Number of branches in diffusion-limited aggregates: The skeleton

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We develop the skeleton algorithm to define the number of main branches \( N_b \) of diffusion-limited aggregation (DLA) clusters. The skeleton algorithm provides a systematic way to remove dangling side branches of the DLA cluster and has successfully been applied to study the ramification properties of percolation. We study the skeleton of comparatively large (\( \approx 10^6 \) sites) off-lattice DLA clusters in two, three, and four spatial dimensions. We find that initially with increasing distance from the cluster seed the number of branches increases in all dimensions. In two dimensions, the increase in the number of branches levels off at larger distances, indicating a fixed number of \( N_b = 7.5 \pm 1.5 \) main branches of DLA. In contrast, in three and four dimensions, we find indications that the skeleton continues to ramify as one proceeds from the cluster center outward, and there may not exist a constant number of main branches. Likewise, we find no indication for a fixed \( N_b \) in a study of DLA on the Cayley tree, the limit of “infinite dimensions.” In two dimensions, we encounter strong corrections to scaling of logarithmic character, which can help to explain recently reported deviations from self-similar behavior of DLA.

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I. INTRODUCTION

The fractal nature of diffusion-limited aggregation (DLA) has been established by a number of studies, and a theoretical framework is being developed [1,2]. In the initial paper written by Witten and Sander [1], power-law behavior of the two-point correlation functions was found. Later, Meakin [3] observed power-law scaling of the radius of gyration of DLA in two (2D), three (3D), and four dimensions (4D). Enhanced on- and off-lattice algorithms facilitated studies of larger aggregates (Meakin et al. [4] and Ball and Brady [5]) and a striking influence of the lattice structure on the structure of the aggregates was found [6]. Advanced off-lattice simulations by Ossadnik [7] grew clusters up to \( M = 5 \times 10^7 \). All these studies indicate or confirm a fractal dimension of \( d_f = 1.715 \pm 0.004 \) as reported by Tolman and Meakin [8]. Apart from the scaling of the radius of gyration with cluster mass, fractality is revealed in studies of the branching structure of the aggregates by, e.g., Alström, Trunfio, and Stanley [9], Hinrichsen et al. [10], Ossadnik [11], and more recently Yekutieli, Mandelbrot, and Kaufman [12].

Self-similarity has also been found in a study by Argoul et al. [13] calculating the moments of the mass distribution of on-lattice clusters [13], by Barabási and Vicsek’s mapping of the cluster perimeter to a self-affine surface [14], in investigations of the fjord geometry of DLA [15,16], and many more (for reviews, see, e.g., [17–22]).

However, almost as old as the model itself are reports about deviations from a simple statistically self-similar structure. In Refs. [5,23] lattice anisotropy and deposition habit effects on the structure of lattice clusters were studied, Meakin and Vicsek [24] found that the tangential and radial two-point-correlation functions in DLA display different exponents. In recent works, however, Hegger and Grassberger [25] point out that off-lattice DLA in strip geometry displays a slow approach to local isotropy. Differences arise in the fractal dimensions measured in circular or strip geometry [26]. Simple self-similar scaling is not sufficient to fully characterize the mass distribution within a DLA cluster, as recognized by Meakin and Havlin [27], Vicsek, Faimly, and Meakin [28], Amitrano, et al. [29], and Mandelbrot [30]. Mandelbrot examined the lacunarity [31,32] of off-lattice DLA and found a trend towards increasing compactness for increasing systems size, which is not reflected in the measurements of the radius of gyration of the clusters.

In our study we will apply the “skeleton” algorithm [33,34] to calculate the number of main branches \( N_b \) in off-lattice DLA (Sec. II). We obtain the skeleton of a DLA cluster by removal of all the small dangling side branches of the cluster (we define the skeleton in the following section). Only the main branches that reach out to a certain fraction of the cluster radius remain. For a self-similar, scale-invariant structure we expect that the large-scale properties, like the number of main branches of the cluster, are not altered under rescaling. Conversely, we take an increase or decrease of the number of main branches with increasing cluster size to suggest that the structure is not self-similar.

In 2D, we find that significant finite-size effects of logarithmic nature are present that may be related to some
of the observations reporting deviations from simple self-similar behavior. Asymptotically, however, the number of main branches of 2D DLA approaches a constant number (Sec. III). In contrast to the behavior in 2D, the number of branches of 3D and 4D DLA increases with cluster mass, indicating that these clusters may not be self-similar (Sec. IV).

II. THE SKELETON

DLA has a loopless tree structure [35]. This observation is apparent in simulations of off-lattice aggregates [36,37] where an incoming particle is added to the cluster when its distance to the cluster is below a specific sticking distance. Its “parent” particle is the one that it was closest to at the moment of incorporation into the cluster. The child-parent relationship allows us to uniquely assign a generation number or “chemical distance from the seed” to every cluster particle. To this end, we assign to the seed particle of the DLA cluster the number \( \ell = 0 \). The children of the seed are assigned the chemical distance \( \ell = 1 \); children of \( \ell = 1 \) particles have chemical distance \( \ell = 2 \). In general, each particle inherits the chemical distance of its parent and increases this number by one to find its own. The chemical distance turns out to be a very useful quantity in studies of the branching properties of DLA [9–11] and we will use it here to define the skeleton of DLA — closely following Ref. [33].

Roughly speaking, we obtain the skeleton by removal of all the small dangling side branches of the cluster. To this end we first identify all the sites that are tips of branches — those sites to which no other particle has been attached during the growth process. The \( \ell \) value of each tip \( \ell_{\text{tip}} \) is then passed back to its parent and grandparent and so forth until we hit an ancestor site that has more than one child. In this case, only the largest of the \( \ell_{\text{tip}} \) values of the children is retained. We iterate this process until all cluster sites “know” their \( \ell_{\text{tip}} \) value.

Now let us choose an arbitrary value \( \ell_c \) less than the largest chemical distance \( \Lambda \) of any site from the seed. Given \( \ell_c \), we define as the skeleton those sites for which both \( \ell_{\text{tip}} \geq \ell_c \) and \( \ell \leq \ell_c \) holds. Thus, side branches that have not grown out to at least \( \ell_c \) do not contribute to the skeleton.

In Fig. 1, we display the skeletons of two randomly selected growing DLA clusters at mass \( M = 5000, 50000, \) and \( 500000 \). In this figure, \( \ell_c \) is chosen to be half the “chemical radius” \( \Lambda/2 \) of the cluster. Note that the termination points of the skeleton are located almost on a circle, although we use the chemical and not the Euclidean distance from the seed to define the skeleton. This indicates that DLA grows radially outward without forming loops. Indeed, it has been found that the fractal dimension \( d_{\text{min}} \) of the path connecting the cluster seed to a given site is equal to 1 [35]. Here, \( d_{\text{min}} \) is defined by \( \ell \sim r^{d_{\text{min}}} \). In other words, for cluster sites at a given distance \( r \) from the cluster seed, the average value \( \langle \ell \rangle \) is proportional to \( r \), \( \langle \ell \rangle \sim r \). For the remainder of this study, we will consider \( \ell \) and \( r \) as equivalent.

One physical interpretation of the skeleton in DLA can be obtained if we consider the aggregate as a conductor situated between the grounded seed and a circular electrode (a sphere or hypersphere in 3D or 4D, respectively) of “radius” \( \ell_c \). Then the skeleton is the collection of paths that contribute to the current through the aggregate.

The skeleton as defined above has some desirable and some less desirable features. First, its definition is certainly simple, which is desirable. Second, we see in Fig. 1 that in an intermediate range of \( \ell \) the number of branches in the skeleton is small, which is another desirable feature since we claim that the skeleton identifies the main branches of the DLA cluster. However, we see that close to \( \ell_c \) a lot of branches appear. Their presence ultimately reflects our ignorance as to whether a specific side branch will keep growing or die out. Thus, these branches are a very “physical” property, but, nevertheless, an undesirable feature. We will address this problem by keeping \( \ell_c \)

![FIG. 1. Skeleton for two growing DLA clusters: (a)–(c) First cluster; (d)–(f) second cluster. Growth of the cluster has been interrupted at cluster masses \( M = 5000 \) (a),(d), 50000 (b),(e), or 500000 (c),(f). The determination of the skeleton is based on a value of \( \ell_c/\Lambda = 0.5 \). The skeleton is then rescaled so that, independent of cluster mass, the same size results.](image-url)
a variable and systematically test the dependence of the skeleton calculations on this parameter.

We define $\lambda \equiv \ell_c/\Lambda$ and consider ensembles of DLA clusters for which the value $\Lambda$ is fixed. For such an ensemble, we define $B(\lambda; \Lambda; \ell)$ as the average number of sites in the skeleton at chemical distance $\ell$ from the cluster seed. Note that $B(\lambda; \Lambda; \ell)$ is defined only for $\ell \leq \ell_c$.

Some general conclusions about properties of the DLA skeleton can be drawn from the definition and are visible in Fig. 1: (i) For $\ell = 0$, only the cluster seed contributes, thus $B(\lambda; \Lambda; \ell = 0) = 1$ independent of spatial dimension $d$ and cluster size $\Lambda$. (ii) Since $\ell \sim r$, and since DLA is essentially loopless so that overhangs can be neglected, we can determine the value $B(\lambda; \Lambda; \ell = \ell_c)$ from the number of intersection points of a “cookie cutter” of radius corresponding to $\ell_c$ with the DLA cluster. As the analog of a cookie cutter in 3D we consider a sphere, and in 4D a hypersphere. The result is $B(\lambda; \Lambda; \ell_c) \sim \ell_c^{d_f-1}$ where $d_f$ denotes the fractal dimension of DLA, which depends on the embedding dimension $d$. The exponent $d_f - 1$ follows from the fact that the codimension of the intersection of two sets is obtained as the sum of the codimensions of the two intersecting sets. (iii) The number of sites in the skeleton increases monotonically with $\ell$, i.e., $B(\lambda; \Lambda; \ell') \geq B(\lambda; \Lambda; \ell)$ if $\ell' \geq \ell$. Qualitatively we observe from Fig. 1 that the increase is slow in the vicinity of the cluster seed, but is followed by a precipitous increase close to $\ell_c$. (iv) For fixed $\ell$ and cluster “radius” $\Lambda$, the number of sites in the skeleton decreases monotonically when $\lambda$ is increased, because some branches that contribute to the skeleton at some value of $\ell_c$ fail to reach larger values. In the extreme case, when $\lambda = 1$, we obtain $B(1, \Lambda; \ell) = 1$, independent of $\Lambda$ and $\ell$.

In the study of the skeleton of percolation clusters [33], a slow increase of the number of arms for small $\ell$ was found to be due to finite-size corrections. Taking this into account, one obtains for small $\ell$ a constant number of branches. In analogy to the percolation example, we use finite-size scaling techniques to extract information about DLA from our 2D, 3D, and 4D skeleton calculations, which we present in the following sections. The DLA results in finite dimensions will be compared to the skeleton of a random Cayley tree model, which can be considered to be the limit of infinite dimensionality of DLA [38].

### III. SKELETON OF 2D DLA

In Fig. 2 we display $B(\lambda = 0.5; \Lambda; \ell)$ as a function of $\ell/\ell_c$ for $\Lambda = 100, 316, 1000, 3162,$ and $5000$. For $\Lambda \leq 1000$, we obtain our data from 500 off-lattice clusters,

![Log-log plot of the 2D skeleton $B(\lambda = 0.5; \Lambda; \ell)$ vs $\ell/\ell_c$ for different $\Lambda$ values as indicated in the legend. For $\Lambda \leq 1300$ data are averaged over 850 off-lattice DLA, for $\Lambda > 1300$ averages over 114 clusters are taken. For comparison, we have also averaged data in the constant $M$ ensemble, instead of the constant $\Lambda$ ensemble used here, and obtain similar results (not shown). Note that in an intermediate region of the plot ($\ell/\ell_c < 0.3$) a slope can be associated with $B(\lambda = 0.5; \Lambda; \ell) \approx \ell/\ell_c$. As $\Lambda$ increases, this slope becomes increasingly smaller, indicating that the asymptotic curve may be flat in this “scaling” region. Inset: The number of branches in the skeleton at $\ell = \ell_c$, $B(\lambda; \Lambda; \ell_c)$ is equal to the number of intersection points of a cookie cutter of radius $\ell_c$ in $\ell$ space with the cluster, i.e., $B(\lambda; \Lambda; \ell_c) \sim \ell_c^{d_f-1}$. The solid line is a guide to the eye and indicates a slope of $d_f - 1 = 0.79$. Different symbols refer to different $\lambda: 0.8 (~C), 0.6 (~□), 0.5 (~∆), 0.4 (~★),$ and $0.2 (~★)$.


for $\Lambda \geq 3162$, 114 clusters are averaged. The clusters are grown stepwise and growth is interrupted for the analysis when $\Lambda$ equals one of the above listed values. For $\Lambda = 5000$, the cluster mass $M$ is typically $M = (1 - 1.5) \times 10^6$ sites.

As a function of $\ell/\ell_c$ we find that after an initial short transient the central part of the data displays approximately linear behavior. The linear region widens when the cluster size $\Lambda$ increases. The approximate linear behavior terminates when $\ell/\ell_c$ approaches $0.3$. Beyond $\ell/\ell_c \approx 0.3$, $B(0.5, \Lambda; \ell)$ increases sharply towards the final value $\sim \ell_{\Lambda}^{2/3}$. However, the linear region of the data tends to become flatter as we proceed to larger $\Lambda$. Since $B(\Lambda, \lambda)$ is monotonically increasing, the slope $\alpha(\lambda, \Lambda)$ in the linear region cannot be negative. Thus, asymptotically $\alpha(\lambda, \Lambda)$ can either approach 0 or assume a finite value $> 0$. We will refer to the restricted range of approximately linear behavior as the "scaling regime" of the skeleton.

In order to distinguish between these two possibilities, we examine the dependence of $\alpha(\lambda, \Lambda)$ on system size, here characterized by $\Lambda$. To this end, we determine the slope $\alpha(\lambda, \Lambda)$ of a linear least-squares fit,

$$\ln B(\lambda, \Lambda; \ell) \approx \alpha(\lambda, \Lambda) \ln \left( \frac{\ell}{\ell_c} \right) + \beta(\lambda, \Lambda),$$

(1)

to $\ln B(\lambda, \Lambda; \ell)$ vs $\ln(\ell/\ell_c)$ for data points satisfying $\ell \geq 4$ and $\ell/\ell_c \leq 0.3$. We use the assumption, familiar from the estimate of critical exponents from finite size tests calculations (see, e.g., [39]), that $\alpha(\lambda, \Lambda)$ may be written as

$$\alpha(\lambda, \Lambda) = \alpha_\infty(\lambda) + f(\lambda, \Lambda) + \cdots.$$  

(2)

Here, the leading correction term $f(\lambda, \Lambda)$ to the asymptotic value $\alpha_\infty(\lambda)$ is typically either (i) a power law $\sim \Lambda^{-\gamma}$ or (ii) a logarithmic term, e.g., $\sim 1/\ln \Lambda$. Both vanish for $\Lambda \to \infty$. The dots indicate corrections that decrease faster than $f(\lambda, \Lambda)$ as $\Lambda \to \infty$.

In Fig. 3(a) we plot $\alpha(\lambda; \Lambda)$ vs $1/\ln \Lambda$ for different values of $\lambda$. From the almost linear behavior of $\alpha(\lambda; \Lambda)$ for large $\Lambda$ we conclude that the leading-order correction term can be approximated by the reciprocal of $\ln \Lambda$ (ii), i.e.,

$$f(\lambda, \Lambda) = A(\lambda)/\ln \Lambda.$$  

(3)

Here, $A(\lambda)$ denotes an amplitude that depends on $\lambda$ as evident from Fig. 3(a). However, our calculations do not allow us to exclude the possibility that $f(\lambda, \Lambda)$ vanishes in the power-law fashion (i), although, in this case the characteristic power $\gamma$ must be small, $\gamma < 0.2$. Larger values of $\gamma$ would lead to significant curvature in the large $\Lambda$ behavior of $\alpha(\lambda, \Lambda)$ [cf. Fig. 3(b)].

From Fig. 3(a) we conclude that $\alpha_\infty(\lambda) = 0$ for $0.3 \leq \lambda \leq 0.8$ by extrapolating the trend displayed by $\alpha(\lambda, \Lambda)$ for the $\Lambda$ range accessed by our simulation. Since for $\Lambda \to 1$, $B(\Lambda, \lambda; \ell) = 1$ for all $\ell$, we expect this result to also hold if $\lambda > 0.8$. For small $\lambda$, the skeleton probes the frozen region of the DLA cluster which effectively stopped growing [29], so that our results should be less affected by possible slow structural changes in the growth zone of the cluster [16,40,41]. Therefore, although the fit region of $B(\lambda, \Lambda; \ell)$ becomes narrower, we assume that $\alpha_\infty(\lambda) = 0$ extrapolates into the small $\lambda$ regime.

The implications of $\alpha_\infty(\lambda) = 0$ are intriguing. Asymptotically, we can assign a fixed number of main branches $N_8$ to 2D DLA growth. We deduce the actual number of these branches from the behavior of the intercepts $\beta(\lambda; \Lambda)$ of the fit (1) in the "flat" region of the skeleton. In Fig. 3(c), $\beta(\lambda; \Lambda)$ is plotted as a function of $1/\ln \Lambda$ for several $\lambda$. The value of $\beta(\lambda; \Lambda)$ slightly increases as the cluster size grows — corresponding to more main branches in larger systems. However, the increase levels off at about $\beta_\infty(\lambda) = 2 \pm 0.2$, so that on average $N_8 = \exp[\beta_\infty(\lambda)] = 7.5 \pm 1.5$ main branches result.

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**FIG. 3.** Finite-size behavior of the slopes $\alpha(\lambda, \Lambda)$ of $B(\lambda, \Lambda; \ell)$ vs $\ell/\ell_c$ calculated from data in the range $\ell \geq 4$ and $\ell/\ell_c < 0.3$. The slopes $\alpha(\lambda, \Lambda)$ for different $\lambda$ [0.8 (O), 0.7 (□), 0.6 (△), 0.5 (▼), 0.4 (●), 0.3 (■)] are plotted vs (a) $1/\ln \Lambda$ and (b) $1/\Lambda$. In (a) the slopes lie on asymptotically straight lines indicating a finite-size correction term $\sim 1/\ln \Lambda$. This behavior should be contrasted to the calculations in (b) where the correction term is assumed to be proportional to $1/\Lambda$ and pronounced curvature is displayed as $\Lambda$ increases. (c) Intercepts $\beta(\lambda; \Lambda)$ of the straight line fit with the ordinates vs $1/\ln \Lambda$. If $\alpha_\infty(\lambda) = 0$, then $\beta_\infty(\lambda)$ denotes the logarithm of the number of arms $N_8$ in the "flat" region of the skeleton. From $\beta_\infty \approx 2 \pm 0.2$ we conclude that $N_8 = 7.5 \pm 1.5$.  

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IV. DLA IN HIGHER DIMENSIONS: 3D AND 4D

We also perform simulations of off-lattice DLA in 3D and 4D. Our findings are quite different from the 2D case, a result not surprising in light of earlier work suggesting that 2D is a special case [42]. In particular, our calculations suggest that in contrast to the 2D case, where $\alpha_\infty(\lambda) = 0$, in the 3D and 4D case $\alpha_\infty(\lambda) > 0$. We discuss in the following the numerical data supporting this result.

A. Skeleton of 3D DLA

In Fig. 4 we display $B(\lambda = 0.5, \Lambda; \ell)$ for 3D DLA. Different curves are characterized by different values of $\Lambda$. The largest value of $\Lambda$ is 560, which corresponds to DLA clusters with $M \approx 10^6$ particles. The skeletons of 450 ($\Lambda \leq 200$) and 75 ($\Lambda > 200$) clusters have been averaged to obtain the displayed data.

The tendency for the central part of the plot to become flatter as $\Lambda$ increases is much less pronounced than in 2D. In fact, by eye there is only a very small change of the slope visible as we increase the cluster size. We take this as one indication that the slope may actually remain finite in the asymptotic limit.

To some extent our observation is supported by the finite-size scaling plots Figs. 5(a) and 5(b), which show the $\Lambda$ dependence of the slope $\alpha(\lambda; \Lambda)$. Similar to 2D, we obtain the $\alpha(\lambda; \Lambda)$ from a linear least-square fit of $\ln B(\lambda, \Lambda; \ell)$ in the region characterized by $\ell \geq 4$ and $\ell/\ell_\infty < 0.3$. As visible in these plots, the functional form of the finite-size correction terms in 3D is much less clear than in 2D. Although, as in 2D, for all values of $\lambda$ in Fig. 5(a) some region of linear behavior is visible, the extrapolation to infinite systems is inconclusive; whereas the curves for large $\lambda$ values (open symbols) seem to extrapolate to values of $\alpha_\infty(\lambda)$ in the range 0.05–0.3, smaller $\lambda$ values indicate $\alpha_\infty(\lambda) \leq 0$. Since $\alpha_\infty(\lambda)$ must be positive, it seems possible that the actual finite-size corrections are stronger than $\sim 1/\ln(\Lambda)$, e.g., vanishing in power-law fashion for $\Lambda \to \infty$.

In contrast to 2D, there is a more significant region of straight line behavior apparent in the plot of $\alpha(\lambda; \Lambda)$ vs $1/\Lambda$ [Fig. 5(b)]. If we here assume that the finite-size corrections are $\sim 1/\Lambda$ then the finite-size behavior rather indicates that $\alpha_\infty(\lambda) > 0$ for all $\lambda$.

From the simulation data of 3D DLA we have thus collected three indications that in 3D the asymptotic values of $\alpha_\infty(\lambda)$ are larger than zero: (i) the very slow change in the slopes of $B(\ell, \Lambda; \lambda)$, (ii) finite-size scaling extrapolations of $\alpha(\lambda; \Lambda)$ to positive values for small $\lambda$ (the behavior of the other $\lambda$’s are pointing to a stronger than logarithmic decay to zero of the finite-size corrections), and (iii) for an assumed $\Lambda^{-1}$ dependence all $\alpha(\lambda; \Lambda)$ extrapolate to possibly different, but nonzero positive values. We have made equivalent plots for correction exponents between 0 and –1 and we have not found a single convincing value yielding asymptotically linear behavior for all $\lambda$ values. In particular for small $\lambda$, where the statistics are worse due to the smaller extension of the scaling regime, larger clusters are required to reach asymptotic behavior. Some observations on other systems, which

![Figure 4](image_url)

**FIG. 4.** Log-log plot of the 3D skeleton $B(\lambda = 0.5, \Lambda; \ell)$ vs $\ell/\ell_\infty$ for different $\Lambda$ values as indicated in the legend. We take averages over 75 ($\Lambda > 200$) and 450 ($\Lambda \leq 200$) clusters. Note that the tendency of the power-law region in the plot to become flatter with increasing $\Lambda$ is much less pronounced than in the 2D case 2 and is in fact rather indicating a nonzero $\alpha_\infty(\lambda) > 0$. Inset: Number of branches in the skeleton at $\ell = \ell_\infty$. We expect $B(\lambda, \Lambda; \ell_\infty) \approx \Lambda^{d_f - 1}$. The solid line is a guide to the eye and indicates a slope of $1.5 \approx d_f - 1$ in 3D. Different symbols refer to different $\lambda$ and are the same as in Fig. 2.
will be discussed in Secs. V and VI, also indicate that in 3D $\alpha_\infty(\lambda)$ could assume values larger than zero.

The significance of such positive values is that in 3D (in contrast to the 2D behavior) the number of branches of the skeleton increases monotonically in power-law fashion as a function of $\ell/\ell_c$, even in the limit of infinite cluster size. However, like in 2D, very strong corrections to the asymptotic behavior are present if we consider finite systems. This is certainly a surprising result when we compare it to, e.g., the behavior of the scaling of the radius of gyration data for DLA [36], which shows a very clean scaling behavior without significant finite-size corrections.

**B. Skeleton of 4D DLA**

The skeletons of 50 off-lattice DLA clusters were averaged to obtain the data displayed in Fig. 6. The clusters with $\Lambda = 190$ contain up to $\approx 500,000$ particles. Similar to the 3D case, the slopes in the central part of the $B(\lambda, \Lambda; \ell)$ data do not tend to become flat with increasing mass and this behavior is reflected in the $\Lambda$ dependence of $\alpha(\lambda; \Lambda)$, which we display in Figs. 7(a) and 7(b). The quality of the 4D calculations is much poorer than that of the 3D and 2D calculations. However, a value (or different values) $\alpha_\infty(\Lambda)$ larger than zero is consistent with the two finite-size scaling plots Figs. 7(a) and 7(b). As in 3D, such a positive value indicates a power-law branching of the skeleton in the scaling regime as the cluster mass goes to infinity.

**FIG. 6.** Log-log plot of the 4D skeleton $B(\lambda = 0.5, \Lambda; \ell)$ vs $\ell/\ell_c$ for different $\Lambda$ values as indicated in the legend. As in the 3D case, the central region of the plot does not display a tendency to become flat as $\Lambda$ increases. However, the system sizes seem still too small to draw a final conclusion, in particular if one considers the size dependence of the slopes in the power-law region and the comparatively poorer quality of the calculations (see Fig. 7). Inset: Number of branches in the skeleton at $\ell = \ell_c$. We expect $B(\lambda, \Lambda; \ell_c) \sim \Lambda^{d_f - 1}$. The solid line is a guide to the eye and indicates a slope of $2.2 \approx d_f - 1$ in 4D. Different symbols refer to different $\lambda$ and are the same as in Fig. 2.
disregard screening of the interior parts of the tree, which in finite dimensions is the essential mechanism to produce low-dimensional fractal structures [43]. We have numerically grown clusters of this tree model with functionality $z = 3, 4$, and $5$ as large as $32$ shells. As a function of the shell number, the number of occupied sites grows exponentially [38]. The skeleton shows no sign of saturation as a function of $\ell / \ell_c$. The increase seems to have even exponential character, with $\ln B(\lambda, \Lambda = 32; \ell) \propto \ell / \ell_c$, and prefactors that depend on $z$ and $\lambda$.

We are thus confronted with two extreme situations. On the one hand, in 2D DLA, where screening is very important, we observe a constant number of branches. On the other hand, without screening the skeleton grows even exponentially as a function of $\ell / \ell_c$. We may thus speculate that if in some finite dimension the screening becomes sufficiently weak there will be a transition from nonbranching to branching behavior. We argue here that this transition occurs from 2D to 3D, albeit the branching behavior has power-law rather than exponential character.

### VI. DISCUSSION

#### A. Previous results on the number of stable branches of DLA

The number of branches $N_b$ of DLA has been used as an important morphological characterizaton of the cluster [4,44–49]. Here, we want to briefly discuss the consequences of our findings with respect to previous results. In 2D using conformal mapping arguments, Ball [46] has proposed that the maximum number of stable branches $N_{\text{max}}$ is related to the mass scaling dimension $D$ of the cluster by

$$N_{\text{max}} / 2 - 1 | (D - 1) = 1.$$  \hspace{1cm} (4)

For $D \approx 1.7$ this relation predicts $N_{\text{max}} \approx 4.9$. Similarly, Turkевич and Scher [44] have modeled the tips of the cluster as the corners of a convex polygonal cluster envelope. The number $N$ of branches then determines the tip angles and those in turn the mass scaling dimension, namely, $D - 1 = N / (N + 2)$. If one replaces $D - 1$ from Eq. (4) and assumes $N = N_{\text{max}}$, we obtain $N_{\text{max}} \approx 4.8$, in agreement with the previous prediction.

Both numbers seem to be significantly smaller than our finding $N_b = 7.5 \pm 1.5$ in 2D. (Turkevich-Scher have pointed out that there exists a relation between the fractal dimension of the cluster, which finds its expression in the scaling behavior of the tip growth probability and the geometric shape of the absorber. Of course there can be 7 absorbers of the same shape, but then the Turkevich-Scher assumption of a simple convex polygonal envelope of the cluster has to be given up. This is a necessary consequence of our findings.) It is not really clear to us why $N_b$ is different from $N_{\text{max}}$ following from Eq. (4). On the one hand, if the actual number of branches is larger than the maximum number of stable ones then the competition among branches for the incoming flux of random walkers.

**FIG. 7.** Finite-size behavior of the slopes $\alpha(\lambda; \Lambda)$ of $B(\lambda = 0.5, \Lambda; \ell) \sim \ell / \ell_c$, calculated from data in the range $\ell > 4$ and $\ell / \ell_c < 0.3$ for 4D DLA. The slopes $\alpha(\lambda, \Lambda)$ for different $\lambda$ with symbols as in Fig. 3 are plotted vs (a) $1 / \ln \Lambda$ and (b) $1 / \lambda$. (c) Intercepts $\beta(\lambda, \Lambda)$ of the straight line fit with the ordinate vs $1 / \ln \Lambda$.

Moreover, if we consider the intersections $\beta(\lambda, \Lambda)$ of the straight line fits to the scaling region we observe that they may diverge as $\Lambda \to \infty$. Such behavior would imply that $B(\lambda, \Lambda; \ell)$ may not converge to a limit as $\Lambda \to \infty$. Thus we expect rather strong deviations from self-similar behavior in 4D and for $\Lambda \to \infty$ no self-similar limit as in the 2D case may exist.

### V. CAYLEY TREE MODEL

The realization of the DLA growth model on a Cayley tree is considered to be the mean-field or infinite dimensionality limit of DLA. DLA on the Cayley tree has been formulated as follows [38]. To the cluster seed (shell 0), we link $z$ surface sites (shell 1) $- z \geq 2$ is the functionality of the model. In each growth step, we select (occupy) one of the surface sites randomly. We then create $z - 1$ new (empty) surface sites and link them to the just occupied site. The shell number of the new empty sites is equal to the shell number of the occupied parent incremented by 1. Note that all occupied sites have $z$ links to neighbor sites and that all empty sites have exactly 1 neighbor site.

Since new cluster sites are randomly picked from the set of all surface sites of the growing tree, we effectively
will cause branches to die (cf. [50]). If, on the other hand, the number of branches is too small, then new branches will be created because the fjords of the aggregate are not sufficiently screened to suppress growth. It is, however, possible that this competition process establishes a number of branches different from (and larger than) the stability limit \( N_{\text{max}} \). Thus the observed number \( N_b \) does not constitute a contradiction to Ball’s arguments.

B. Relation to “lacunarity” measurements on DLA

Recently, Mandelbrot has investigated the lacunarity of 2D off-lattice DLA [30]. To calculate the lacunarity of a cluster of span \( \Lambda \), he (i) selects the frozen interior region of the cluster (of span \( k\Lambda \), \( k < 1 \)) and rescales all the particles coordinates into the unit circle, then (ii) centers solid disks of radius \( \epsilon \ll 1 \) on the rescaled particle positions and (iii) determines the area fraction \( f \) of the unit circle covered by the \( \epsilon \) disks. For a “normal” fractal, one expects \( f \) to saturate at a value less than unity when \( \Lambda \) increases and in this case to provide a measure for the lacunarity of the object. For DLA, however, \( f \) seems to increase continuously with \( \Lambda \) to a value of unity. Mandelbrot argues accordingly that either DLA displays a “massive transient” or a “limitless drift” towards increasing compactness.

We argue that the logarithmic finite-size corrections that we see in 2D in the approach of \( B(\Lambda, \Lambda; \ell) \) to its asymptotics can explain the “massive transient” picture. As we have seen above, \( B(\Lambda, \Lambda; \ell) \) increases with system size corresponding to structurally different clusters with more main branches \( N_b \). Consequently, the area fraction \( f \) increases.

Thus, 2D off-lattice DLA shows a very slow structural change towards its asymptotic structure, similar to its on-lattice siblings.

C. Why is 2D DLA different from higher-dimensional DLA?

The remaining open question is why on the one hand \( \alpha_{\infty}(\lambda) \to 0 \) in 2D, but, on the other hand, \( \alpha_{\infty}(\lambda) > 0 \) in 3D and 4D and in the latter case possibly even a divergence of \( \beta(\lambda, \Lambda) \) with \( \Lambda \). Let us first note that 2D is known to be a “critical dimension” for radial DLA with respect to the scaling properties of the growth probabilities of the surface sites [42,51,52]. We find two more indications supporting our result, one from a simple consideration of the cluster density in different spatial dimensionalities, the other from a comparison of the correlated DLA tree to a random tree model, for which the scaling behavior of the skeleton is known [34].

1. Cluster “density” and screening

We have already noted above that the number of branches in the skeleton at its termination points at \( \ell_c \) is \( \sim \ell_c^{d_f - 1} \). Thus, for given \( \ell = \ell_c \), there are many more branches present in the 3D and 4D skeleton than in the 2D skeleton. In order that these branches contribute to the skeleton in the scaling region they must originate deep in the cluster interior.

With increasing dimension, the screening of the cluster interior from incoming particles decreases. This decrease manifests itself, for example, in the strong semi-exponential screening of the cluster fjords in 2D [53–57] in contrast to the power-law type [42] screening in 3D or in the screening free Cayley-tree model — where we also observe \( \alpha_{\infty}(\lambda) > 0 \). Qualitatively, the reduction of screening can also be seen in the mean-field expression for the particle penetration depth \( \sim \rho^{-1} \sim \Lambda^{d - d_f} \). The penetration depth increases with \( d \), because \( d_f \) approaches \( d - 1 \) from above. (It is, however, known from simulations that at least in 2D this relation does not hold and the penetration depth is \( \sim \Lambda \).)

Therefore we conclude that the weaker screening in dimensions \( d > 2 \) is responsible for the further increase in the number of branches of the skeleton both as a function of \( \ell / \ell_c \) and as a function of cluster size \( \Lambda \).

2. Comparison to a random tree model

DLA has the structure of a tree. Thus, we find it instructive to compare our findings for the off-lattice DLA skeleton to results for the skeleton of a random tree model [34]. This tree model is defined on a 2D square lattice and characterized by a tunable “intrinsic” dimension \( d_f \), which determines how the mass \( M(\ell) \) of the tree object depends on the chemical distance \( \ell \) from the cluster seed, i.e., \( M(\ell) \propto \ell^{d_f} \). To this end, we occupy randomly exactly \( N(\ell) = \ell^{d_f - 1} \) sites in the \( \ell \)th chemical shell (or fewer if necessary) — all the other sites are blocked. Only those sites in the \( \ell \)th shell shall be occupied that do not close loops, i.e., that are not neighbors to sites in previous shells. Note that apart from possible constraints due to the embedding lattice at small \( \ell \), the topological structure of the clusters does not depend on the dimension of the embedding lattice.

Reference [34] then considers the skeleton \( B(1, \Lambda; \ell) \) of a tree with intrinsic dimension \( d_f \). In contrast to DLA, here in general \( B(1, \Lambda; \ell) \geq 1 \), because the last chemical shell contains \( N(\Lambda) = \ell^{d_f - 1} \geq 1 \) sites. In these random tree models there exists a sharp transition value \( d_f^* = 1.65 \pm 0.05 \) which separates two regions with different types of behavior. For trees with \( d_f < d_f^* \) the skeleton displays a flat — approximately constant — region as a function of \( \ell \), like the one observed in the 2D DLA case (nonbranching skeleton). However, for larger values of \( d_f \) the skeleton displays a well-defined power-law increase as a function of \( \ell \), i.e., \( \alpha_{\infty}(1) > 0 \) (branching skeleton).

Thus, if we consider DLA as such a random tree, we expect a branching skeleton for spatial dimension \( d > 2 \), since \( d_f = d_f^* > d_f^* \). In 2D, \( d_f^* \) is only slightly larger than \( d_f^* \), or even equal to \( d_f^* \) within the error bars, which is consistent with the observed nonbranching skeleton.
VII. SUMMARY

We have determined the skeleton of comparatively large (∼ 10^6 sites) off-lattice DLA clusters in 2D, 3D, and 4D. We find that, asymptotically, in 2D the skeleton of DLA suggests a fixed number of N_k ≈ 7.5 ± 1.5 main branches and a self-similar structure. In 3D and 4D, and possibly all spatial dimensions d > 4, the DLA skeleton is a ramified object, which displays branching over the whole range of t values for which it is defined.

For all dimensions we find strong finite-size effects corresponding to a slow change in the structure of DLA as the cluster size increases. The presence of strong corrections to scaling of logarithmic character in 2D DLA is in agreement with findings of deviations from self-similar behavior reported elsewhere.

Moreover, in 4D, it is possible that the structure of the aggregate keeps changing even asymptotically, such that there is no self-similar limit.

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