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Role of dimensionality in complex networks

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Deep connections are known to exist between scale-free networks and non-Gibbsian statistics. For example, typical degree distributions at the thermodynamical limit are of the form $P(k) \propto e_q^{-k/\kappa}$, where the q -exponential form $e_q^z \equiv [1 + (1 - q)z]^{1/(1-q)}$ optimizes the nonadditive entropy S_q (which, for $q \rightarrow 1$, recovers the Boltzmann-Gibbs entropy). We introduce and study here d -dimensional geographically-located networks which grow with preferential attachment involving Euclidean distances through $r_{ij}^{-\alpha}$ ($\alpha_A \geq 0$). Revealing the connection with q -statistics, we numerically verify (for $d = 1, 2, 3$ and 4) that the q -exponential degree distributions exhibit, for both q and k , universal dependences on the ratio α_A/d . Moreover, the $q = 1$ limit is rapidly achieved by increasing α_A/d to infinity.

Networks emerge spontaneously in many natural, artificial and social systems. Their study is potentially important for physics, biology, economics, social sciences, among other areas. For example, many empirical studies have identified peculiar properties in very different networks such as the Internet and online social networks (e.g., Facebook), citations networks, neurons networks¹⁻³, to quote but a few. An ubiquitous class of such networks is constituted by the scale-free ones (more precisely, asymptotically scale-free). As we shall soon verify, these networks can be seen as a particular application of nonextensive statistical mechanics, based on the nonadditive entropy $S_q = k \frac{1 - \sum_i p_i^q}{q-1}$ ($q \in \mathcal{R}$; $S_1 = S_{BG} = -k \sum_i p_i \ln p_i$, where BG stands for *Boltzmann-Gibbs*)⁴⁻⁶. This current generalization of the BG entropy and corresponding statistical mechanics has been widely successful in clarifying the foundations of thermal statistics as well as for applications in complex systems in high-energy collisions at LHC/CERN (CMS, ALICE, ATLAS and LHCb detectors) and at RHIC/Brookhaven (PHENIX detector)⁷⁻¹⁶, cold atoms¹⁷, dusty plasmas¹⁸, spin-glasses¹⁹, trapped ions²⁰, astrophysical plasma^{21,22}, biological systems²³, type-II superconductors²⁴, granular matter²⁵, the Kuramoto model at the edge of chaos²⁶, low-dimensional maps, for instance the (area-preserving) standard map²⁷ (see Bibliography in <http://tsallis.cat.cbpf.br/biblio.htm>). Many other physical situations are described which are analogous, such as long-range-interacting Hamiltonians, for example, gravitational problems like globular clusters, spins systems, like the Ising, XY and Heisenberg long-range models. We may also point out random-walk anomalous diffusion where the jumps obey a power law probability distribution function. Some (naturally not all) of the properties of long-range-interacting systems may be described as forming complex network where the sites are linked according to power-law preferential attachment. In the present work we address a wide class of this kind of problems focusing on some basic universality relations.

The deep relationship between scale-free networks and q -statistics started being explored in 2005²⁸⁻³⁰, and is presently very active³¹⁻³⁵. The basic connection comes (along the lines of the BG canonical ensemble) from the fact that, if we optimize the functional $S_q[P(k)] = k \frac{1 - \int dk [P(k)]^q}{q-1}$ with the constraint $\langle k \rangle \equiv \int dk k P(k) = \text{constant}$ or analogous (k being the degree of a generic site, i.e., the number of links connected to a given site; $P(k)$ denotes the degree or connectivity distribution), we straightforwardly obtain $P(k) = P(0) e_q^{-k/\kappa} = P(0) [1 + (q-1)k/\kappa]^{1/(1-q)}$, which turns out to be the generic degree distribution for virtually all kinds of scale-free networks. The q -exponential function is defined as $e_q^z \equiv [1 + (1-q)z]^{1/(1-q)}$ ($e_1^z = e^z$). We verify that, for $q > 1$ and $k \rightarrow \infty$, $P(k) \sim 1/k^\gamma$ with $\gamma \equiv 1/(q-1)$. The classical result $\gamma = 3$ ³⁶ corresponds to $q = 4/3$.

In the present work we address the question of how universal such results might be, and more specifically, how $P(k)$ varies with the dimension d of the system?

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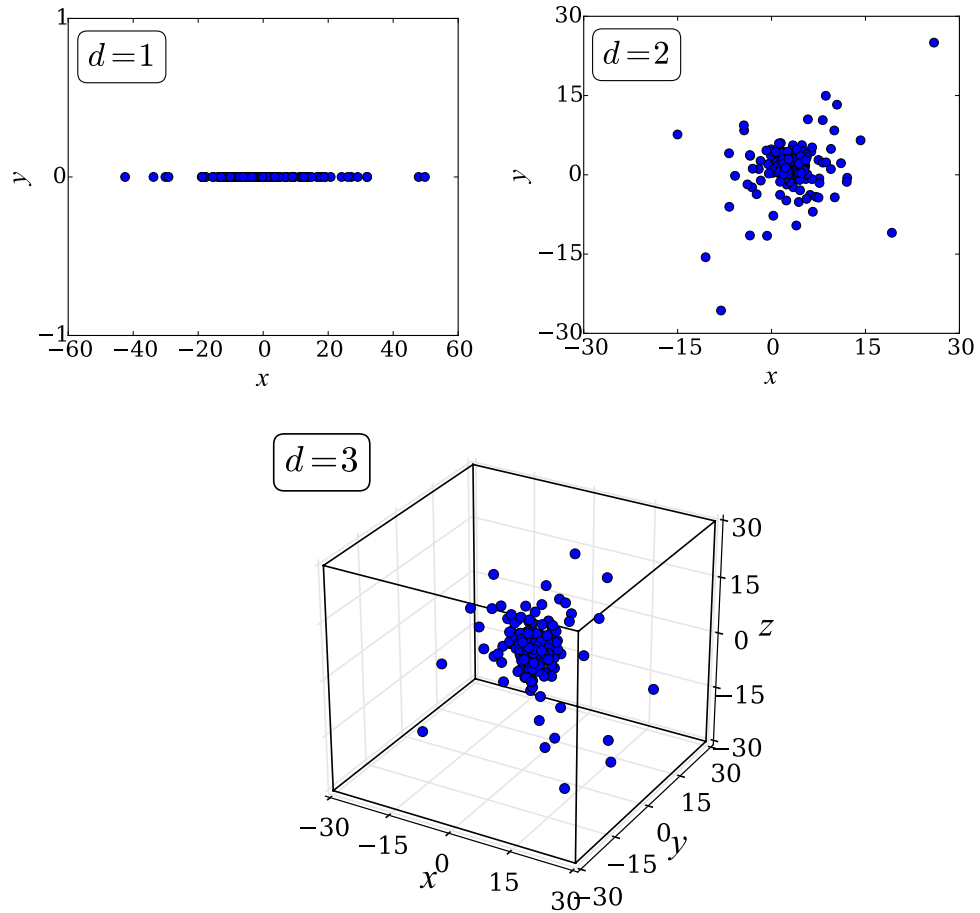


Figure 1. Distribution of $N=500$ sites obtained with Eq. (1) for $\alpha_A=2.0$, $\alpha_G=0.0$, and $d=1, 2, 3$.

Our growing model starts with one site at the origin. We then stochastically locate a second site (and then a third, a fourth, and so on up to N) through the d -dimensional isotropic distribution

$$p(r) \propto \frac{1}{r^{d+\alpha_G}} \quad (\alpha_G > 0; d = 1, 2, 3, 4), \tag{1}$$

where $r \geq 1$ is the Euclidean distance from the newly arrived site to the center of mass of the pre-existing system (in one dimension, $r=|x|$; in two dimensions, $r = \sqrt{x^2 + y^2}$; in three dimensions $r = \sqrt{x^2 + y^2 + z^2}$, and so on); we assume angular isotropy; $p(r)$ is zero for $0 \leq r < 1$; the subindex G stands for *growth*. We consider $\alpha_G > 0$ so that the distribution $P(r)$ is normalizable; indeed, $\int_1^\infty dr r^{d-1} r^{-(d+\alpha_G)} = \int_1^\infty dr 1/r^{1+\alpha_G}$, which is finite for $\alpha_G > 0$, and diverges otherwise. See Fig. 1.

Every new site which arrives is then attached to one and only one site of the pre-existing cluster. The choice of the site to be linked with is done through the following preferential attachment probability:

$$\Pi_{ij} = \frac{k_i r_{ij}^{-\alpha_A}}{\sum_i k_i r_{ij}^{-\alpha_A}} \in [0, 1] \quad (\alpha_A \geq 0), \tag{2}$$

where k_i is the connectivity of the i -th pre-existing site (i.e., the number of sites that are already attached to site i), and r_{ij} is the Euclidean distance from site i to the newly arrived site j ; subindex A stands for *attachment*.

For α_A approaching zero and arbitrary d , the physical distances gradually lose relevance and, at the limit $\alpha_A = 0$, all distances become irrelevant in what concerns the connectivity distribution, and we therefore recover the Barabási-Albert (BA) model³⁶, which has topology but no metrics. The BA model was extended³⁷ in such a way that it would be able to yield an exponent γ such that $2 < \gamma < 3$, thus making the model more realistic. In this work they showed a topological phase transition which ranges from scale-free networks to exponential networks through three control parameters (addition of links, redirection of edges, and addition of new sites). In the present paper we show that, for arbitrary dimensionality, γ can be controlled in a simpler manner, namely by metric changes through only one control parameter (namely the ratio α_A/d) in the structure of the network. Notice, however, that the BA generalized model is not a particular case of our model, and neither the other way around.

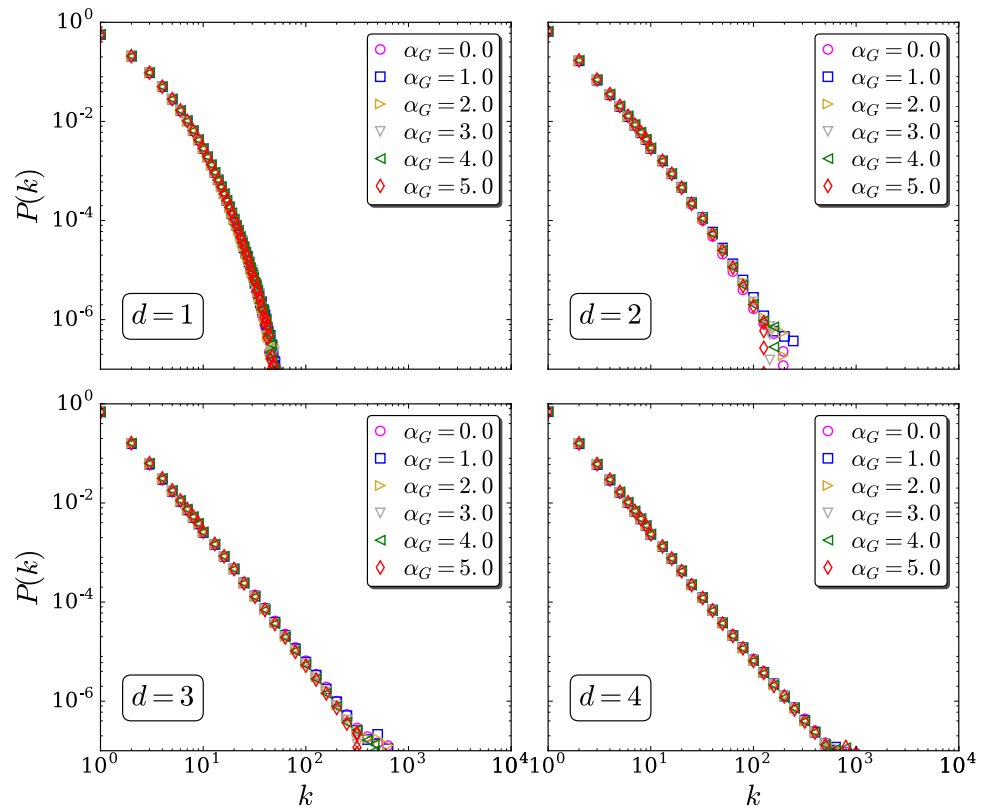


Figure 2. Connectivity distribution for $d = 1, 2, 3, 4$, $\alpha_A = 2.0$ and typical values for α_G . The simulations have been run for 10^3 samples of $N = 10^5$ sites each. We verify that $P(k)$ is independent from α_G ($\forall d$). Logarithmic binning was used whenever convenient.

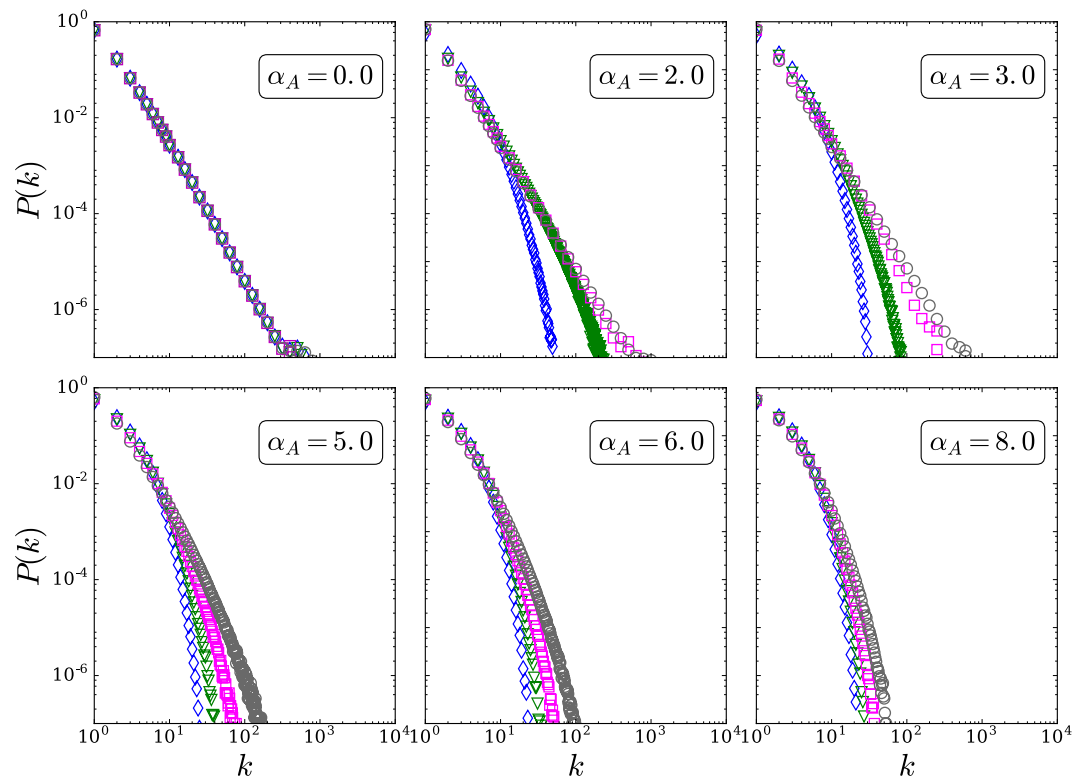


Figure 3. Degree distribution for $d = 1$ (blue diamonds), 2 (green triangles), 3 (magenta squares), 4 (grey circles), and typical values of α_A , with $\alpha_G = 2.0$. The simulations have been run for 10^3 samples of $N = 10^5$ sites each. Logarithmic binning was used whenever convenient.

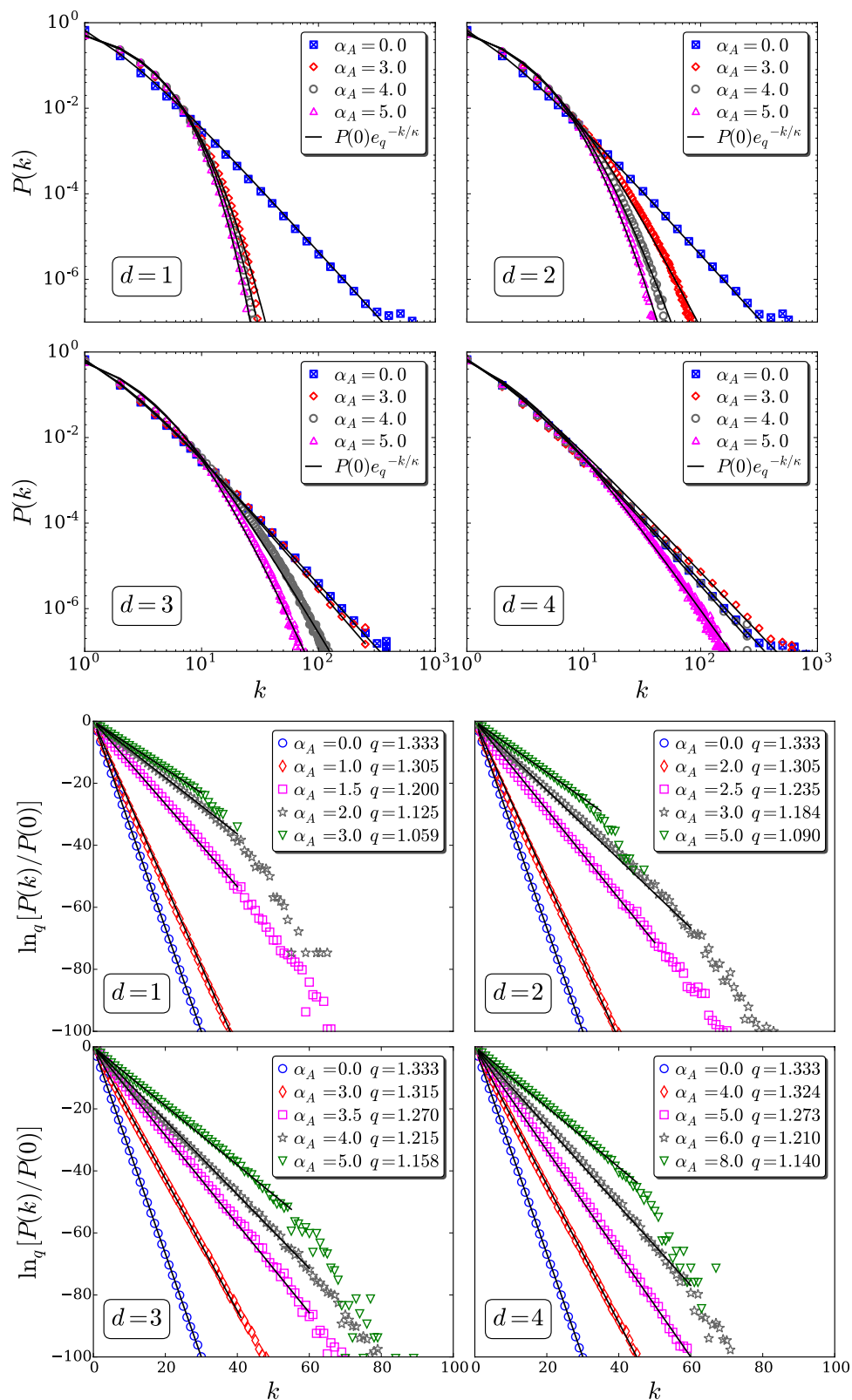


Figure 4. Fittings of the $d=1, 2, 3, 4$ connectivity distributions with the function $P(k) = P(0)e_q^{-k/\kappa}$, where $e_q^z \equiv [1 + (1 - q)z]^{1/(1-q)}$. The data are those of Fig. 3. *Top:* log-log representation. *Bottom:* $\ln_q[P(k)/P(0)]$ versus k representation. Notice that straight lines in a \ln_q -linear representation univocally determine the q -exponential function. The fitting parameters are exhibited in Fig. 5. The numerical failure, at large enough values of k , with regard to straight lines are finite-size effects that gradually disappear when we approach the thermodynamic limit $N \rightarrow \infty$. Logarithmic binning was used whenever convenient.

Dimension	α_A	q	κ	p
$d=1$	0.0	1.333 ± 0.01	0.30 ± 0.10	0.98
	1.0	1.305 ± 0.01	0.38 ± 0.10	1.00
	1.5	1.200 ± 0.02	0.75 ± 0.13	1.00
	2.0	1.125 ± 0.02	1.10 ± 0.11	0.84
	3.0	1.059 ± 0.01	1.30 ± 0.10	0.99
$d=2$	0.0	1.333 ± 0.01	0.30 ± 0.10	0.99
	2.0	1.305 ± 0.01	0.36 ± 0.14	0.94
	2.5	1.235 ± 0.02	0.66 ± 0.12	0.96
	3.0	1.184 ± 0.02	0.85 ± 0.06	0.96
	5.0	1.090 ± 0.01	1.21 ± 0.11	0.93
$d=3$	0.0	1.333 ± 0.01	0.30 ± 0.10	0.98
	3.0	1.315 ± 0.01	0.33 ± 0.12	0.94
	3.5	1.270 ± 0.01	0.49 ± 0.12	0.99
	4.0	1.215 ± 0.02	0.75 ± 0.14	0.99
	5.0	1.158 ± 0.01	0.95 ± 0.12	0.88
$d=4$	0.0	1.333 ± 0.01	0.30 ± 0.10	0.96
	4.0	1.324 ± 0.01	0.31 ± 0.10	0.99
	5.0	1.273 ± 0.01	0.55 ± 0.12	0.73
	6.0	1.210 ± 0.02	0.78 ± 0.10	0.79
	8.0	1.140 ± 0.01	1.01 ± 0.11	1.00

Table 1. Some q and κ parameters from the Fig. 4 for the best fit to a q -exponential function for $d = 1, 2, 3, 4$ connectivity distributions. The first two columns are the parameters of the network. The next two columns are the fit parameters to the q -exponential. The last column corresponds to the p -values for nonparametric statistical Kolmogorov-Smirnov test. In this test fitting curves are considered acceptable if $p > 0.05$. We can see in our results $p \geq 0.73$ indicating that connectivity distributions are very well described for q -exponential functions. We analyzed networks of the 10^5 sites and 10^3 samples. The errors that are indicated were calculated through upper bounds of those obtained by using the chi-square method.

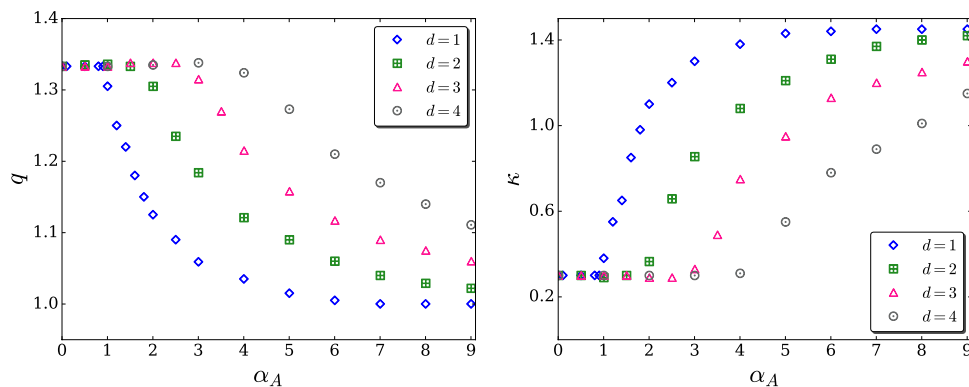


Figure 5. q and κ for $d = 1, 2, 3, 4$. For $\alpha_A = 0$ and $\forall d$, we recover the Barabási-Albert universality class $q = 4/3$ (hence $\gamma = 3$)³⁶, which has no metrics.

Large-scale simulations have been performed for the ($d = 1, 2, 3, 4$) models for fixed (α_G, α_A) , and we have verified in all cases that the degree distribution $P(k)$ is completely independent from α_G ; see Fig. 2. Using this fact, we have arbitrarily fixed $\alpha_G = 2$, and have numerically studied the influence of (d, α_A) on $P(k)$; see Figs 3 and 4. In all cases, the q -exponential fittings $P(k) = P(0)e_q^{-k/\kappa}$ with $q > 1$ and $\kappa > 0$ have been remarkably good. To test the goodness of fit, we performed Kolmogorov-Smirnov test³⁸ (see Table 1). To deal with the problem that the data are very sparse in the tail, we excluded data points with sample probability less than 10^{-6} . The best fitting values for (q, κ) are indicated in Fig. 5. From normalization of $P(k)$, $P(0)$ can be expressed as a straightforward function of (q, κ) .

Our most remarkable results are presented in Fig. 6, namely the fact that both the index q and the characteristic degree (or “effective temperature”) κ do not depend from (α_A, d) in an independent manner but *only from the ratio* α_A/d . This nontrivial fact puts the growing d -dimensional geographically located models that have been introduced here for scale-free networks, on similar footing as long-range-interacting many-body classical Hamiltonian systems such as the inertial XY planar rotators^{39–42} (possibly the generic inertial n -vector rotators

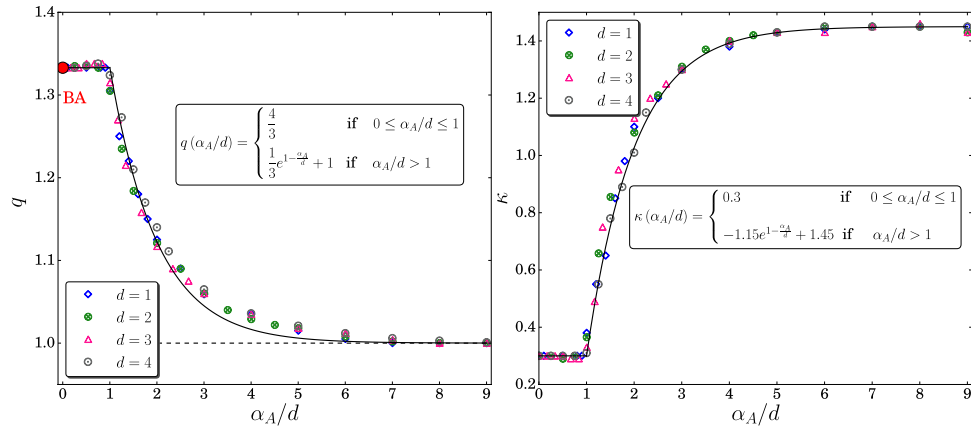


Figure 6. q and κ versus α_A/d (same data as in Fig. 5). We see that $q = 4/3$ for $0 \leq \alpha_A/d \leq 1$, and a nearly exponential behavior emerges for $\alpha_A/d > 1$ ($\forall d$); similarly for κ . These results exhibit the universality of both q and κ . The red dot indicates the Barabási-Albert (BA) universality class $q = 4/3$ ³⁶. In what concerns the universal $q = 4/3$ cut-off (i.e., the $1/(q - 1) = 3$ cut-off), see⁵⁰ and references therein.

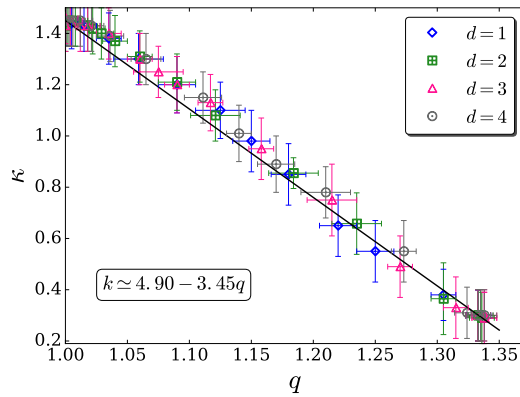


Figure 7. All the values of q and κ for the present $d = 1, 2, 3, 4$ models follow closely the linear relation Eq. (3) (continuous straight line). The upmost value of q is $4/3$, yielding $\kappa \simeq 0.3$ ($\forall d$).

as well^{43,44}) and Fermi-Pasta-Ulam oscillators, assuming that the strength of the two-body interaction decreases with distance as $1/(\text{distance})^\alpha$. Moreover, as first pointed out generically by Gibbs himself⁴⁵, we have the facts that the BG canonical partition function of these classical systems anomalously diverges with size for $0 \leq \alpha/d \leq 1$ (long-range interactions, e.g., gravitational and dipole-monopole interactions) and converges for $\alpha/d > 1$ (short-range interactions, e.g., Lennard-Jones interaction), and the internal energy per particle is, in the thermodynamical limit, constant for short-range interactions whereas it diverges like $N^{1-\alpha/d}$ for long-range interactions, N being the total number of particles.

If all these meaningful scalings are put together, we obtain a highly plausible scenario for the respective domains of validity of the Boltzmann-Gibbs (additive) entropy and associated statistical mechanics, and that of the nonadditive entropies S_q (with $q \neq 1$) and associated statistical mechanics.

Finally, we notice in Fig. 6 that both q and κ approach quickly their BG limits ($q = 1$) for $\alpha_A/d \rightarrow \infty$. Moreover, the same exponential $e^{1-\alpha/d}$ appears in both heuristic expressions for q and κ . Consequently, the following linear relation can be straightforwardly established:

$$\kappa \simeq 4.90 - 3.45q. \tag{3}$$

In fact, this simple relation is numerically quite well satisfied as can be seen in Fig. 7. Its existence reveals an interesting peculiarity of the nature of q -statistics. If in the celebrated BG factor $e^{-\text{energy}/kT}$, corresponding to $q = 1$, we are free to consider an arbitrary value for T , how come in the present problem, κ is not a free parameter but has instead a fixed value for each specific model that we are focusing on? This is precisely what occurs in the high-energy applications of q -statistics, e.g., in quark-gluon soup⁴⁶ where $q = 1.114$ and $T = 135.2 \text{ MeV}$, as well as in all the LHC/CERN and RHIC/Brookhaven experiments⁷. Another example which is reminiscent of this type of behavior is the sensitivity to the initial conditions at the edge of chaos (Feigenbaum point) of the logistic map; indeed, the inverse q -generalized Lyapunov exponent satisfies the linear relation $1/\lambda_q = 1 - q^{47,48}$. The cause of this interesting and ubiquitous feature comes from the fact that q -statistics typically emerges at critical-like regimes

and is deeply related to an hierarchical occupation of phase space (or Hilbert space or Fock space), which in turn points towards asymptotic power-laws (see also⁴⁹). In other words, κ plays a role analogous to a critical temperature, which is of course not a free parameter but is instead fixed by the specific model.

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Author Contributions

C.T. conceived the research, analyzed the results, contributed to the manuscript text and revised it. S.B. developed and carried out the numerics, prepared figures, analyzed the results, contributed to the manuscript text and revised it. L.R.d.S. offered constructive suggestions, analyzed the results, contributed to the manuscript text and revised it.

Additional Information

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