

## Parameter-free scaling relation for nonequilibrium growth processes

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We discuss a parameter-free scaling relation that yields a complete data collapse for large classes of non-equilibrium growth processes. We illustrate the power of this scaling relation through various growth models, such as the competitive growth model with random deposition and random deposition with surface diffusion or the restricted solid-on-solid model with different nearest-neighbor height differences, as well as through a deposition model with temperature-dependent diffusion. The scaling relation is compared to the familiar Family-Vicsek relation, and the limitations of the latter are highlighted.

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The study of growing interfaces has been a very active field for many years [1–3]. Many studies focus on the technologically relevant growth of thin films or nanostructures, but growing interfaces are also encountered in various other physical, chemical, or biological systems, ranging from bacterial growth to diffusion fronts. Over the years important insights into the behavior of nonequilibrium growth processes have been gained through the study of simple model systems that capture the most important aspects of real experimental systems [4,5].

In their seminal work, Edwards and Wilkinson [6] investigated surface growth phenomena generated by particle sedimentation under the influence of gravity. They proposed to describe this process in  $(d+1)$  dimensions by the following stochastic equation of motion for the surface height  $h(\mathbf{x}, t)$ , now called the Edwards-Wilkinson (EW) equation:

$$\frac{\partial h(\mathbf{x}, t)}{\partial t} = \nu \nabla^2 h(\mathbf{x}, t) + \eta(\mathbf{x}, t), \quad (1)$$

where  $\nu$  is the diffusion constant (surface tension), whereas  $\eta(\mathbf{x}, t)$  is a Gaussian white noise with zero mean and covariance  $\langle \eta(\mathbf{x}, t) \eta(\mathbf{y}, s) \rangle = D \delta^l(\mathbf{x} - \mathbf{y}) \delta(t - s)$ . Since Eq. (1) is linear, it can be solved exactly by Fourier transformations [2,4,6]. Later, Family [7] discussed the random deposition (RD) and random deposition with surface relaxation (RDSR) processes. RD [3,7] is one of the simplest surface growth processes. In this lattice model particles drop from randomly chosen sites over the surface and stick directly on the top of the selected surface site. Since there is no surface diffusion, the independently growing columns yield an uncorrelated and never-saturated surface. The RDSR process is realized by adding surface diffusion, which allows particles just deposited on the surface to jump to the neighboring site with lowest height. This diffusion step smooths the surface and limits the maximum interface width  $W(t)$ , defined at deposition time  $t$  as the standard deviation of the surface height  $h$  from its mean value  $\bar{h}$ :  $W(t) = \sqrt{\langle (h - \bar{h})^2 \rangle}$ . Starting from an initially flat surface, RDSR yields at very early times, with  $t < t_1 \sim 1$  (we assume here that one layer is deposited per unit time), a surface growing in the same way as for the RD process since no (or only very few) diffusion steps occur in that regime. For  $t > t_1$  the width increases as a power law of time with a growth exponent  $\beta$  before entering a saturation

regime after a crossover time  $t_2$ ; see Fig. 1. Both the saturation width  $W_2$  and the crossover time  $t_2$  are powers of the substrate size  $L$ :

$$W_2 \sim L^\alpha, \quad t_2 \sim L^z, \quad (2)$$

where  $\alpha$  is the roughness exponent and  $z$  is the dynamical exponent, with  $z = \alpha/\beta$ . In his study Family [7] noticed that the scaling exponents obtained through numerical simulations of the RDSR process agree with those obtained from the solution of the EW equation. The dependence of the growing interface on the substrate size  $L$  is summarized in the celebrated Family-Vicsek scaling relation [8]

$$W = L^\alpha f(t/L^z). \quad (3)$$

Combining this with the relations given in Eq. (2), we see that this scaling relation mainly consists of shifting the crossover points for the different system sizes to a common point with the new coordinates  $t' = t/L^z \sim t/t_2$  and  $W' = W/L^\alpha \sim W/W_2$ . It is worth noting that the Family-Vicsek relation neglects the RD regime at early times and exclusively focuses on the two regimes connected by the crossover point at  $t = t_2$ .

Scalings (2) and (3) are generic for growing interfaces and have been verified analytically, numerically, and experi-

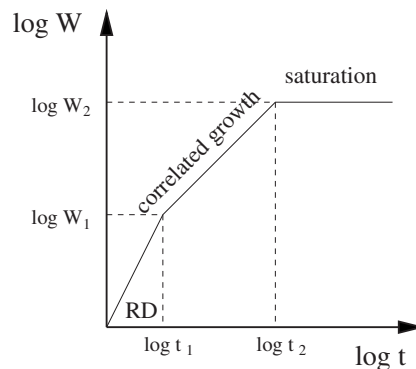


FIG. 1. Schematic plot of the interface width as a function of time for a typical deposition process. The early-time behavior (with  $t < t_1$ ) is that of the RD process. For  $t > t_1$  correlated growth sets in. Finally, the finite system displays at late times  $t > t_2$  a crossover to a saturation regime where the width of the interface remains constant,  $W = W_2$ .

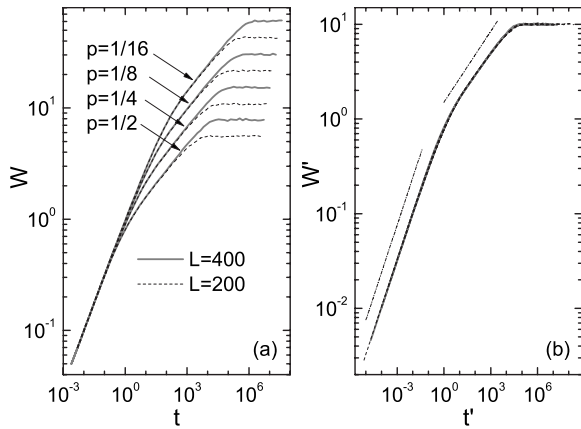


FIG. 2. (a) Log-log plot of the surface width vs time for the RD/RDSR process in systems with different sizes  $L$  and different probabilities  $p$ . (b) A complete data collapse of all data sets is achieved when using scaling relation (5). The curves shown in the left panel completely fall on top of each other and are no longer distinguishable. The axis labels are  $t' = t^\lambda / t_1^\lambda$  and  $W' = W^\lambda / W_1^\lambda$ ; see main text. The dashed lines indicate the expected slopes in the RD and EW regimes.

mentally in a large variety of systems. Various universality classes have been identified which differ by the values of the scaling exponents. Thus the RDSR process belongs to the Edwards-Wilkinson universality class with the exponents  $\beta = 1/4$  and  $z=2$  for a one-dimensional substrate. The RD process is in a universality class of its own which for a  $d=1$  substrate is characterized by the values  $\beta=1/2$  and  $z=\infty$ . Other well-known universality classes, directly related to technologically relevant growth processes, are the Kardar-Parisi-Zhang (KPZ) [9] and the conserved KPZ universality classes [10,11].

In recent years the study of nonequilibrium growth systems has shifted to more complex cases such as, for example, competitive growth models; see, e.g., [12–23]. In a competitive growth model one considers a mixture of two different deposition processes where one of them takes place with probability  $p$ , whereas the other takes place with probability  $1-p$ . One example is the RD/RDSR model [12], where the deposition happens according to the RDSR rules with probability  $p$  and to the RD rules with probability  $1-p$ . Whereas for  $p=1$  and  $p=0$  only one of the processes is realized, for general values of  $p$  the mixture of the two processes leads to a crossover between the two regimes where the crossover time and width depend on the value of  $p$  [see Fig. 2(a)]. A similar dependence on system parameters is also observed in the restricted solid-on-solid (RSOS) model [24], which exhibits a crossover from the RD regime to the important KPZ universality class. In this model new particles are incorporated into the growing surface only if the height differences between the deposition site and its neighboring sites remain smaller than some maximum height  $S$ . As discussed in [25] and shown in Fig. 3(a), the crossover time and width depend on the value of  $S$ .

In simple growth processes the random deposition regime is restricted to very early times. This is fundamentally different in more complex systems where the initial regime can

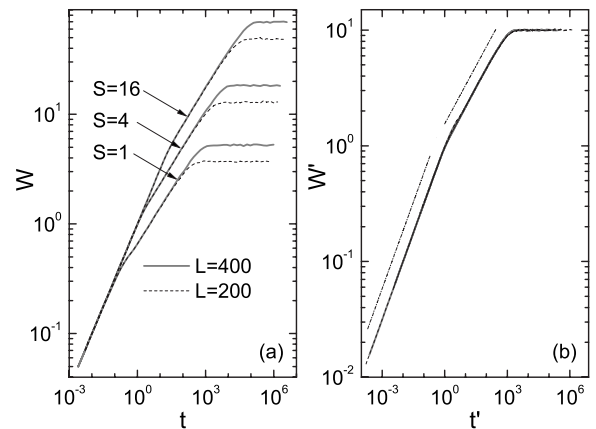


FIG. 3. The same as Fig. 2, but now for the RSOS model with different values  $S$  of the maximal nearest-neighbor height difference. Again a perfect data collapse is achieved when using scaling relation (5). The dashed lines indicate the expected slopes in the RD and KPZ regimes.

extend over very large times [12–23,25]. As already mentioned, Family-Vicsek scaling relation (3) assigns a new set of coordinates to the second crossover point. This does however not yield a complete data collapse for growth processes with two crossover points if one considers systems of different sizes. For the competitive growth models some phenomenological scaling relations have been proposed in the past, but these modified scaling relations also only allow a partial collapse of the different curves [12,16,18,23].

However, a scaling relation leading to a complete data collapse of *all* curves obtained for different system sizes and different values of the system parameters can indeed be obtained for any growth system that exhibits two different crossover points. This data collapse is achieved in a two-step process. First we translate all curves in a log-log plot such that the first crossover point is now located at the origin. This is achieved by plotting  $\log(\frac{W}{W_1})$  as a function of  $\log(\frac{t}{t_1})$ . In the second step we rescale both axes by the common scale factor  $\lambda = 1/\log(\frac{W_2}{W_1})$  such that in the log-log plot the second crossover point is fixed at the rescaled width  $\log W'_2 = 1$ . This isotropic rescaling, which conserves the slope of the region between the two crossover points, results in that the length of the line connecting the two crossover points is the same for all curves, and a complete data collapse, encompassing all three regimes, follows. Our proposed scaling relation can be cast in the following equation:

$$\frac{\log\left(\frac{W}{W_1}\right)}{\log\left(\frac{W_2}{W_1}\right)} = F\left[\frac{\log\left(\frac{t}{t_1}\right)}{\log\left(\frac{W_2}{W_1}\right)}\right], \quad (4)$$

where  $F(x)$  is a scaling function. Introducing  $\lambda = 1/\log(\frac{W_2}{W_1})$ , we can rewrite this as

$$W^\lambda = W_1^\lambda G\left[\left(\frac{t}{t_1}\right)^\lambda\right], \quad (5)$$

with a new scaling function  $G(y)$ . As shown in Figs. 2 and 3 for the RD/RDSR and RSOS processes, the proposed scaling

relation yields a complete data collapse for different system sizes  $L$  and different values of the system parameters. This perfect scaling behavior should be compared with the incomplete scaling proposed in the literature [12,16,18,23].

Obviously, scaling relation (5) is of universal use in growth systems with two crossover points and replaces the Family-Vicsek relation in these systems. This class of systems encompasses competitive growth models, but also the simple growth systems, for which the Family-Vicsek relation has been proposed originally, belong to this class. It is also worth noting that the properties of the different models only enter in our relation (5) implicitly through the dependence of the positions of the crossover points on the different system parameters.

We can also state the scaling relation in an alternative way which makes the difference to the Family-Vicsek relation more transparent. Indeed, a complete collapse can also be achieved when first moving the second crossover point to the origin in a log-log plot, yielding the relation

$$\frac{\log\left(\frac{W}{W_2}\right)}{\log\left(\frac{W_2}{W_1}\right)} = \tilde{F} \left[ \frac{\log\left(\frac{t}{t_2}\right)}{\log\left(\frac{W_2}{W_1}\right)} \right] \quad (6)$$

or

$$W^\lambda = W_2^\lambda \tilde{G} \left[ \left(\frac{t}{t_2}\right)^\lambda \right], \quad (7)$$

where  $\tilde{F}$  and  $\tilde{G}$  are again scaling functions. This scaling relation is completely equivalent to the first one, only the scales are shifted. In fact, relation (7) allows a direct com-

parison with Family-Vicsek relation (3): recalling that the scaling behaviors of  $W_2$  and  $t_2$  are given by relations (2), we immediately see that we recover the Family-Vicsek relation by setting  $\lambda=1$ . This nicely shows that it is the isotropic rescaling in the log-log plot by the factor of  $\frac{1}{\log(W_2/W_1)}$  that ultimately is responsible for the success of the scaling relation.

Competitive growth models have the peculiar feature that at every deposition one has to decide which of the two deposition rules is followed by the newly added particle. We propose in the following a deposition model with similar features as the competitive growth models, but where the competition is intrinsic and governed by the value of the substrate temperature. This is a much more realistic situation, especially since in the growth of thin films and nanostructures the substrate temperature is an important parameter that shapes to a large extent the morphology of growing structures [26].

The deposition model discussed in the following is based on the original RDSR process of Family [7] and differs from this model by the diffusion step. In the RDSR process a particle deposited on the surface is allowed to jump to one of the neighboring sites if this site has a lower height than the site of deposition. In our model we assign an energy  $E(\mathbf{x},t)=fh(\mathbf{x},t)$  to the column  $\mathbf{x}$ , where  $h(\mathbf{x},t)$  is the height of that column at time  $t$ . The constant  $f$  can be thought to be the gravitation constant. Starting from an initially flat substrate, particles are deposited on randomly chosen sites and then allowed to diffuse locally after deposition. For a diffusion step taking place at time  $t$ , we select one of the neighboring sites  $\mathbf{y}$  at random and accept the jump with the temperature- and time-dependent (Metropolis-like) probability

$$P_{\mathbf{x}\rightarrow\mathbf{y}}(T,t) = \begin{cases} 1 & \text{if } E(\mathbf{y},t) \leq E(\mathbf{x},t) \\ e^{-[E(\mathbf{y},t)-E(\mathbf{x},t)]/k_B T} = e^{-f[h(\mathbf{y},t)-h(\mathbf{x},t)]/k_B T} & \text{otherwise.} \end{cases}$$

In the following we choose units so that  $f/k_B=1$ , where  $k_B$  is the Boltzmann constant.

In contrast to the original RDSR model, we have in the present model a nonvanishing probability that a deposited particle jumps to a neighboring site with a higher height than the deposition site. We assume this jump to be thermally activated and to depend on the temperature  $T$  of the substrate. As we discuss elsewhere [27], the substrate temperature is a parameter that allows the study of novel properties of growing interfaces.

In Fig. 4(a) we show the temporal evolution of the width for various temperatures and system sizes. As for the RDSR process one distinguishes three regimes separated by two crossover points: a RD regime, followed by a EW regime, with a final crossover to the saturation regime. In contrast to the RDSR process, the random deposition process is not confined to the very-early-time regime  $t \leq 1$  but can extend to

larger times. In fact, the crossover time  $t_1$  between the random deposition and the EW regimes is shifted to higher values for increasing temperatures and diverges in the limit of infinite temperatures. Of special interest is that the surface widths shown in Fig. 4(a) can be directly obtained from Edwards-Wilkinson equation (1) for a system of size  $L$  with a temperature-dependent diffusion constant. For example, for the temperatures shown in Fig. 4(a), the values of  $\nu$  are  $\nu(T=100)=0.005$ ,  $\nu(T=10)=0.04$ , and  $\nu(T=1)=0.18$ .

We first check in Fig. 4(b) that also for the present model scaling relation (5) yields the full data collapse. Due to the simplicity of the model, we can obtain the full information on the location of the two crossover points [27]. In this way we find that, as usual, only the crossover to the saturation regime depends on the system size. In addition, the coordinates of both crossover points display a linear dependence on the substrate temperature. Taking these observations into ac-

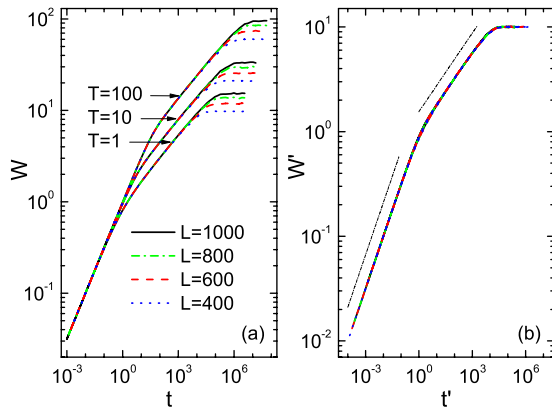


FIG. 4. (Color online) (a) Log-log plot of the surface width vs time for the temperature-dependent deposition model. Systems of different sizes at different temperatures are shown. (b) A complete data collapse of all data sets is also achieved for this model. The dashed lines indicate the expected slopes in the RD and EW regimes.

count, we can rewrite scaling relation (5) in the form

$$W^\lambda = (a + bT)^{\lambda/2} G \left[ \frac{t^\lambda}{(a + bT)^\lambda} \right], \quad (8)$$

where  $a=0.59$  and  $b=0.29$ , with  $\lambda(L)=1/\log(cL^\alpha)$  and  $c=0.55$ , whereas  $\alpha=1/2$  is the roughness exponent of the EW universality class. Equation (8) directly reveals for our model the dependence of the generalized scaling relation on the system size and on the temperature.

In conclusion, we have presented in this paper a parameter-free scaling relation that yields a complete data collapse for large classes of nonequilibrium growth processes with two crossover points. Examples include all simple growth processes as well as more complex growing interfaces as encountered, for example, in competitive growth systems. A deposition model with temperature-dependent diffusion allows us to discuss the dependence of the scaling relation on the relevant system parameters.

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