Finite-size effects in roughness distribution scaling

T. J. Oliveira* and F. D. A. Aarão Reis†
Instituto de Física, Universidade Federal Fluminense, Avenida Litorânea s/n, 24210-340 Niterói RJ, Brazil
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We study numerically finite-size corrections in scaling relations for roughness distributions of various interface growth models. The most common relation, which considers the average roughness \( \langle w_2 \rangle \) as a scaling factor, is not obeyed in the steady states of a group of ballisticlike models in 2+1 dimensions, even when very large system sizes are considered. On the other hand, good collapse of the same data is obtained with a scaling relation that involves the root mean square fluctuation of the roughness, which can be explained by finite-size effects on second moments of the scaling functions. We also obtain data collapse with an alternative scaling relation that accounts for the effect of the intrinsic width, which is a constant correction term previously proposed for the scaling of \( \langle w_2 \rangle \). This illustrates how finite-size corrections can be obtained from roughness distributions scaling. However, we discard the usual interpretation that the intrinsic width is a consequence of high surface steps by analyzing data of restricted solid-on-solid models with various maximal height differences between neighboring columns. We also observe that large finite-size corrections in the roughness distributions are usually accompanied by huge corrections in height distributions and average local slopes, as well as in estimates of scaling exponents. The molecular-beam epitaxy model of Das Sarma and Tamborenea in 1+1 dimensions is a case example in which none of the proposed scaling relations work properly, while the other measured quantities do not converge to the expected asymptotic values. Thus although roughness distributions are clearly better than other quantities to determine the universality class of a growing system, it is not the final solution for this task.

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I. INTRODUCTION

In order to understand the basic mechanisms of a growth process, it is a common practice to associate it to a certain universality class [1,2]. This is usually done with the calculation of scaling exponents of surface roughness and assuming that Family-Vicsek scaling is valid [3] (with a suitable generalization when anomalous scaling holds). However, huge corrections to scaling make this calculation difficult in finite-size systems, both for simulation work and for the analysis of images of real surfaces. These problems and the interest in finding more universal properties of growth processes motivated recent studies of other quantities, such as the distributions of surface roughness [4–7]. Most efforts were devoted to calculation of those distributions in steady states (roughness saturation), but the comparison of distributions obtained under conditions that parallel experimental work (the so-called window boundary conditions) is also a promising tool [8–11].

It is expected that roughness distributions scale as

\[
P(w_2) = \frac{1}{\langle w_2 \rangle} f \left( \frac{w_2}{\langle w_2 \rangle} \right),
\]

where \( P(w_2) \) is the probability density of the roughness \( w_2 = h^2 - h^2 \) of a given configuration to lie in the range \([w_2, w_2 + dw_2]\). \( f \) is a scaling function, the overbars denote spatial averages and the angular brackets denote configurational averages. Equation (1) is usually obtained in analytical work, such as in the calculation of distributions for 1/f\(^a\)-noise interfaces [6]. An alternative scaling relation is

\[
P(w_2) = \frac{1}{\sigma} g \left( \frac{w_2 - \langle w_2 \rangle}{\sigma} \right),
\]

where

\[
\sigma = \left( \langle w_2^2 \rangle - \langle w_2 \rangle^2 \right)^{1/2}
\]

is the root mean square (rms) deviation of \( w_2 \).

It is reasonable to claim that comparison of full roughness distributions is better than the calculation of one or two exponents to determine the class of a growth process. However, these distributions may also be deviated from the expected scaling relations by finite-size corrections. For instance, finite-size data of some models in the Kardar-Parisi-Zhang (KPZ) class [12] show small deviations from relation (1), while good data collapse of numerical data for different lattice sizes is obtained with relation (2) [13]. The advantage of a scaling relation equivalent to Eq. (2) was also illustrated in the study of maximal height distributions (MHD) of 1/f\(^a\) signals [14]. Moreover, the leading corrections to scaling in MHD of some one-dimensional solid-on-solid models were analytically predicted and numerically confirmed in Ref. [15], which clearly shows that such corrections are not related to low accuracy of simulation data. The presence of huge scaling corrections may be much worse in distributions of other quantities, such as height distributions [16–19].

For the above reasons, investigation on the failure of finite-size scaling of roughness distributions is essential to ensure their reliability for comparison with real or model systems. Moreover, understanding the reasons for the failure of a scaling relation and the advantages of other ones may...
also be helpful to find the form of the main finite-size corrections and to propose alternative methods of analysis. One particularly important case is KPZ growth in 2+1 dimensions, since this class contains many models with large corrections to scaling of the average roughness, such as ballistic deposition (BD) and Eden growth, as well as a large number of applications (see, e.g., Refs. [20–22]).

The aim of this work is to study finite-size corrections in roughness distribution scaling by analyzing numerical simulation data of various interface growth models and to discuss the possible relations with other quantities that characterize those interfaces. The numerical approach is essential to study roughness distribution scaling by analyzing numerical simulations, since this class contains many models with large corrections to scaling relation. However, we discard the usual interpretation that the deviations from relation (1) are expected to be better than Eq. (2) by avoiding that finite-size effects are reflected in its second moment. This means that any finite-size dependence of the ratio $r$ (between the average square roughness and its rms fluctuation) will lead to a finite-size dependence of the second moment. However, in the scaling relation (2), $g(y)$ is a function of the variable $y=(w_z-(w_2))/\sigma$, so that the first moments of $g$ are

$$ M_g^{(0)} = 1, \quad M_g^{(1)} = 0, \quad M_g^{(2)} = 1, \quad M_g^{(3)} = \frac{(w_3^3)}{\sigma^4} - 3r - r^3. \quad (5) $$

This means that the finite-size dependence of $r$ does not lead to deviations in the second moment of this scaling function, leaving possible corrections to the third moment.

In works based on data collapse methods, the fluctuations in the first few moments of a scaling function are the ones most easily related to visual deviations of the plots of data from different system sizes. Consequently, scaling relation (2) is expected to be better than Eq. (1) by avoiding that finite-size effects are reflected in its second moment. This was already observed for some KPZ models in 2+1 dimensions in relatively small system sizes [13]. Recent work on maximal height distributions of 1/f signals also illustrated this feature and stressed the fact that finite-size scaling of the higher cumulants was related to the deviations [14].

The knowledge of the particular properties of a growth model is certainly important to guide the proposal of suitable scaling relations. For instance, this was the case of a recent work on elastic lines in random environments, in which the scaling relation (1) was generalized to incorporate time and temperature effects [29]. Here, we will consider a series of ballisticlike models whose deposits contain holes and overhangs, among other steep features, thus our approach is based on properties of related models.

Simulation work on models which produce deposits with holes and overhangs, such as the Eden model, suggested that the leading correction to the scaling of the average squared roughness is a constant term [23,24,30–32]. The time evolution of the average squared roughness, starting from a flat interface, obeys

$$ \xi_2(L,T) = L^{2\alpha}F(t) + W_2, \quad (6) $$

where $W_2$ is a constant called the intrinsic width, $\alpha$ is the roughness exponents, and $z$ is the dynamic exponent. The usual interpretation of relation (6) is that the first term on the
right-hand side (the original Family-Vicsek relation [3]) is associated with fluctuations of all wavelengths, while the intrinsic width is thought to be related to short wavelength features, represented by overhangs, holes, and high steps at the film surface—see, e.g., Refs. [24,30].

The present work is only concerned with steady state properties, where \( F(x) \to \text{const} \). Otherwise, the aggregation attempt is rejected. In the universality class of the model, and a constant contribution, which is the intrinsic width. Following this reasoning, we change relation (1) by subtracting a constant from \( w_2 \) and \( w_2 \):

\[
\langle w_2 \rangle = AL^{2\alpha} + W_2, \quad (7)
\]

with \( A \) constant. Large values of \( W_2 \) are responsible for remarkable deviations of finite-size estimates of exponent \( \alpha \) from its asymptotic value. However, numerical work on ballisticlike models show that finite-size corrections to the scaling of \( \sigma \) are smaller than the corrections in \( \langle w_2 \rangle \) [33].

These results lead us to propose that, for the ballisticlike models, the roughness of a given interface configuration is a sum of a fluctuating value, which is distributed according to the universality class of the model, and a constant contribution, which is the intrinsic width. Following this reasoning, we expect a constant from \( w_2 \) and \( w_2 \):

\[
P(w_2) = \frac{1}{\langle w_2 \rangle - C} h \left( \frac{w_2 - C}{\langle w_2 \rangle - C} \right).
\]

Here \( C \) must be viewed as a fitting constant and \( h \) is a scaling function. The first moments of the function \( h(z) = (\langle w_2 \rangle - C)P(w_2) \), where \( z = (w_2 - C)/\langle w_2 \rangle - C \), are

\[
M_h^{(0)} = 1, \quad M_h^{(1)} = 1, \quad M_h^{(2)} = \frac{\sigma^2}{\langle (w_2) - C \rangle^2} + 1. \quad (9)
\]

If our assumptions are correct and \( C \) is chosen to match the intrinsic width \( W_2 \), then the finite-size dependence of the second moment \( M_h^{(2)} \) will be canceled, or at least it will be greatly reduced. Consequently, the quality of a data collapse plot will be similar to that based on Eq. (2). Certainly this proposal has to be validated by numerical (or eventually analytical) work, and it is also important to notice that it is independent on the particular interpretation given to the intrinsic width.

III. DISCRETE GROWTH MODELS AND UNIVERSALITY CLASSES

The discrete models analyzed in this paper are the original BD model, the RSOS model, a conserved RSOS model, the grain deposition models, and the DT model. In all cases, deposition begins with a flat substrate.

In BD, each incident particle is released from a randomly chosen position above the deposit, follows a trajectory perpendicular to the substrate, and sticks upon first contact with a nearest neighbor occupied site [34,35].

In the RSOS model [36,37], the incident particle can stick at the top of a column only if the differences of heights of all pairs of neighboring columns do not exceed \( \Delta H_{\text{max}} \) after aggregation. Otherwise, the aggregation attempt is rejected. In this work, results for \( \Delta H_{\text{max}} = 1, \Delta H_{\text{max}} = 10, \) and \( \Delta H_{\text{max}} = 20 \) will be presented.

The original conserved RSOS model (CRSOS) was proposed in Ref. [37], but here we will consider an extension of that model that has the same symmetries and, consequently, belong to the same universality class. In this generalized CRSOS model [38], hereafter simply called CRSOS, if the column of incidence does not obey the above condition for aggregation, with \( \Delta H_{\text{max}} = 1 \), then the incident particle executes random walks among neighboring columns until finding a position where it is satisfied and aggregation can occur.

In the DT model [25], the arriving particle sticks at the top of the column of incidence if it has one or two lateral neighbors at that position. Otherwise, the neighboring columns are consulted and if one of them satisfies that condition, then the incident particle aggregates at that point. However, if no neighboring column satisfies the condition, then the particle sticks at the top of the column of incidence, and if both neighboring columns satisfy the condition, then one of them is randomly chosen.

The grain deposition models studied here were introduced in Ref. [39] for the study of a crossover in local roughness scaling similar to experimental systems. They are defined in a simple cubic lattice where the length unit is the lattice parameter. The grains have cubic shapes and lateral size \( l \). They sequentially flow towards the film surface with trajectories perpendicular to the flat substrate, having two of their faces parallel to the substrate. The incident grain permanently aggregates to the deposit when its bottom touches a previously aggregated grain (thus there is no lateral aggregation). The process is shown in Fig. 1. Here we will simulate the model with grain sizes ranging from \( l = 2 \) to 16. Despite the different aggregation rules, the grain deposition models and the BD model form the group to which hereafter we will refer to as ballisticlike models.

The symmetries of the ballisticlike models and of the RSOS models indicate that they are in the KPZ universality class (see, e.g., the discussion in Ref. [40]), in which the continuous description of the growth process is provided by the KPZ equation

\[
\frac{\partial h}{\partial t} = \nu_2 \nabla^2 h + \lambda_2 (\nabla h)^2 + \eta(x,t). \quad (10)
\]

Here, \( \nu_2 \) and \( \lambda_2 \) are constants, \( \eta \) is a Gaussian noise with zero mean and variance \( \langle \eta(x,t) \eta(x',t') \rangle = D \delta(x-x') \delta(t-t') \), with \( D \) constant, and \( d \) is the dimension of the sub-
in the following sections, we will study those models in 2+1 dimensions (in 1+1 dimensions, some exact results and highly accurate numerical results are already available).

The DT and the CRSOS models in 1+1 dimensions are in the class of the nonlinear fourth order growth equation [41–44]

$$\frac{\partial h}{\partial t} = \nu_4 \nabla^4 h + \lambda_{22} \nabla^2 (\nabla h)^2 + \eta(x, t), \quad (11)$$

where $\nu_4$ and $\lambda_{22}$ are constants. This is also known as the Villain-Lai-Das Sarma (VLDS) equation [45,46], and the discrete models are said to belong to the VLDS class. The DT model is particularly interesting in 1+1 dimensions due to the presence of strong finite-size corrections which leads to exponents estimates far from the VLDS values (although noise reduction methods are able to provide estimates closer to the VLDS ones [47]).

For all models, only steady state properties will be studied in this work. The roughness distributions are typically obtained from $10^7$ to $10^8$ different configurations, and height distributions from a number of configurations which is larger by a factor $L^2$ (the square of the lattice size).

IV. SCALING OF ROUGHNESS DISTRIBUTIONS

In order to compare the roughness distributions of the above models, we use two of them as representatives of their growth classes: the RSOS model with $\Delta H_{max}=1$, whose distribution hereafter will be referred to as the RSOS/KPZ one (2+1 dimensions, $L=256$), and the CRSOS model with $\Delta H_{max}=1$, whose distribution hereafter will be referred to as the CRSOS/VLDS one (1+1 dimensions, $L=256$). The use of these models data as standards is possible because previous works have already shown that they have very small finite-size dependence [13].

The original BD model illustrates the need of choosing a suitable scaling relation for the analysis of roughness distributions from small system sizes. In Figs. 2(a)–2(c) we show its steady state distributions in 2+1 dimensions obtained from $L=128$ to 512 and scaled according to Eqs. (1), (2), and (8), respectively. For comparison, the RSOS/KPZ scaled distributions are also shown. In Fig. 2(c), $C=12$ is used to match the peaks of the curves for BD and for RSOS/KPZ.

It is clear that the scaling with the rms fluctuation $\sigma$ [Eq. (2)] is superior if compared with the scaling with Eq. (1). Indeed, there is no data collapse in Fig. 2(a), thus that plot is unable to confirm the KPZ scaling of the BD model. From the discussion of Sec. II, the deviations may be related to finite-size corrections in the ratio $r$, and our estimates of $r$ for different sizes $L$ confirm this large size-dependence.

However, the quality of the data collapse using Eq. (8) is similar to the one using Eq. (2). This result suggests that the assumptions leading to the proposal of Eq. (8) are reasonable, i.e., the roughness of each configuration is a constant plus a KPZ-distributed fluctuating part (Sec. II). It also gives an estimate of the intrinsic width, which is the constant $C=12$ [Fig. 2(c)].

The scaled distributions of the grain deposition model with $l=16$ in lattice sizes $L=2048$ and 4096 are shown in Figs. 3(a)–3(c), considering Eqs. (1), (2), and (8), respectively. The RSOS/KPZ distribution is also shown. Deviations from the scaling relation (1) are also large here, but using Eqs. (2) and (8) we obtain good data collapse with these large lattice sizes. Similar results were obtained with an extended model where the grains could have Poisson-distributed sizes, with an average size $\bar{l}=16$ (the aggregation mechanism was the same of Fig. 1).

In BD and in this grain deposition model we observe deviations from data collapse with sizes smaller than those shown in Figs. 3(a)–3(c). The deviations for a certain lattice size $L$ become smaller when the grain size decreases, thus good data collapse is obtained for intermediate values of $l$ ($l=2–8$) with lattice sizes $L$ intermediate between those in Figs. 2(a)–2(c) and 3(a)–3(c). Typically, significant deviations appear only for $L\leqslant 64l$ ($l=1$ for BD), probably due to further corrections to scaling, whose effects are reduced as $L$ increases.

Our results for this group of ballisticlike models suggest that the roughness is actually distributed in a KPZ-like form except for an additional constant, which may be interpreted as the intrinsic width. Additional support to the constant correction term in the scaling of $\langle w_2 \rangle$ [Eq. (7)] is given in Fig. 4, where we plot $\langle (w_2)-C \rangle/L^{2\alpha}$ versus $L$, using $\alpha=0.39$ [18,19]
FIG. 3. Steady state square roughness distributions of the grain deposition model with \( L = 16 \) and lattice sizes \( L = 2048 \) (squares) and \( L = 4096 \) (triangles), scaled according to (a) Eq. (1), (b) Eq. (2), and (c) Eq. (8). In each plot, the solid curve is the RSOS/KPZ distribution scaled accordingly.

and the values of \( C \) obtained from collapse of roughness distributions. The saturation value of \( (\langle w \rangle - C)/L^{2\alpha} \) for each model gives an estimate of the amplitude \( A \) in Eq. (7). These estimates and the correction constants \( C \) are shown in Table I, where we also present the corresponding values of both terms on the right-hand side of Eq. (7) for the largest lattice size used for each model \( (L_{\text{MAX}}) \). The ratios between the leading correction \( (C) \) and the dominant term \( AL^{2\alpha}_{\text{MAX}} \) are very large for all models, in some cases exceeding 50%, which quantitatively confirm the strong finite-size effects.

However, it is important to notice that the scaling with Eq. (8) advances over Eq. (2) because it not only confirms the universality class of the model (by data collapse) but also provides evidence on the leading finite-size correction to the roughness scaling. These results also advance over previous work where the intrinsic width was identified only as a correction to scaling of the average roughness [23, 24, 30–32].

The intrinsic width is usually associated to steep surface features, such as large local slopes, holes, and overhangs. In order to test this hypothesis, it is interesting to consider other models which also have large local slopes. This is the case of RSOS models with large \( \Delta H_{\text{max}} \).

In Figs. 5(a) and 5(b) we show the roughness distributions for the RSOS models with \( \Delta H_{\text{max}} = 10 \) and 20 scaled according to Eqs. (1) and (2), respectively. Both plots show excel-

![FIG. 4. \( (\langle w \rangle - C)/L^{2\alpha} \) versus lattice size \( L \) for the ballisticlike models, where \( C \) is the constant used to fit roughness distributions to the new scaling relation. Symbols correspond to the ballistic deposition model (squares) and the grain deposition models with \( l = 2 \) (crosses), \( l = 4 \) (diamonds), \( l = 8 \) (circles), and \( l = 16 \) (triangles).](image1)

![FIG. 5. Steady state normalized square roughness distributions of the RSOS model with \( \Delta H_{\text{max}} = 10 \) (squares) and \( \Delta H_{\text{max}} = 20 \) (triangles) scaled according to (a) Eq. (1) and (b) Eq. (2). In each plot the solid curve is the RSOS/KPZ distribution scaled accordingly.](image2)
FIG. 6. (a) and (b) Steady state normalized square roughness distributions of the DT model in lattice sizes $L=64$ (squares) and $L=128$ (triangles) scaled according to Eqs. (1) and (2), respectively; (c) distributions for the original and the noise-reduced DT models in lattice size $L=64$; $m=1$ (original model—dashed line), $m=10$ (crosses), $m=20$ (squares), and $m=30$ (triangles). In all plots the solid curve is the CRSOS/VLDS distribution scaled accordingly.

FIG. 7. Correction to scaling $C$ as a function of the average square nearest neighbor height difference $\langle \delta h_{nn} \rangle$ of the ballisticlike models (open circles) and of the RSOS models (triangles).

and 30. Here, $m$ is the number of times that a certain column has to be chosen for aggregation of a new particle before the aggregation actually takes place [47]. In Fig. 6(c) we show the scaled distributions for these models in $L=64$, which converge to the CRSOS/VLDS curve as $m$ increases (although reasonable data collapse is not obtained in the tails).

This noise-reduction scheme is not expected to change the universality class of the process because symmetries are preserved (see, e.g., the discussion in Ref. [40]). Thus the above changes in the distributions when the parameter $m$ is increased [Fig. 6(c)] are evidence that the asymptotic behavior of the DT distribution was not attained yet. Consequently, the above discrepancies with the CRSOS/VLDS curve do not mean that the DT model does not belong to the VLDS class, but only that the finite-size effects on the roughness distributions have much more complex forms than those suggested here. Indeed, we recall that Ref. [41] derived the VLDS equation from the master equation of the DT model in $1+1$ dimension.

In the following sections, we will show that finite-size corrections obtained here in the roughness distributions are accompanied by corrections in other quantities.

V. ROLE OF THE LOCAL SLOPES

Here we characterize the interfaces at small length scales by the average squared height difference between nearest neighbor columns, $\langle \delta h_{nn} \rangle$, also calculated at the steady states. Our main aim is to test the hypothesis that the intrinsic width is related to local height fluctuations of the interface.

In the ballisticlike models, $\langle \delta h_{nn} \rangle$ has a significant finite-size dependence for small lattice sizes. However, it is approximately constant in the largest lattice sizes analyzed for each $l$ in Sec. III (the same sizes used to collapse the roughness distributions). However, in the RSOS models $\langle \delta h_{nn} \rangle$ attains a saturation value for very small system sizes.

In Fig. 7 we plot the correction term $C$ obtained in the scaling with Eq. (8) (i.e., the intrinsic width) as a function of the large $L$ value of $\langle \delta h_{nn} \rangle$ for both groups of models. The result for the ballisticlike models could naively suggest that the intrinsic width was only related to local height fluctuations. However, the result for the RSOS models show that
this interpretation is not valid in general: \( C = 0 \) in that case and the values of \( \langle \delta h_{nn}^2 \rangle \) are in the same range of those for ballisticlike models with \( l = 4 \).

In the DT model in 1+1 dimensions, we observe a rapid increase of the average local slope with the lattice size: \( \langle \delta h_{nn}^2 \rangle = 22 \) for \( L = 32 \), \( \langle \delta h_{nn}^2 \rangle = 58 \) for \( L = 64 \), and \( \langle \delta h_{nn}^2 \rangle = 139 \) for \( L = 128 \). A comparison with \( \langle \delta h_{nn}^2 \rangle \) for the other models shown in Fig. 7 is not possible because we are not able to perform a reliable extrapolation of these data to \( L \rightarrow \infty \). On the other hand, these data confirm that the DT model is very far from its asymptotic limit (the continuum VLDS model) in this range of lattice sizes.

The above results suggest that the presence of large fluctuations at small length scales is insufficient to explain the intrinsic width as the main correction term of roughness scaling in the ballisticlike models, since this is also a property of models which have much weaker scaling corrections (the RSOS ones). On the other hand, we observe that finite-size effects on roughness distributions are accompanied by the same type of effect on \( \langle \delta h_{nn}^2 \rangle \).

VI. FINITE-SIZE EFFECTS IN HEIGHT DISTRIBUTIONS

Finite-size effects in height distributions are typically larger than those in roughness distributions, as illustrated for KPZ models in Ref. [10]. Thus in order to study finite-size effects on steady state height distributions, the best procedure is to compare their skewness \( S \) and kurtosis \( Q \) in different lattice sizes (other previous works which aimed at characterizing height distributions also focused on those quantities—see, e.g., Refs. [16–19]). They are defined from the moments of the height distributions, \( W_n = \langle (h - \bar{h})^n \rangle \), as

\[
S = \frac{W_3}{W_2^{3/2}}
\]

and

\[
Q = \frac{W_4}{W_2^2} - 3.
\]

The sign of the skewness indicates which type of singular feature is dominant at the interface: deep valleys or grooves lead to negative \( S \), while sharp peaks give positive \( S \). In KPZ models the sign of \( S \) is the same as that of the coefficient of the nonlinear term \( \lambda \) of the corresponding KPZ equation (see, e.g., the discussion in Ref. [19]). Thus since ballisticlike models correspond to positive \( \lambda \) and RSOS models to negative \( \lambda \), we must compare values of \( S \) from the former with \( -S \) from the latter.

In Figs. 8(a) and 8(b) we show \( \pm S \) and \( Q \) versus \( 1/L^{1/2} \), respectively, for BD, for the grain deposition model with \( l = 8 \) and for the RSOS model with \( \Delta H_{\text{max}} = 10 \) (\( -S \) only for the RSOS model). Data for BD in the largest sizes were extracted from Ref. [19]. In Figs. 8(a) and 8(b), the variable \( 1/L^{1/2} \) was used in the abscissa because it is the one which provides the best linear fits among other variables of the form \( 1/L^\alpha \), with one digit variation in \( \Delta \). Since the asymptotic values of \( S \) and \( Q \) are not known exactly, as well as their finite-size corrections, such numerical extrapolations are necessary.

We observe very weak finite-size corrections in the RSOS data, so that a simple extrapolation procedure to \( L \rightarrow \infty \) \((1/L^{1/2} \rightarrow 0)\) leads to an asymptotic estimate of \( S \) in agreement with the best known value \(|S| = 0.26 \pm 0.01 \) of the KPZ class [18,19]. However, there are large finite-size corrections in the data for the BD model. It is remarkable that up to \( L = 500 \), the skewness of BD is negative, while the values for larger \( L \) are positive. The finite-size dependence of the data for the grain deposition model is much stronger: only for \( L = 4096 \) the skewness crosses over from negative to positive values.

The change in sign of \( S \) of the ballisticlike models shows that the typical steady state configurations for small \( L \) are very different from the typical ones for large \( L \). In other words, the steady states for small \( L \) are not representative of the asymptotic limit of the model (large \( L \) and \( t \)), which is the KPZ limit. For the grain deposition model with \( l = 16 \), even the steady state for \( L = 4096 \) has \( S = 0 \) and, consequently, is very far from the KPZ limit.

Despite these problems, the data for the BD ballisticlike models indicate that \( S \) is positive when \( L \rightarrow \infty \). Asymptotic estimates (not very accurate) of \( S \) can be obtained from extrapolations of those data; they are not very distant from the best known values for KPZ in 2+1 dimensions. However, the finite-size effects on roughness distributions are certainly smaller than those found in height distributions. Indeed, the good data collapse in Figs. 2(b), 2(c), 3(b), and 3(c) is reflected in values of skewness (of the roughness distributions) close to the RSOS/KPZ curve. The analysis of the kurtosis of the above models [Fig. 8(b)] leads to the same conclusions, although there the finite-size effects are much larger, as well as the error bars of the data.

The values of \( S \) and \( Q \) for the DT and the CRSOS models are shown in Figs. 9(a) and 9(b) as a function of \( 1/L^{1/2} \). They were estimated up to \( L = 256 \) for the DT model (this work) and up to \( L = 2048 \) for the CRSOS model (Ref. [38]), with a relatively low accuracy for the largest \( L \) data. Here the variable in the abscissa was chosen by analogy with Fig. 8, but not for extrapolation purposes.

The asymptotic skewness of the VLDS class, \(|S| = 0.32 \pm 0.02 \), can be obtained from extrapolation of the CR-
For various interface growth models, height and roughness distributions and average square neighboring heights differences \(\langle \Delta h_{mn}^2 \rangle\) were numerically calculated in the steady states in order to analyze their finite-size effects.

First, we considered a group of ballisticlike deposition models in which the most typical scaling relation for roughness distributions [Eq. (1)] fails to provide collapse of finite-size systems data. On the other hand, good data collapse was obtained with a scaling relation involving the roughness fluctuation [Eq. (2)] and with an alternative relation which includes the effect of the intrinsic width. This shows that this constant term is the main correction to the asymptotic KPZ distribution, thus extending previous work which suggested that quantity only as a correction to the scaling of the average roughness. The comparison with other models with large \(\langle \Delta h_{mn}^2 \rangle\) (the RSOS ones) show that the average local slope is not sufficient to explain the intrinsic width, in contrast to previous interpretation. Height distributions for those models also show strong finite-size effects, although results for sufficiently large lattice sizes indicate convergence of amplitude ratios to the expected asymptotic values.

We also studied the model of Das Sarma and Tamburenea (DT) in 1+1 dimensions. Its roughness distributions up to size \(L=128\) are very different from the ones representative of the asymptotic VLDS class. Strong finite-size effects are also observed in skewness and kurtosis of height distributions and in \(\langle \Delta h_{mn}^2 \rangle\), with no evidence of convergence to reliable asymptotic values.

Our results show that the deviations from finite-size scaling in the roughness distributions are usually accompanied by significant finite-size dependence of dimensionless amplitude ratios of moments of height distributions and of average local slopes. One important point is that this conclusion is not restricted to the KPZ class. The analysis of simple quantities, such as \(\langle \Delta h_{mn}^2 \rangle\), is useful to search for finite-size effects and may reveal slow convergence to a continuum limit. In some situations (e.g., the DT model here), this convergence may be so slow that no quantity analyzed here is able to show the correct class of the growth process. In the other cases, roughness distributions are certainly better than other quantities (including scaling exponents) to classify the process.

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