_1(t)$ shown in Fig. 5, the grain added at t=14 stays at site 1 up to time t=41 and then goes out of the site 1 at t=42 (i.e., just after addition of the 42nd grain), and so $T_1(14)=28$. As $h_1(13)=h_1(12)$, the grain added at t=12comes out immediately, and hence $T_1(12)=0$.

Let $\operatorname{Prob}_L(T_1 \ge t | h_1)$ be the conditional probability that a grain stays at site 1 for time greater than *t*, given that it was added when the height was h_1 . Since $\operatorname{Prob}_L(h_1)$ is the probability that height was h_1 when the grain was added, we have the following, summing over all possible values of h_1 :

$$\operatorname{Prob}_{L}(T_{1} \ge t) = \sum_{h_{1}=h_{min}}^{h_{max}} \operatorname{Prob}_{L}(h_{1}) \operatorname{Prob}_{L}(T_{1} \ge t | h_{1}). \quad (2)$$

But $\operatorname{Prob}_{I}(T_{1} > t | h_{1})$ can also be written as the conditional probability that the height of the pile at site 1 would remain above h_1 for an interval $\geq t$, given that the height is h_1 in the steady state. This probability can be calculated from the general theory of Markov chains as the probability of the first return to a height less than or equal to h_1 , given that we start with height h_1 in the steady state, and add one grain per unit time. The probability that no return has occurred up to time t decreases as $\exp[-\lambda(h_1)t]$ for large t, where $\lambda(h_1)$ is the largest eigenvalue of the the reduced Markov matrix, with rows and columns corresponding to configurations with heights at site 1, below or equal to h_1 , removed [17,18]. While it is not very easy to calculate $\lambda(h_1)$ exactly, clearly it decreases as h_1 decreases. For $h_1 = h_{max}$, it is $+\infty$ as the height at the site cannot be higher than h_{max} and T_1 must always be zero. Also it is very small for h_1 near h_{min} , as the pile returns to very low values of h_1 only rarely.

For large *t*, in the sum in the right-hand side (r.h.s.) of Eq. (2), only terms with h_1 near h_{min} make a significant contribution. In this case, it is a reasonable approximation to replace the function Prob_L $(T_1 \ge t | h_1)$ by a simple exponential, with $\lambda(h_1) = \langle T_1 \rangle_{h_1}$. Thus we write, for large *t*,

PROBABILITY DISTRIBUTION OF RESIDENCE TIMES ...

$$\operatorname{Prob}_{L}(T_{1} \ge t | h_{1}) \simeq \exp(-t/\langle T_{1} \rangle_{h_{1}}).$$
(3)

It is easy to write the conditional expectation value of the residence time at the first site $\langle T_1 \rangle_{h_1}$, given that the grain was added when the height at site 1 is h_1 , in terms of the stationary probability distribution $\text{Prob}_L(h_1)$ exactly as

$$\langle T_1 \rangle_{h_1} = \frac{\operatorname{Prob}_L(\operatorname{height} > h_1)}{p_1 \operatorname{Prob}_L(h_1)} \tag{4}$$

where p_1 is the probability of adding a grain at site i=1. When we add grains only at first site, $p_1=1$ and when we add grains uniformly everywhere, $p_1=1/L$.

Proof. Define an indicator function $\eta_{n,t} = 1$, if the *n*th grain is at height h_1 at time *t*, and zero otherwise. Clearly, the sum of $\eta_{n,t}$ over *t* is the residence time of the *n*th grain at height h_1 . Then, averaging over *n* we get the mean residence time. But the sum of $\eta_{n,t}$ over *n* and *t* both gives a contribution whenever there is a grain at height h_1 , and hence is equal to $N\text{Prob}_L(\text{height} > h_1)$ where *N* is the total number of grains added and *N* is very large. Dividing this sum by the average number of grains added at height h_1 , which is equal to $p_1N\text{Prob}_L(h_1)$, we get $\langle T_1 \rangle_{h_1}$. Hence, Eq. (4) follows.

We substitute this estimate of $\langle T_1 \rangle_{h_1}$ in Eq. (3). We note that for large *t*, the terms in the summation that contribute significantly correspond to h_1 near h_{min} . For these values of h_1 , Prob(height> h_1) is nearly 1, and $\langle T_1 \rangle_{h_1}$ may be replaced, with small error, by $1/p_1 \text{Prob}_L(h_1)$ [see Eq. (4)]. Then Eq. (2) can be approximately written as given below:

$$\operatorname{Prob}_{L}(T_{1} \ge t) \simeq \sum_{h_{1}=h_{min}}^{h_{max}} \operatorname{Prob}_{L}(h_{1})e^{-tp_{1}\operatorname{Prob}_{L}(h_{1})}.$$
 (5)

Thus, the distribution of residence times T_1 can be expressed in terms of the probability distribution $\text{Prob}_L(h_1)$ of height h_1 .

B. Behavior of $\operatorname{Prob}_L(T_1 \ge t)$ for large *t*

Now we shall use the knowledge of the behavior of $\operatorname{Prob}_L(h_1)$ to explain the steplike structures in the distribution function $\operatorname{Prob}_L(T_1 \ge t)$.

For $h_1 \ll h_1$, the probability distribution of height $\operatorname{Prob}_L(h_1)$ falls very rapidly. Actually, it will be argued in Sec. V that for $h_1 \ll \overline{h}_1$ the ratio $\operatorname{Prob}_L(h_1 - 1)/\operatorname{Prob}_L(h_1)$ is of order $exp(-aL^2)$ where a is a constant and hence is very much less than 1. The values of $\operatorname{Prob}_L(h_1)$ for different h_1 's can differ by several orders of magnitude from each other, if h_1 is sufficiently near h_{min} . Now in the interval of $1/\operatorname{Prob}_L(h_1-1) \gg t \gg 1/\operatorname{Prob}_L(h_1)$, only a single term corresponding to h_1 contributes significantly to the summation, and then the summation is nearly independent of t. It is clearly seen from Fig. 6, where $\operatorname{Prob}_{L}(T_1 \ge t)$ is plotted versus t for L=20 in the 1D Oslo model. We can identify three steps in the plot. Each step in the curve can be associated with a unique value of h_1 (h_1 =28, 29, and 30) and steps appear at the corresponding value of $\operatorname{Prob}_L(h_1)$ along the y axis.

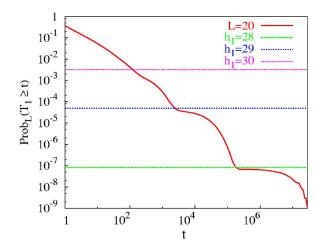


FIG. 6. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ versus time t for lattice size L=20.

This explains the steplike structure of $\operatorname{Prob}_L(T_1 \ge t)$ as a function of *t*. Also, the function decays roughly as 1/t since we must have $\operatorname{Prob}_L(h_1) \sim 1/t$ for the term to contribute in Eq. (5). If $h_1^*(t)$ is the value of h_1 that contributes most in Eq. (5), the value of $h_1^*(t)$ is given by the condition $\operatorname{Prob}_L[h_1^*(t)] \approx 1/t$. Substituting this condition in Eq. (1), we get

$$g\left(\frac{h_1^*(t) - \bar{h}}{L^{\omega_1}}\right) \approx L^{\omega_1/t}.$$
 (6)

Thus the *t* dependence of $h_1^*(t)$ comes through the scaling variable $tL^{-\omega_1} = \tau$ (say). Then for τ large the argument establishing the 1/t dependence of $\operatorname{Prob}_L(t_1 > t)$ given above is quite robust. However more careful analysis of Eq. (5) shows that there is also a logarithmic multiplicative correction factor with the 1/t decay of $\operatorname{Prob}_L(T_1 \ge t)$.

For large *L* and *t*, the terms that contribute to $\operatorname{Prob}_L(T_1 \ge t)$ in Eq. (5) correspond to the values of h_1 for which $h_1 \ll \overline{h}_1$. Substituting the scaling form of $\operatorname{Prob}_L(h) = (1/L^{\omega_1})g[(h-\overline{h}_1)/L^{\omega_1}]$ [see Eq. (1)] in Eq (5) and putting $x = (h-\overline{h}_1)/L^{\omega_1}$, we get the following:

$$\operatorname{Prob}_{L}(T_{1} \ge \tau L^{\omega_{1}}) \sim \int dx \ g(x) \exp[-g(x)\tau]$$
(7)

where $\tau = t/L^{\omega_1}$. We have assumed that the probability distribution $\operatorname{Prob}_L(h_1) \ll 1$, but is not rapidly decaying so that $\operatorname{Prob}_L(h_1-1)/\operatorname{Prob}_L(h_1) \approx 1$, and then the summation of Eq. (5) can be replaced by an integral over the scaled variable *x*. Actually in the real simulation (or experiment) for large *L*, this is the region of *t* we explore; we cannot go too far down the tail of $\operatorname{Prob}_L(T_1 \ge t)$.

Now it is easy to see the origin of logarithmic correction if we choose a particular form of the scaling function g(x) as $\exp(-|x|^{\alpha})$ for $x \leq 1$, with $\alpha > 0$, and try to find out the large *t* behavior of the above equation in terms of the scaling variable $\tau = t/L^{\omega_1}$. We first substitute $s = \exp(-|x|^{\alpha})$ in Eq. (7) and get

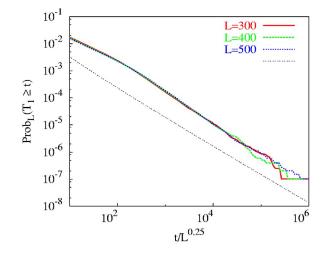


FIG. 7. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ for residence time at the first site has been plotted against the scaled residence time $t/L^{0.25}$ for lattice sizes L=300, 400, and 500 in the 1D Oslo rice pile model. Total 10^7 grains were added. The scaling function is fitted with $0.05/[x(\ln x)^{0.5}]$.

$$\operatorname{Prob}_{L}(T_{1} \ge \tau L^{\omega_{1}}) \sim \frac{1}{\alpha} \int \frac{ds}{\left[-\ln(s)\right]^{(\alpha-1/\alpha)}} \exp(-s\tau). \quad (8)$$

The asymptotic behavior of the above integral for τ large is easy to evaluate, giving

$$\operatorname{Prob}_{L}(T_{1} \ge \tau L^{\omega_{1}}) \sim 1/[\tau(\ln \tau)^{(\alpha - 1/\alpha)}]$$
(9)

In Fig. 7 we have plotted $\operatorname{Prob}_L(T_1 \ge t)$ against the scaled variable t/L^{ω_1} with $\omega_1 = 0.25$ for large values of system sizes L=300, 400, and 500 in the 1D Oslo model. We fit the scaled curves with a functional form given in Eq. (9) with $\alpha = 2$ since the scaling function g(x) is Gaussian near x=0.

Now it is clear that there is a logarithmic multiplicative factor in the 1/t decay and we take account of this logarithmic multiplicative correction by writing the cumulative probability as given below:

$$\operatorname{Prob}_{I}(T_{1} \ge t) \simeq L^{\omega_{1}} [t \ln^{\delta_{1}}(tL^{-\omega_{1}})]^{-1}$$
(10)

where we have used the fact that the answer is a function of the scaling combination $tL^{-\omega_1}$ and $\delta_1 = (\alpha - 1)/\alpha$. As a check we calculate the average residence time at site 1 as given below:

$$\langle T_1 \rangle \simeq \int_1^{T_{1,max}} \operatorname{Prob}_L(T_1 \ge t) dt \sim L^{\omega_1} \ln(T_{1,max})^{1-\delta_1}.$$
(11)

The upper cutoff on the time scale is provided by $1/\operatorname{Prob}_L(h=h_{\min})$, which is the average time interval between successive returns to the minimum height. Assuming the scaling function g(x) varies as $\exp(-|x|^{\alpha})$ for $x \ll -1$ and then putting $T_{1,max} = \exp[kL^{\alpha(1-\omega_1)}]$ in the above equation, we see that $\langle T_1 \rangle$ is proportional to *L*.

For the 1D Oslo model, a numerical estimate from the simulation gives $\omega_1 \approx 0.25$. Assuming the value $\alpha \approx 4$ (argued in Sec. V), we get $\delta_1 \approx 0.75$. In Fig. 8 we have plotted

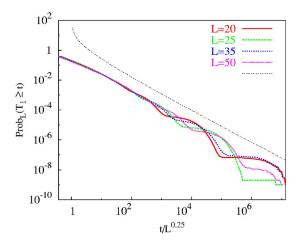


FIG. 8. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ for residence time at the first site has been plotted against the scaled time $t/L^{0.25}$ for lattice sizes L=20, 25, 35, and 50 in the 1D Oslo rice pile model. Total 10⁹ grains were added. The envelope is fitted with $5/[x(\ln x)^{0.75}]$.

Prob_L($T_1 \ge t$) versus a scaled variable t/L^{ω_1} where $\omega_1 = 0.25$ for L=20, 25, 35, and 50 and fitted the envelope formed by the steps in the curves with a function $1/x(\ln x)^{\delta_1}$ where $\delta_1 = 0.75$. We see that we get a reasonable fit to the data. We note that the multiplying logarithmic factor is necessary to get a good fit to the 1/t dependence.

IV. THE DISTRIBUTION OF RESIDENCE TIMES T IN THE OSLO MODEL

The arguments given in the previous section are easily extended to distribution of the residence times T_i with $i \neq 1$, and we conclude that they would also have a similar 1/t distribution with the same logarithmic correction factor as for T_1 , so long as *i* is not near the right end. Hence the distribution of their sum $T = \sum_i T_i$ would also be of the same form.

Even though the cumulative distribution of residence times T_i at any site *i* has a steplike structure, the step structure may be washed out in the sum $\Sigma_i T_i$.

Results of the numerical simulation for distribution of the total residence times using a total 5×10^7 grains are shown in Fig. 9. We see that steps are not seen in the distribution $\operatorname{Prob}_L(T \ge t)$ for different values of L for the range of the total residence times reached in the simulation $(T \le 10^8)$. The function $\operatorname{Prob}_L(T \ge t)$ is much smoother than the function $\operatorname{Prob}_L(T_1 \ge t)$. However, for small values of L (say for $L \le 20$), various curves of $\operatorname{Prob}_L(T \ge t)$ still cross each other at large times. But for larger values of L, we do not see any intercrossing of the curves in the times reached in our simulation (except at the tail where the data are less reliable due to the statistical fluctuations).

In analogy with results for the the distribution of T_1 , We can expect the behavior of the cumulative distribution $\operatorname{Prob}_L(T \ge t)$ to be a scaling function of t/L^{ω} , where the exponent ω is different from ω_1 defined earlier. So we write

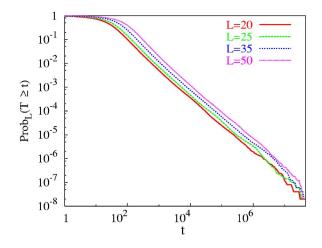


FIG. 9. (Color online) The cumulative probability $\operatorname{Prob}_L(T \ge t)$ versus time *t* for lattice sizes L=20, 25, 35, and 50 in the 1D Oslo rice pile model. Total 5×10^7 grains were added.

$$\operatorname{Prob}_{L}(T \ge t) = f\left(\frac{t}{L^{\omega}}\right) \tag{12}$$

where the scaling function f(x) varies as $1/[x(\ln x)^{\delta}]$ for large *x*, and the exponent δ will also be different from δ_1 defined earlier. Using the condition that the mean residence time in the pile is equal to the mean active mass in the pile, and hence scales as L^2 , we can determine δ in terms of ω and γ by integrating $\operatorname{Prob}_L(T \ge t)$ over *t* up to the cutoff time scale $\exp(\kappa L^{\gamma})$. Now we get

$$\delta = 1 - (2 - \omega)/\gamma. \tag{13}$$

In Fig. 10 we have plotted $\operatorname{Prob}_L(T \ge t)$ versus the scaled variable t/L^{ω} where $\omega \approx 1.25$ [10]. We get the value of δ approximately equal to 0.81 from Eq. (13), assuming $\gamma=3$ (argued in Sec. V). The fit is seen to be very good. In the numerical analysis of Christensen *et al.* [5], no logarithmic

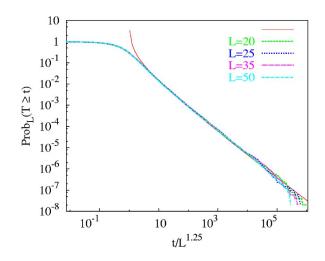


FIG. 10. (Color online) Scaling collapse of $\operatorname{Prob}_L(T \ge t)$ versus scaled residence time $t/L^{1.25}$ in the 1D Oslo rice pile model for lattice sizes L=20, 25, 35, and 50. Total 5×10^7 grains were added. Scaling function is fitted with $0.28/[x[(\ln x)^{0.81}]]$.

factor was used, and the data were fitted with a larger effective exponent, i.e., $1/t^{1.22}$ decay.

V. PROBABILITY OF MINIMUM SLOPE IN THE OSLO MODEL

The function $\operatorname{Prob}_L(h_1)$ can be exactly calculated numerically for small *L* using the operator algebra satisfied by addition operators [13]. We denote any stable configuration by specifying slope values at all sites from *i*=1 to *L*, *e.g.*, $|122...21\rangle$. Whenever the slope z_i becomes 2 after addition or toppling at site *i*, we denote the slope by $\overline{2}$, i.e., $|...\overline{2}...\rangle$. The overbar denotes that the site may topple or become stable with probability *q* or *p*, respectively.

Using these two toppling rules repeatedly and the Abelian property of the 1D Oslo rice pile model, we can relax any unstable configurations. For example, if we relax $|\overline{22}\rangle$ for L=2, we get the following sequence: $|\overline{22}\rangle \rightarrow p|2\overline{2}\rangle + q|1\overline{2}\rangle \rightarrow p^2|22\rangle + pq|12\rangle + pq|1\overline{2}\rangle + q^2|\overline{21}\rangle \rightarrow \cdots \rightarrow p^2|22\rangle + (p + p^2)q|12\rangle + (p + p^2)q^2|21\rangle + (p + p^2)q^3|02\rangle + (1 + p)q^4|11\rangle.$

The probability of maximum slope configuration (i.e., when $h_1=2L$) can be easily calculated. We start with the unstable configuration $|\overline{22}...\overline{2}\rangle$. The probability that no site topples in this unstable configuration is p^L and this is the probability of the maximum slope configuration (i.e., $h_1 = 2L$) in the steady state. That this probability varies exponentially with *L* can be incorporated in the scaling hypothesis by assuming that the scaling function g(x) in Eq. (1) varies as $\exp(-ax^{(1/1-\omega_1)})$ for $x \ge 1$ where *a* is a constant.

The probability of the minimum slope configuration cannot be calculated so easily. However we argue below that this probability asymptotically varies as $\exp(-\kappa L^3)$ where κ is a constant.

First, the above calculation for L=2 showed that, in the steady state, the probability of the minimum slope configuration is $\mathcal{O}(q^4)$. For L=3, we calculated this probability explicitly [19]; it is $\mathcal{O}(q^{10})$. A similar analysis, for other values of L=1-20, shows that the probability of the minimum configuration is $\mathcal{O}(q^{m_L})$, where m_L is exactly given by the formula L(L+1)(L+2)/6. The coefficient of q^{m_L} in the probability is harder to compute explicitly for large L. We conjecture that this simple formula holds true for all L. Then for sufficiently small q, the probability of the minimum height configuration in the 1D Oslo model varies as $\exp[-\kappa(q)L^3]$, where $\kappa(q)$ is a q-dependent function. Then, as there is no change in the behavior of the Oslo rice pile expected, as a function of q, this behavior should persist for all nonzero q. For the scaling function, this would imply that g(x) varies as $\exp[-\kappa(q)|x|^{(3/1-\omega_1)}]$ for $x \ge 1$.

We have calculated $\operatorname{Prob}_L(\operatorname{slope}=1)$, i.e., the probability of the minimum slope configuration, exactly numerically for q=0.50, 0.60, 0.75 for L=1-12 and the logarithm (base *e*) of $\operatorname{Prob}_L(\operatorname{slope}=1)$ has been plotted versus L(L+1)(L+2)/6 in Fig. 11.

More specifically, consider a very low-slope unstable configuration $|11...1\overline{2}\rangle$ which has a total number *L* of grains with slopes 1 at all sites except at the last site with slope 2

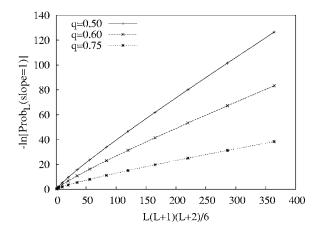


FIG. 11. Logarithm (base *e*) of the probability of occurrence of minimum slope configuration (calculated exactly) is plotted versus L(L+1)(L+2)/6 in the 1D Oslo rice pile model. We calculated for L=1-12.

and estimate the probability to go to the minimum slope from this configuration. To do this, we have to remove *L* grains, and each grain has to be moved a distance of $\mathcal{O}(L)$ on the average. Thus we need $\mathcal{O}(L^2)$ steps for large *L*, and each step requires a factor *q* in the probability. Actually the probability of transition from this configuration with height h_{min} +1 to the minimum slope configuration (with height h_{min}) is $\mathcal{O}(q^{L(L+1)/2})$ and the coefficient of $q^{L(L+1)/2}$ in this case is exactly 1. Now the probability of the minimum slope can be written in a general form as given below:

$$\operatorname{Prob}(\operatorname{slope} = 1) \sim \exp[-\kappa(q)L^3]$$
(14)

where $\kappa(0) = \infty$ and $\kappa(1) = 0$. The different asymptotic behavior of large deviations in g(x) is somewhat unexpected, but has been seen in other problems, such as the distribution of the large deviation of current in the asymmetric exclusion process in a ring [20].

VI. GENERALIZATION TO OTHER MODELS

In this section we present the simulation results in other models and show that the cumulative distributions $\operatorname{Prob}_L(T_1 \ge t)$ and $\operatorname{Prob}_L(T \ge t)$ have same 1/t power law behavior for large *t*, but with different logarithmic corrections.

A. Model B: Rice pile model in two dimensions

Now we present the simulation results for the 2D rice pile model. We add marked grains at the corner site, i.e., at (1,1). We simulate this model choosing p=0.75 and q=0.25, and study the residence time distribution of grains at the corner site (1,1). The standard deviation $\sigma_{h_{1,1}}$ of height $h_{1,1}$ at the corner site about the mean varies as L^{ω_1} where we estimated $\omega_1 \approx 0.2$ from the simulation. We added a total of 10⁶ grains.

We have plotted various curves for cumulative distribution function $\operatorname{Prob}_L(T_{1,1} \ge t)$ of residence time $T_{1,1}$ at the corner site versus times *t* for L=12, 15, and 20 in Fig. 12 and we see a steplike structure appearing for $t \ge 50$. The various curves for different *L* intercross each other many times as

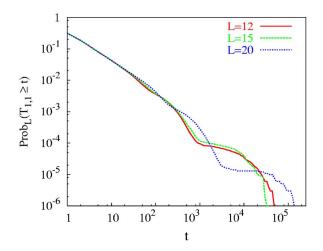


FIG. 12. (Color online) The residence time distribution $\operatorname{Prob}_L(T_{1,1} \ge t)$ of grains at the corner site versus time *t* in model B for lattice sizes L=12, 15, and 20. Total 10⁶ grains were added.

seen in the 1D Oslo rice pile model. In this case also, the step like structures are not log periodic as the step length in each curve increases on a logarithmic scale when going down the curve. Any simple finite size scaling in the whole range of $t(t \ge 1)$ does not work as in the case of the 1D Oslo rice pile model. The probability distribution of height at the corner site has a scaling form as given in Eq. (1) for the 1D Oslo rice pile model. From the simulation we determined the exponent $\omega_1 \approx 0.2$ which is very small. Since, for $t \le L^{\omega_1}$, $\operatorname{Prob}_L(T_{1,1} \ge t)$ scales with t/L^{ω_1} for large L [see Eq. (7)], we plotted various cumulative distributions $\operatorname{Prob}_L(T_{1,1} \ge t)$ versus a scaled time t/L^{ω_1} with $\omega_1=0.2$ in Fig. 13. We see that decay of the envelope formed by the various steps in different curves fits well with the function $2/[x(\ln x)^{0.5}]$ where the logarithmic correction factor is according to Eq. (9).

Using a similar argument as in the 1D Oslo model, in this case, we must have $\gamma=4$. To get to the minimum slope configuration, we will have to topple $\mathcal{O}(L^3)$ grains and each

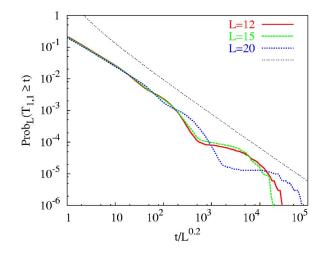


FIG. 13. (Color online) The cumulative probability distribution function $\operatorname{Prob}_L(T_{1,1} \ge t)$ versus scaled residence time $t/L^{0.2}$ in model B for lattice sizes L=12, 15, and 20. Total 10⁶ grains were added. The envelope formed by the steps is fitted with $2.0/[x[(\ln x)^{0.5}]]$.

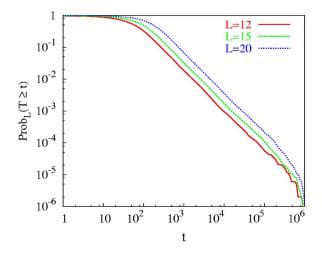


FIG. 14. (Color online) The residence time distribution $\operatorname{Prob}_L(T \ge t)$ in model B for lattice sizes L=12, 15, and 20. Total 10^6 grains were added.

grain $\mathcal{O}(L)$ times. As the average mass of the pile, in this case, varies as L^3 , Eq. (13) is modified as given below:

$$\delta = 1 - (3 - \omega)/\gamma.$$

In Fig. 14 we have plotted various distributions of the total residence time, $\operatorname{Prob}_L(T \ge t)$, versus time *t* for different L=12, 15, and 20. In Fig. 15 we have plotted $\operatorname{Prob}_L(T \ge t)$ for different *L* against the scaling variable t/L^{ω} where $\omega \approx 2.0$. Now we can estimate δ to be approximately 0.75 from the above equation. In Fig. 15 we fit the scaling function for $\operatorname{Prob}_L(T \ge t)$ with $0.2/[x(\ln x)^{0.75}]$ which seems to be a reasonable fit.

B. Model C: The local limited model

Residence time T_1 . Since in model C, grains are added randomly everywhere in the pile, the average residence time

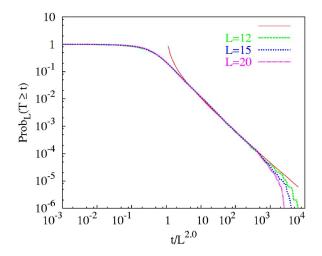


FIG. 15. (Color online) Scaling collapse of various $\operatorname{Prob}_L(T \ge t)$ versus scaled variable $t/L^{2.0}$ in model B for lattice sizes L = 12, 15, and 20. Total 10⁶ grains were added. The scaling function is fitted with $0.2/[x(\ln x)^{0.75}]$.

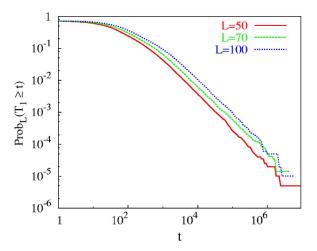


FIG. 16. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ versus time T_1 for lattice sizes L=50, 70, and 100 in model C. Total 10⁶ grains were added.

 $\langle T_1 \rangle_{h_1}$ of a grain added at height h_1 varies as $1/[p_1 \operatorname{Prob}_L(h_1)]$ according to Eq. (4). Now there will be an extra 1/L factor inside the exponential in the Eq. (7). Consequently the scaling variable $\tau = t/L^{\omega_1}$ in Eq. (7) is replaced by $\tau = t/L^{1+\omega_1}$ and Eq. (7) is modified to

$$\operatorname{Prob}_{L}(T_{1} \ge \tau L^{1+\omega_{1}}) \sim 1/[\tau(\ln \tau)^{(\alpha-1)/\alpha}].$$
(15)

Similarly the average residence time at the first site equals $\langle h_1 \rangle / p_1$ [the proof is similar to that for $\langle T_1 \rangle_{h_1}$ in Eq. (4)] which, in this case, varies as L^2 . This can be checked directly by integrating the above equation up to the cutoff time scale as done in Eq. (11).

The total number of grains added in the pile is different for different L so that 10⁵ grains are added at the first site. The standard deviation σ_{h_1} of height fluctuations at the first site varies as L^{ω_1} with system size L, where $\omega_1 \approx 1/3$ [21]. We have plotted $\operatorname{Prob}_L(T_1 \ge t)$ for different values of L in the logarithmic scale in Fig. 16. We note that, unlike in the 1D

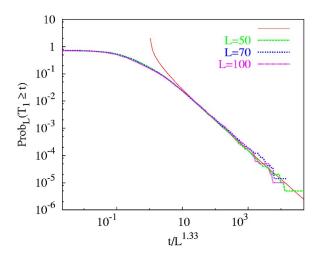


FIG. 17. (Color online) Scaling collapse of $\operatorname{Prob}_L(T_1 \ge t)$ for lattice sizes L=50, 70, and 100 in model C. Total 10^6 grains are added. The scaling function is fitted with $0.41/[x(\ln x)^{0.5}]$.

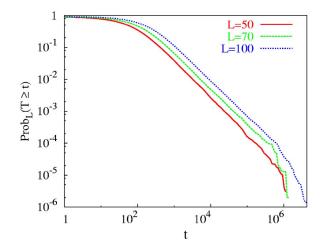


FIG. 18. (Color online) The distribution function $\operatorname{Prob}_L(T \ge t)$ versus time *t* in model C for lattice sizes L=50, 70, and 100. Total 10⁶ grains were added.

Oslo rice pile model, the cumulative probability here is a smooth (except at the tail due to statistical fluctuations) and monotonic function of *L* for a fixed *t*. This is due to the fact that the probability distribution $\operatorname{Prob}_L(h_1)$ of height at the first site is not as sharply a decaying function for $h_1 \ll \overline{h_1}$ as it was in the 1D Oslo model. In fact, in Fig. 17 we get a good scaling collapse of various $\operatorname{Prob}_L(T_1 \ge t)$ for different *L* using the scaled residence time $t/L^{1+\omega_1}$ where $\omega_1 \approx 1/3$. The scaling function is fitted well with the function $1/[x(\ln x)^{(\alpha-1)/\alpha}]$ for $x \ge 1$, taking $\alpha = 2$ [see Eq. (9)].

Total residence time T. In Fig. 18 we have plotted various $\operatorname{Prob}_L(T \ge t)$ versus residence time t for lattice sizes L=50, 70, and 100. Total 10⁶ grains were added in this case. In Fig. 19 we have collapsed various $\operatorname{Prob}_L(T \ge t)$ curves for different L using the scaled variable t/L^{ω} where $\omega \approx 1.5$. We fitted the scaling function of the cumulative distribution with $1/2x[\ln(2x)]^{\delta}$ for $x \ge 1$ where $\delta \approx 0.63$.

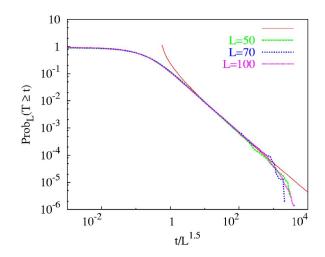


FIG. 19. (Color online) Scaling collapse of $\operatorname{Prob}_L(T \ge t)$ versus scaled time $t/L^{1.5}$ for lattice sizes L=50, 70, and 100 in model C. Total 10⁶ grains were added. The scaling function is fitted with $0.37/\{2x[\ln(2x)]^{0.63}\}$.

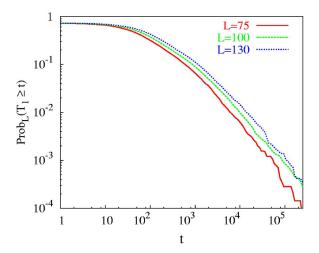


FIG. 20. (Color online) The cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$ versus time *t* for lattice sizes L=75, 100, and 130 in model D. 10⁴ grains were added at the first site.

C. Model D: Variation of the local limited model

In model D, the standard deviation σ_{h_1} of the hight fluctuation at site 1 scales with L as L^{ω_1} where we found $\omega_1 \approx 0.53$. In Fig. 20 we have plotted various $\operatorname{Prob}_L(T_1 \ge t)$ against the residence time t at the first site for L=75, 100, and 130. We added 10^4 grains at the first site. In Fig. 21 we have plotted $\operatorname{Prob}_L(T_1 \ge t)$ versus scaled time $t/L^{1+\omega_1}$ with $\omega_1 \approx 0.53$ and get a good scaling collapse of all the curves for various L. We fitted the scaling function with $0.73/[0.5x(\ln 0.5x)^{(\alpha-1)/\alpha}]$, using Eq. (9) and putting $\alpha=2$, as done in model C.

In Fig. 22 we have plotted various $\operatorname{Prob}_L(T \ge t)$ for the total residence time versus *t* for lattice sizes L=50, 70 and 100. Total 10⁵ grains were added in this case. In Fig. 23 we have plotted $\operatorname{Prob}_L(T \ge t)$ versus scaled time t/L^{ω} where $\omega \approx 1.5$ and we get a good collapse for the scaling function which fits reasonably well with the function $0.34/2x[\ln(2x)]^{\delta}$ for $\delta \approx 0.63$.

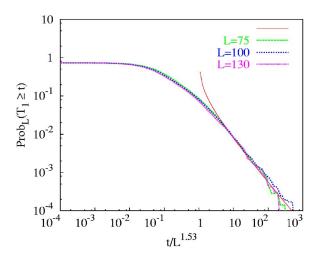


FIG. 21. (Color online) Scaling collapse of $\operatorname{Prob}_L(T_1 \ge t)$ for lattice sizes L=75, 100, and 130 in model D. The scaling function is fitted with $0.12/[x(\ln x)^{0.5}]$.

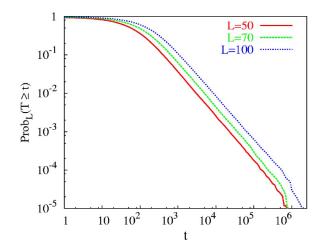


FIG. 22. (Color online) The cumulative probability $\operatorname{Prob}_L(T \ge t)$ versus time t for lattice sizes L=50, 70, and 100 in model D. Total 10⁵ grains were added.

VII. SUMMARY AND CONCLUDING REMARKS

To summarize, in this paper, we studied the distribution of the residence times of grains in various rice pile models. We reduced the problem of finding the residence time distribution of grains at a particular site to that of determining the distribution of first return times of the height at the site to the same value. The result that the probability of the residence times T_i at site *i* or the total residence time T in the pile, being greater than or equal to t, decays as a power law 1/t is valid for a large class of sandpile models, where the height fluctuation at a particular site grows with the system size, and is independent of dimensions. It depends only on the fact that there are some deeply buried grains which come out only in rare fluctuations, i.e., when the slope of the pile becomes very close to the minimum. It is important to note that, since the total residence time T is the sum of the T_i 's, the probability of T=0 decreases with increasing system size, and clearly our analysis cannot predict the small t behavior of the cumulative probability $\operatorname{Prob}_{I}(T \ge t)$, i.e., for $t < L^{\omega}$.

We also found that the cumulative probability $\operatorname{Prob}_L(T_1 \ge t)$, for a fixed *t*, is nonmonotonic with system size *L* for

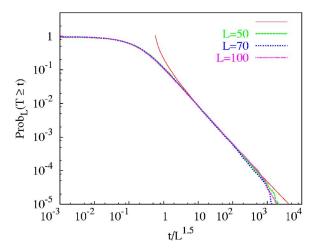


FIG. 23. (Color online) Scaling collapse of $\operatorname{Prob}_L(T \ge t)$ against the scaled variable $t/L^{1.5}$ for lattice sizes L=50, 70, and 100 in model D. The scaling function is fitted with $0.34/\{2x[\ln(2x)]^{0.63}\}$.

some of the rice pile models. The nonmonotonic behavior of the cumulative probability distribution $\operatorname{Prob}_L(T_i \ge t)$ of residence times at site *i* with system size *L* is possible when the probability distribution function $\operatorname{Prob}_L(h_i)$, where h_i is the height at site *i*, sharply decays for $h_i \ll \overline{h_i}$. However this nonmonotonicity is seen only for $t \ge t^*(L)$ where $t^*(L)$ increases with increasing values of *L*, and hence may be harder to observe in real experiments.

It is important to note that if we change the transfer rule of grains, the distribution of residence times may change completely. The rule chosen in this paper is called the firstin-last-out rule. We may employ some other rules, such as the first-in-first-out rule, adding the grain at the top of the stack but taking out grains from the bottom of the stack. A different rule would be to add and take out grains from a stack in random order. In these cases, there are no sites with deeply buried grains and the residence time distribution will be similar to that in the critical height models studied earlier by us.

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