

Scale-free Networks and Nonextensive Statistics

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OUR GOALS

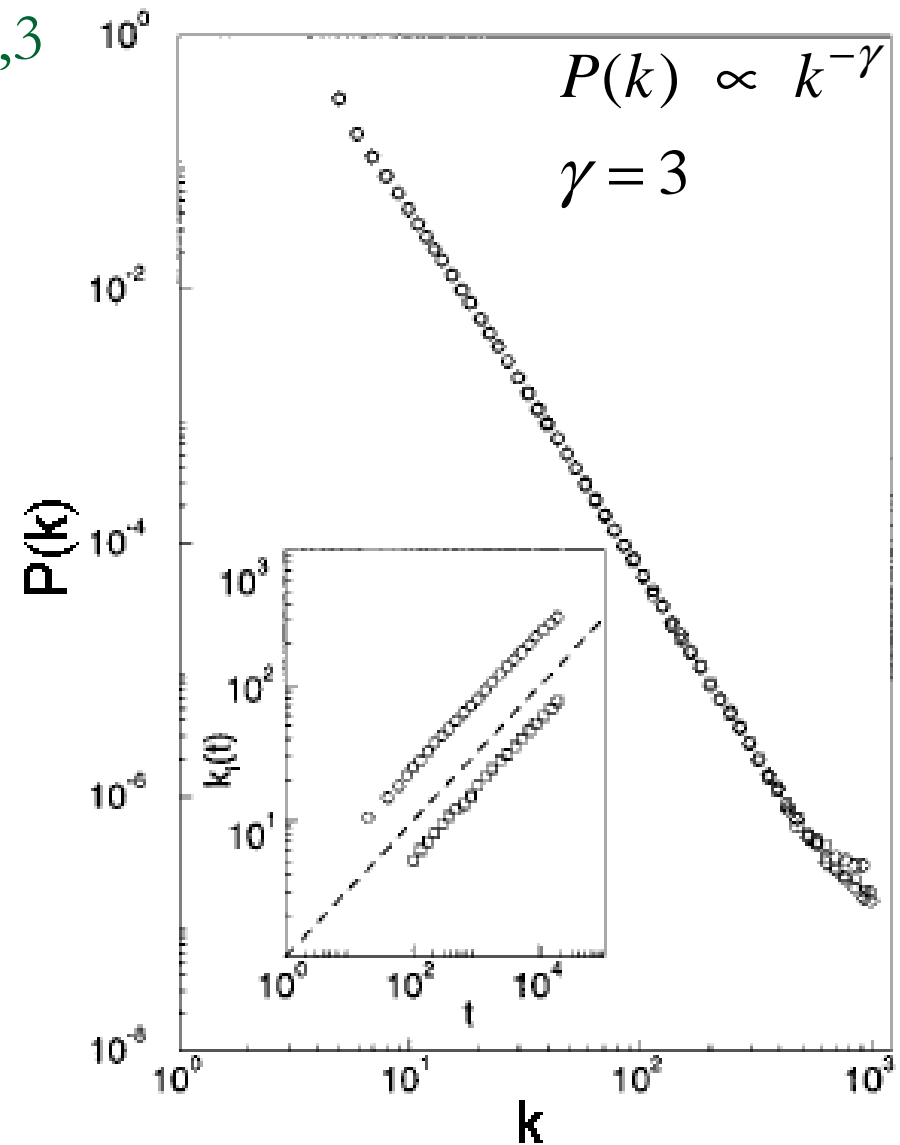
- Growth of an asymptotically scale-free network including metrics.
- Growth of a geographically localized network (around its baricenter).
- To exhibit effects of competition between metrical neighborhood, connectivity, affinity and fitness.
- To analyze the influence of considering a fitness power-law distributed.
- To analyse the influence of the affinity.
- Last but not least, to exhibit the connection between scale-free networks and **nonextensive statistics**.

Scale-free Networks^{1,2,3}

- Barabási and Albert¹;

$$\Pi(k_i) = \frac{k_i}{\sum_{j=1}^{N'} k_j} \quad (01)$$

$$\langle k_i \rangle = \left(\frac{t}{i} \right)^\beta \quad (02)$$



¹Science **286**, 509 (1999) ; Rev. Mod Phys. **74**, 47 (2002)

²M. Boguñá and R. Pastor-Satorras, Physical Review E **68**, 036112 (2003)

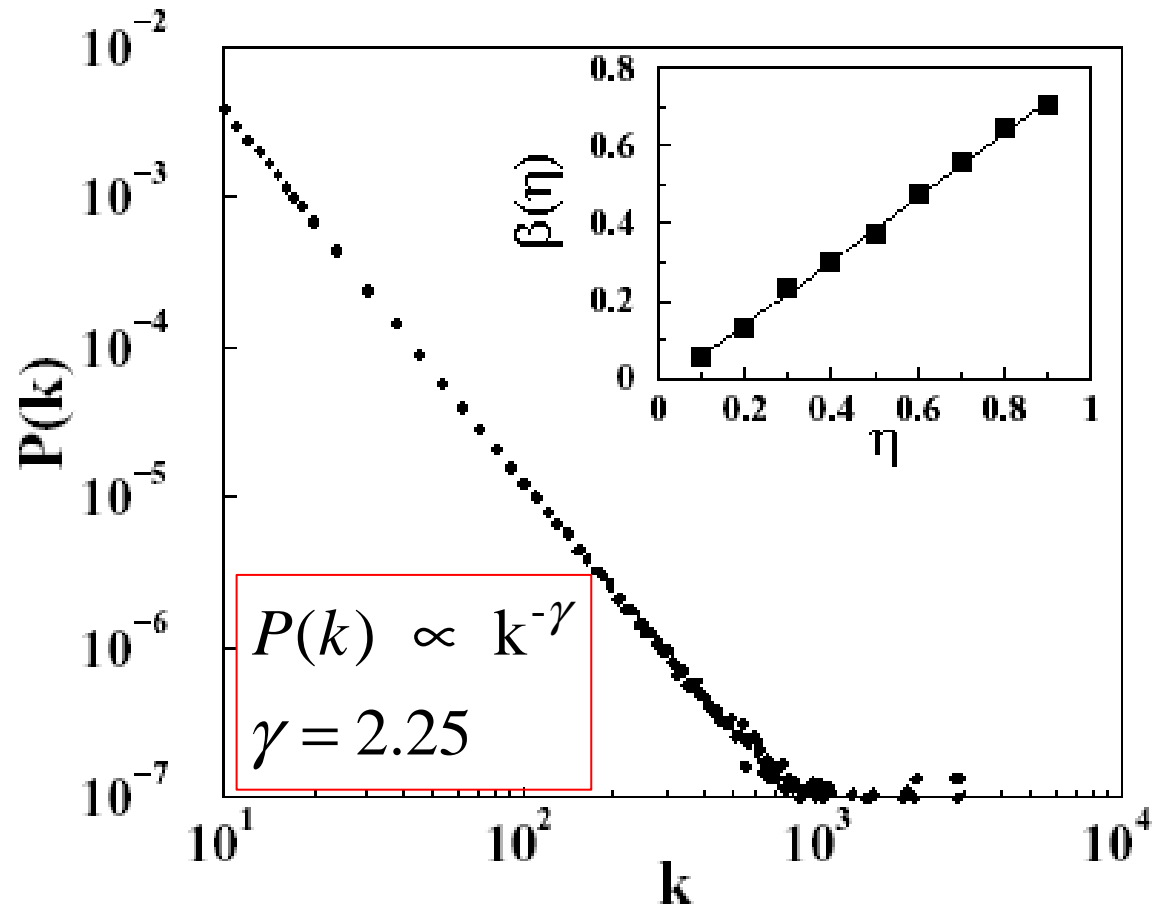
³S. Thurner and C. Tsallis, Europhys Letters **72**, 197 (2005)

Fitness Model

- Bianconi and Barabási⁴;
- Albert and Barabási⁵;

$$\Pi(k_i) = \frac{k_i \eta_i}{\sum_{j=1}^{N'} k_j \eta_j} \quad (04)$$

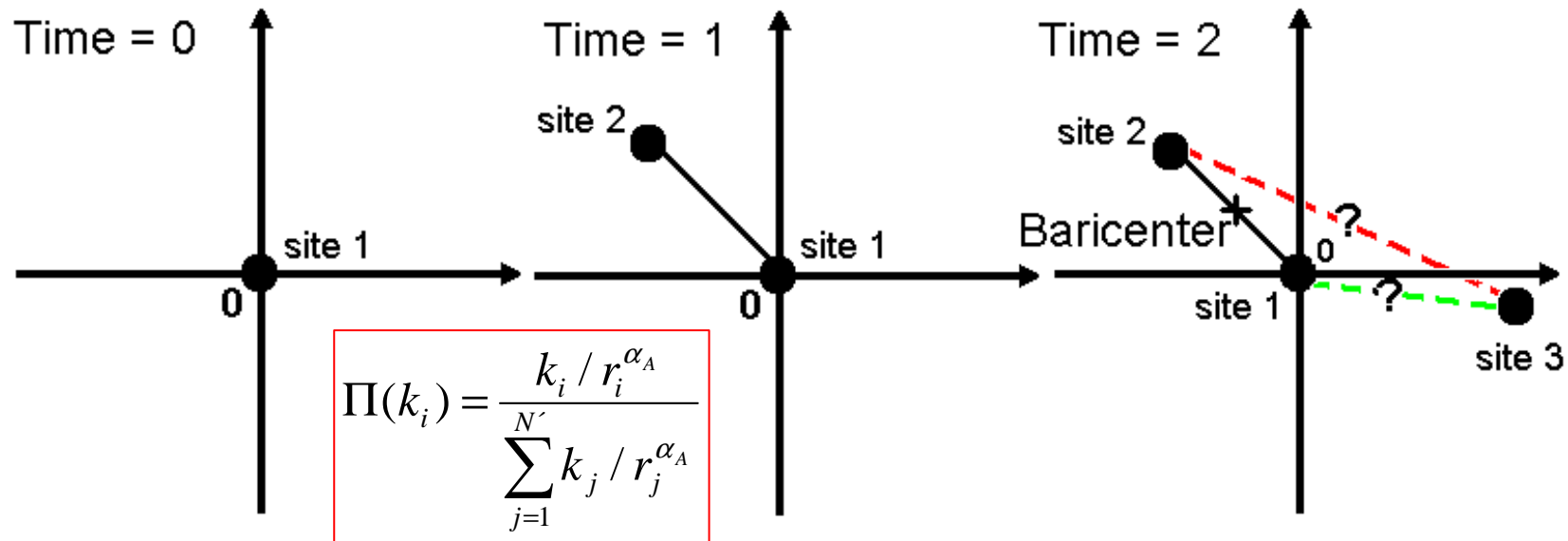
$$\langle k_i \rangle = \left(\frac{t}{i} \right)^{\beta(\eta_i)} \quad (05)$$



⁴Europhys. Lett. **54**,436 (2001) ; ⁵Rev. Mod Phys. **74**, 47 (2002)

Barabási-Albert Model with Euclidean Distance Power-law Distributed

Network Construction:



$$P(r) \propto r^{-\gamma_G}$$

$$\gamma_G > 1$$

where

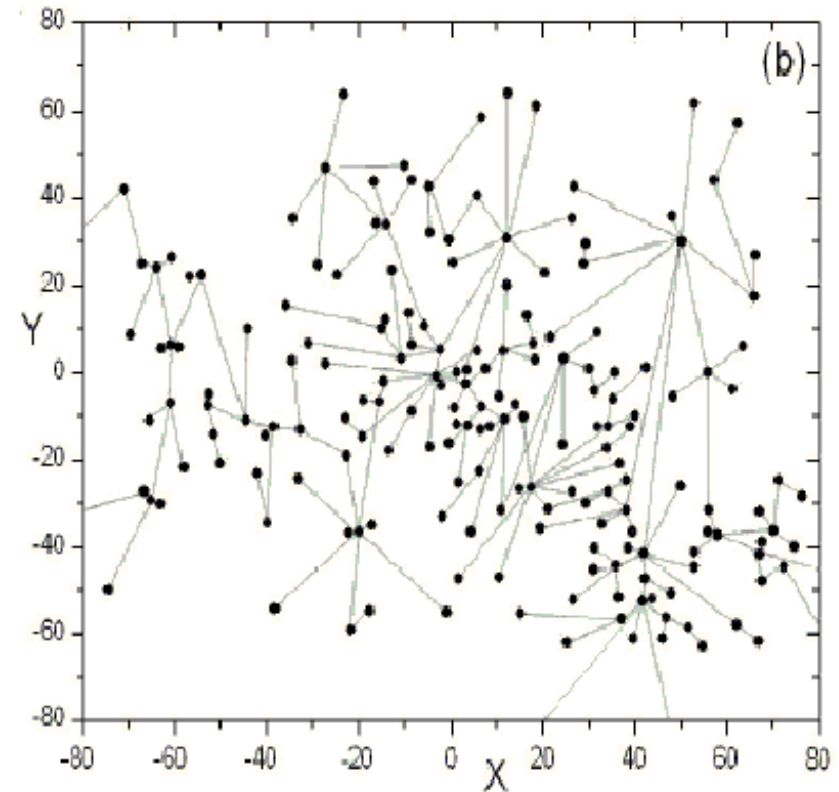
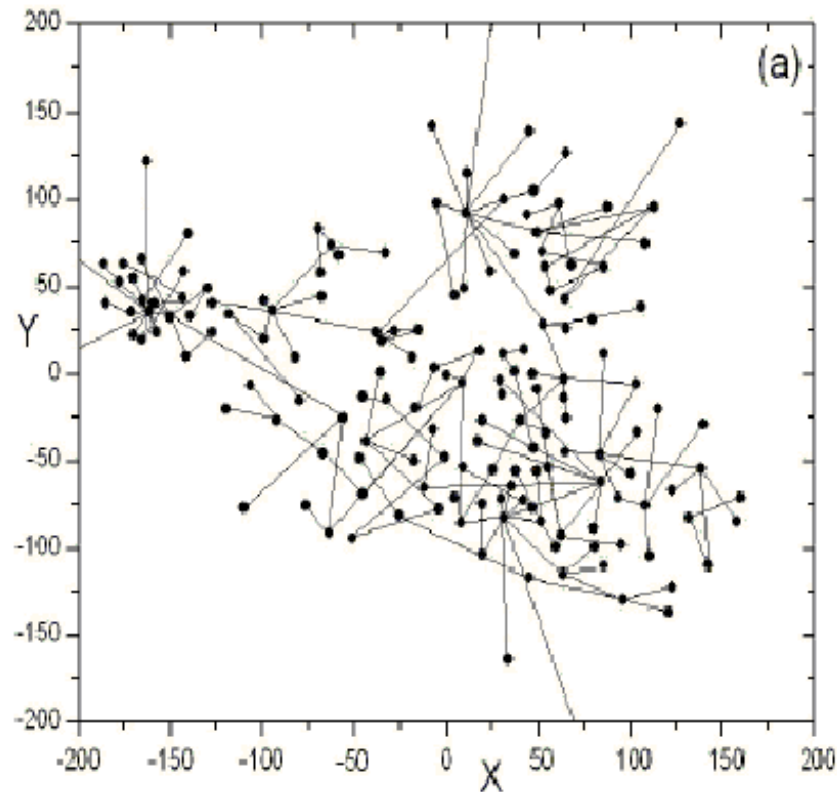
$$\gamma_G = \frac{3 + \alpha_G}{2 + \alpha_G}$$

$$\alpha_G \geq 0$$

$$r = (1 - \xi')^{-(2 + \alpha_G)}$$

$$\theta = 2\pi\xi'$$

Examples

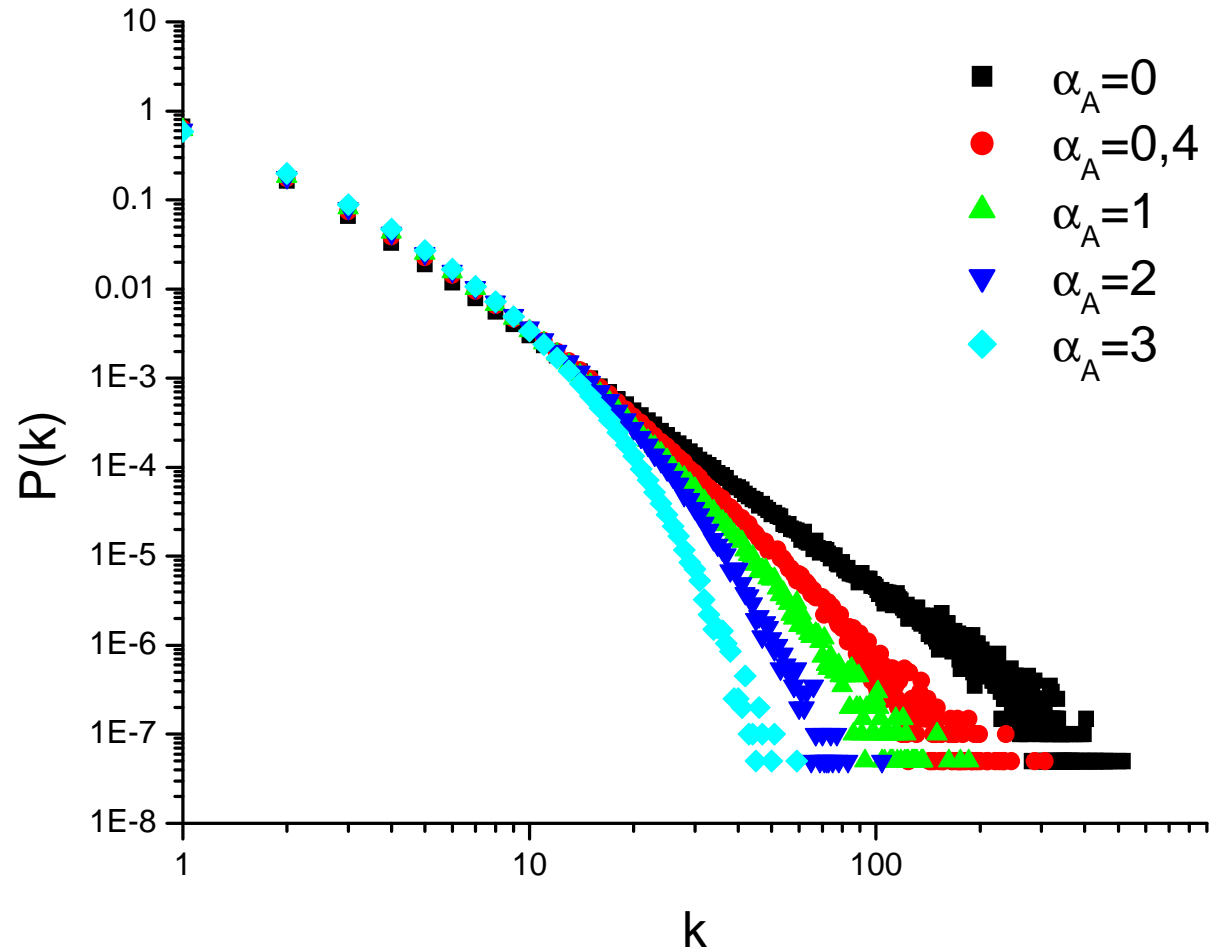


$N = 250$ nodes (a) $(\alpha_G, \alpha_A) = (1, 0)$ and (b) $(\alpha_G, \alpha_A) = (1, 4)$.

The starting site is at $(X, Y) = (0, 0)$. Notice the spontaneous emergence of hubs.

Barabási-Albert Model with Euclidean Distance Power-law Distributed

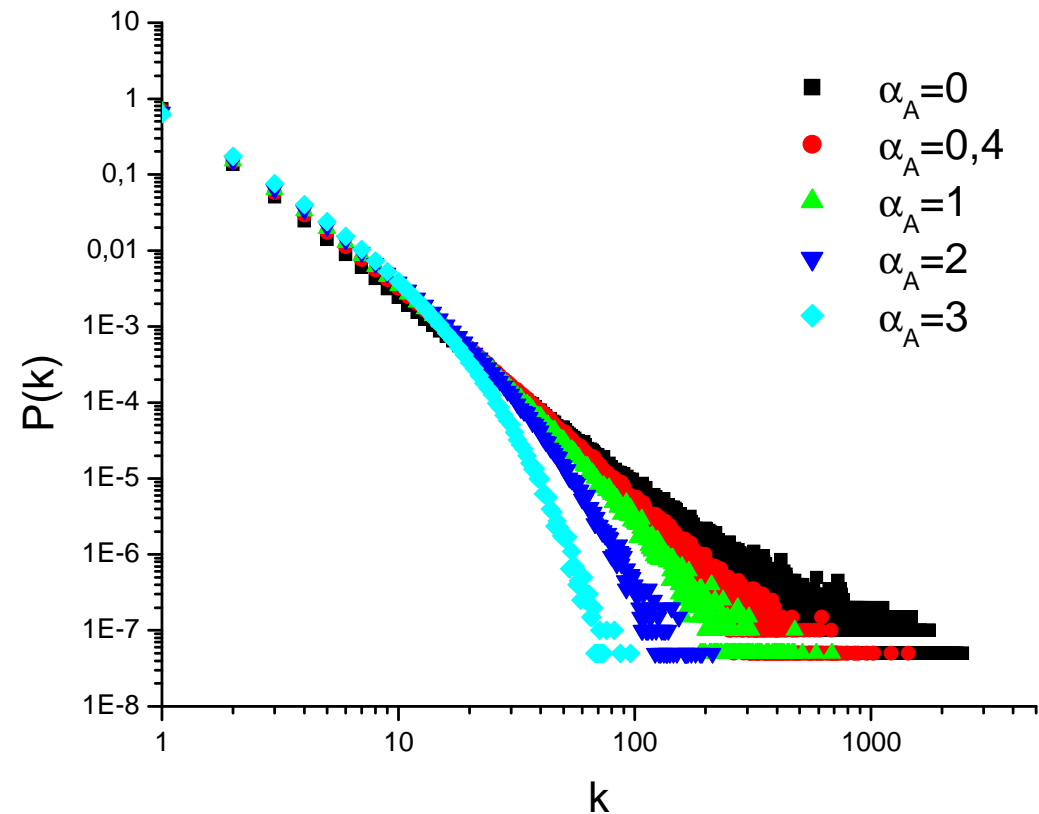
$$\Pi(k_i) = \frac{k_i / r_i^{\alpha_A}}{\sum_{j=1}^{N'} k_j / r_j^{\alpha_A}} \quad (03)$$



Fitness Model with Euclidean Distance Power-law Distributed

- Inspired in the works of Soares, Tsallis, Mariz and da Silva³, and Bianconi and Barabasi⁴.

$$\Pi(k_i) = \frac{k_i \eta_i / r_i^{\alpha_A}}{\sum_{j=1}^{N'} k_j \eta_j / r_j^{\alpha_A}} \quad (06)$$



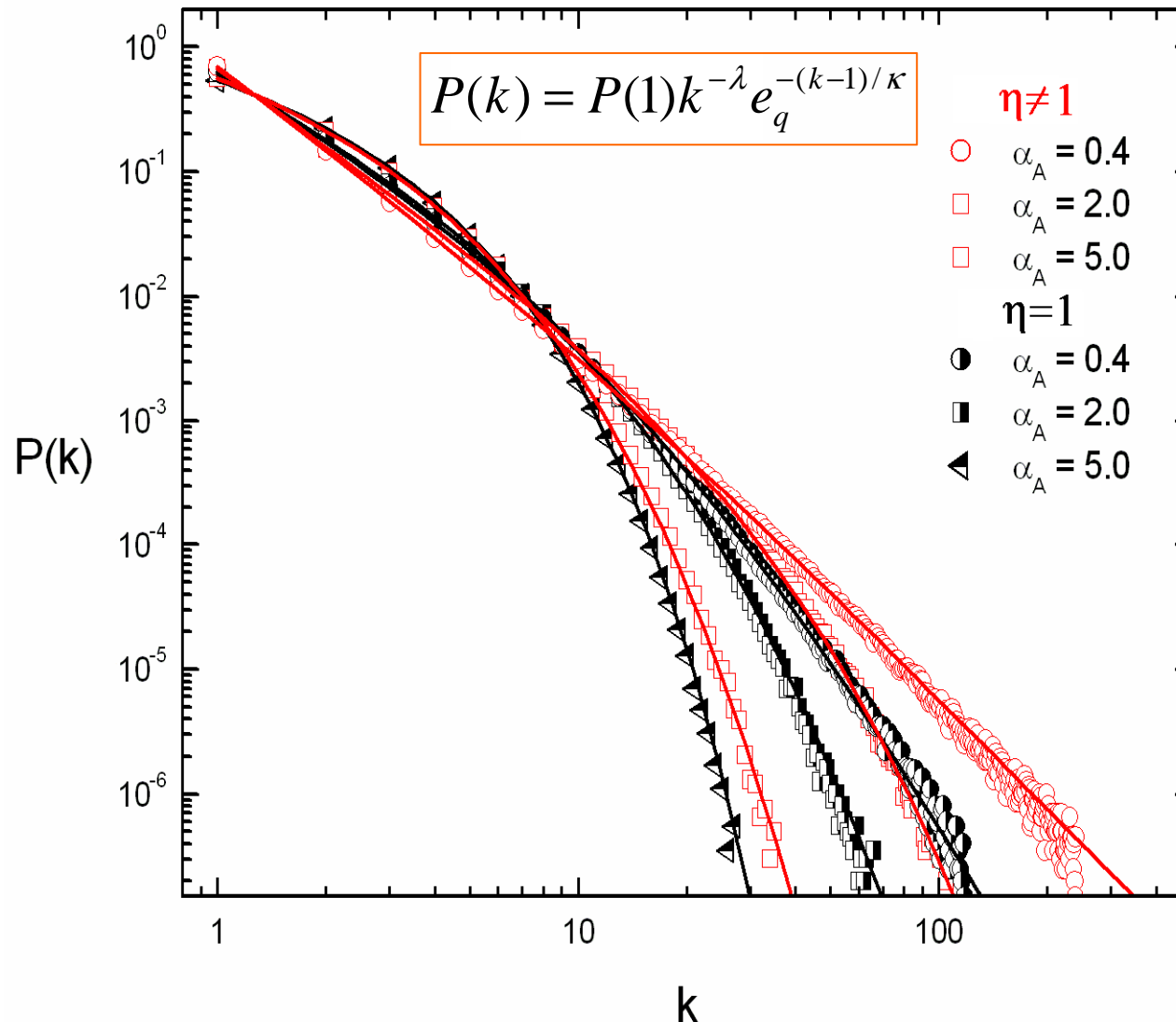
Meneses, Cunha, Soares, and da Silva,
Progress of Theoretical Physics Supplement **162**, 131 (2006)

Tsallis Nonextensive Statistical Mechanics

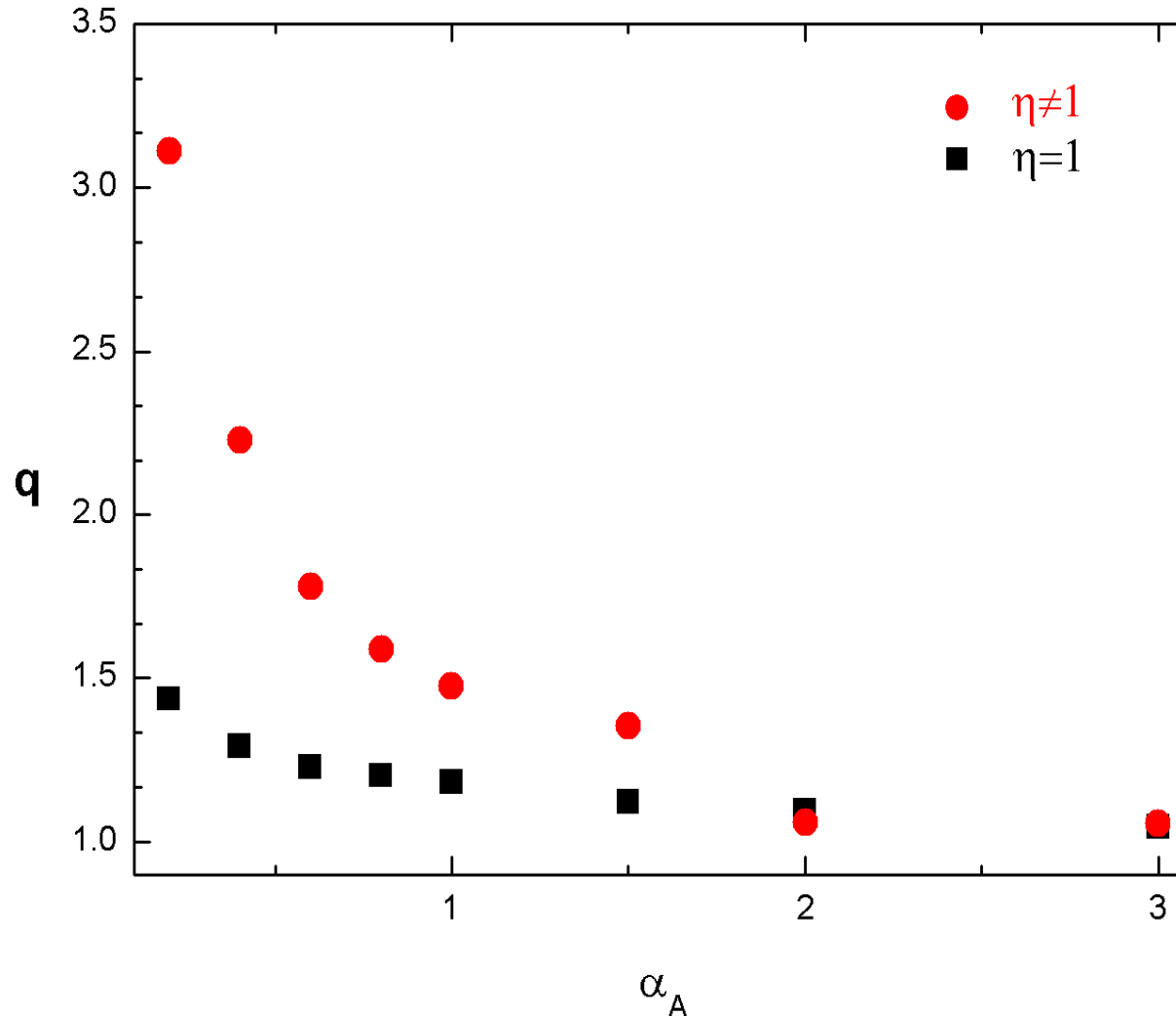
$$s_q = \frac{1 - \int dk [P(k)]^q}{q-1} \quad (q \in \mathfrak{R}; S_1 = S_{BG})$$

$$P(k) = P(1) k^{-\lambda} e_q^{-(k-1)/\kappa} \quad (07)$$

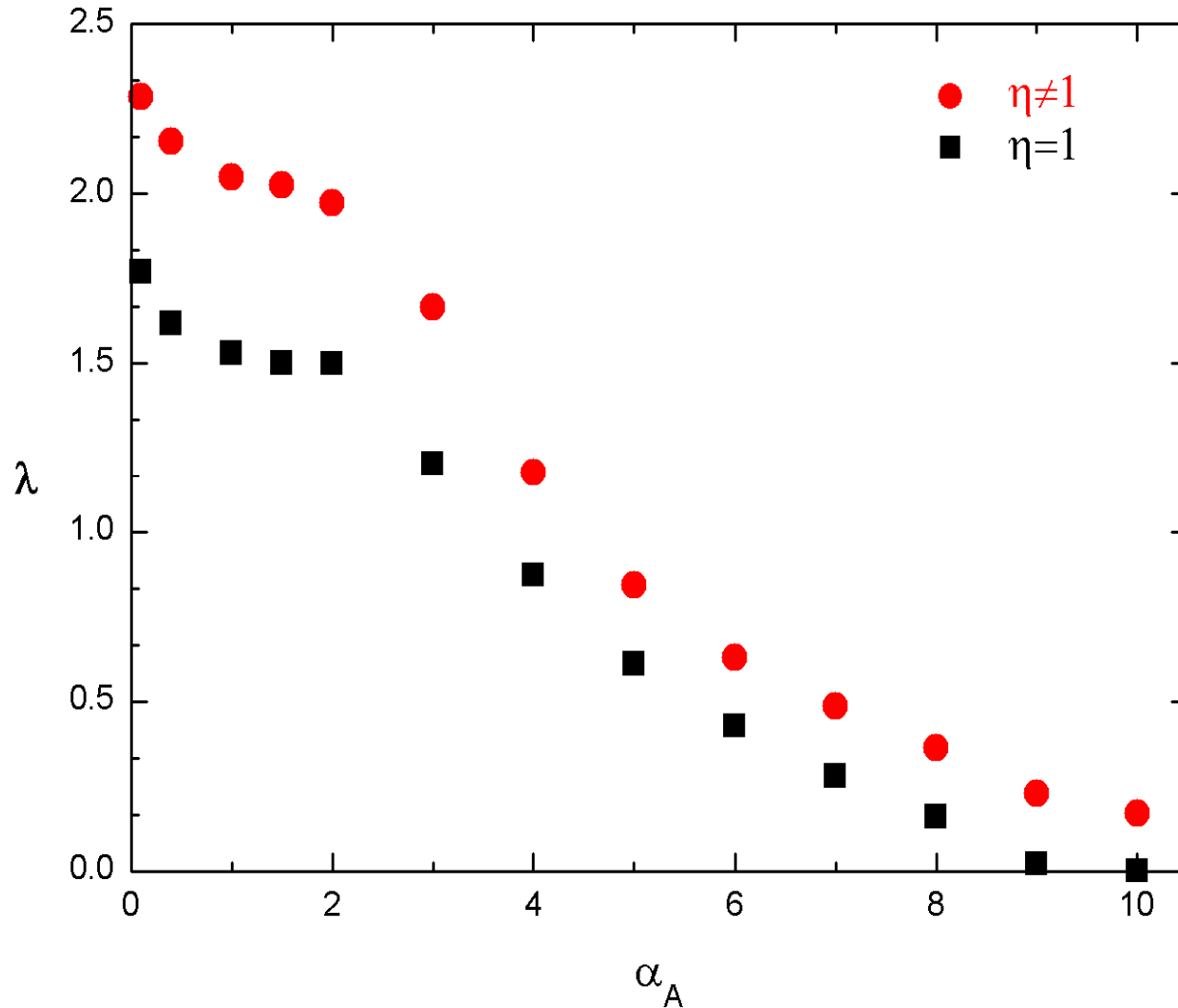
$$e_q^x \equiv [1 + (1-q)x]^{1/(1-q)}$$



Connectivity distribution $P(k)$ for typical values α_A for $\eta \neq 1$ and $\eta = 1$ models. The symbols are numerical results and continuous lines are the best fits according to $P(k)$.

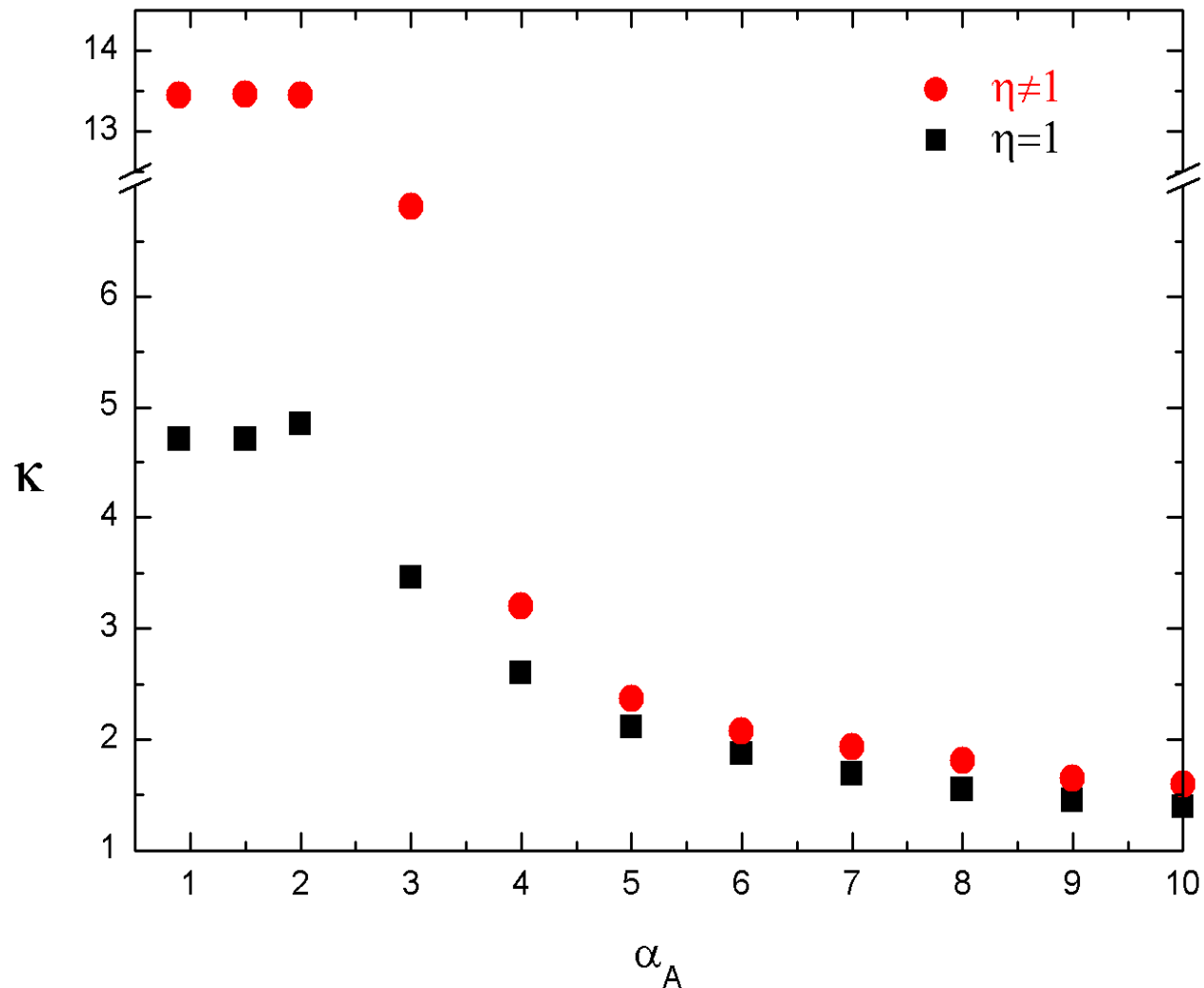


α_A -dependence of q for both $\eta \neq 1$ and $\eta = 1$ models. In this graph we observe some kinds of changes of regimes at $\alpha_A = 2$ (which coincides with the space dimensionality).

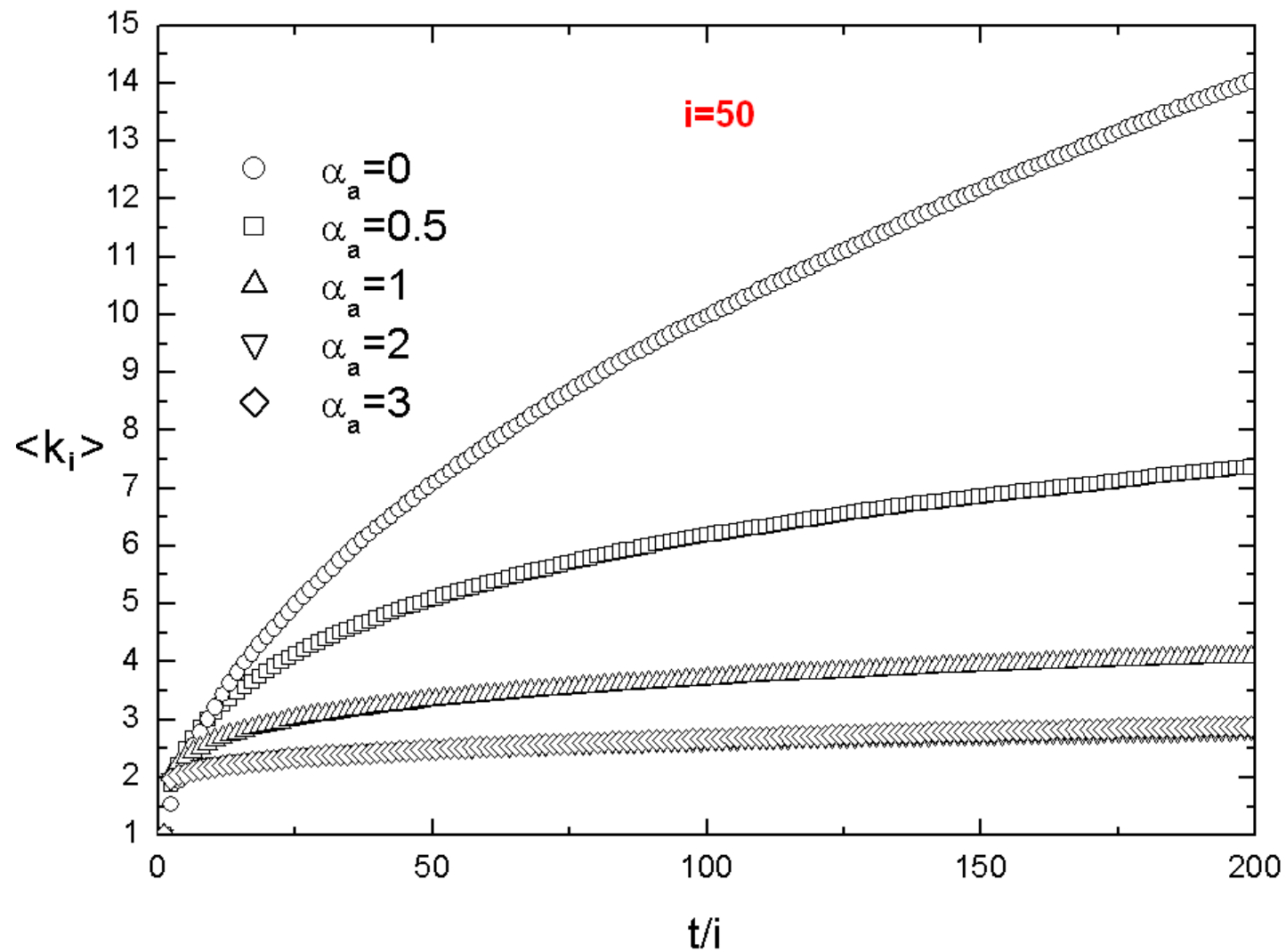


α_A -dependence of λ for both $\eta \neq 1$ and $\eta = 1$ models. In this graph we observe some kinds of changes of regimes at $\alpha_A = 2$ (which coincides with the space dimensionality).

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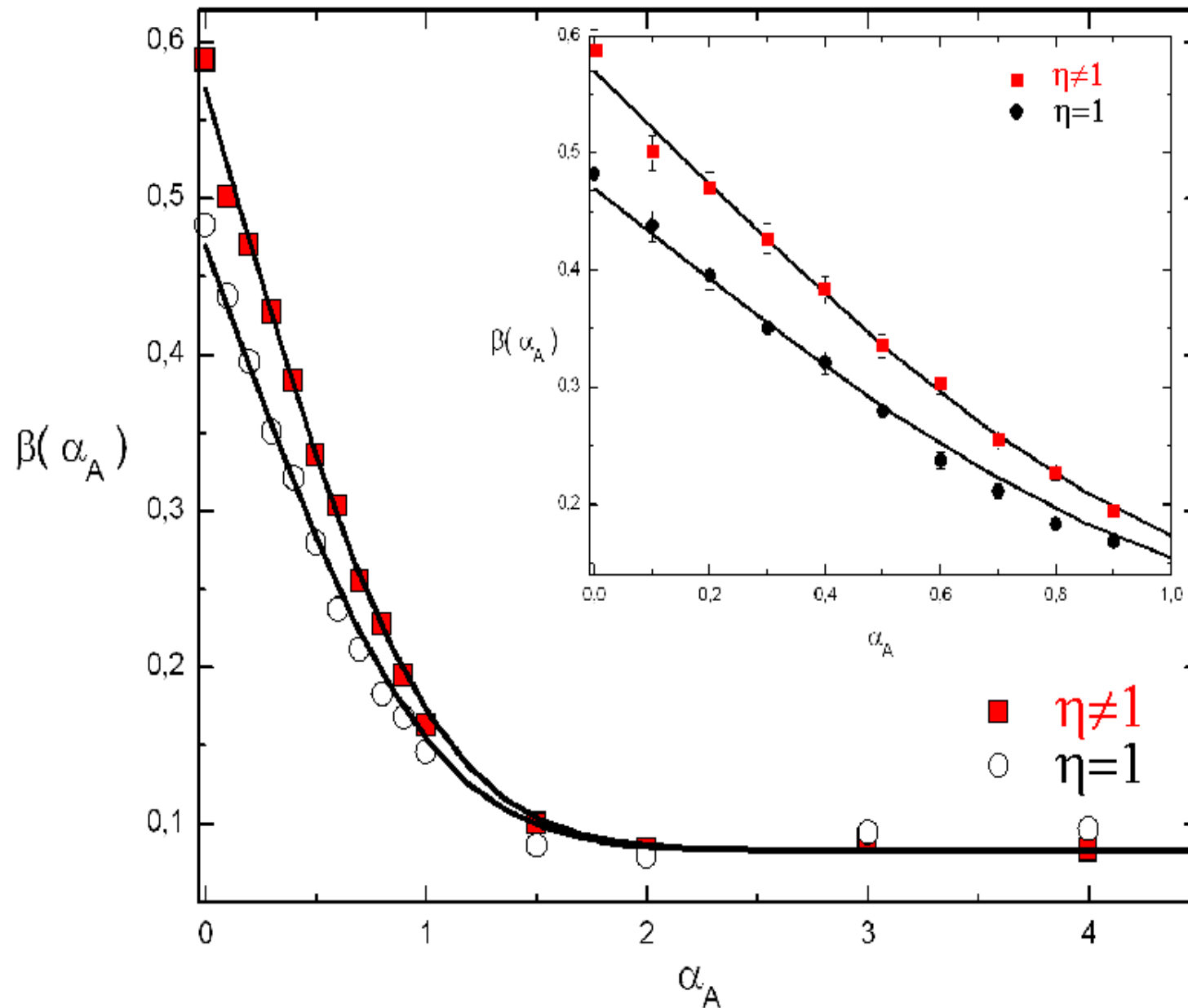


α_A -dependence of q for both $\eta \neq 1$ and $\eta = 1$ models. In this graph we observe some kinds of changes of regimes at $\alpha_A = 2$ (which coincides with the space dimensionality).

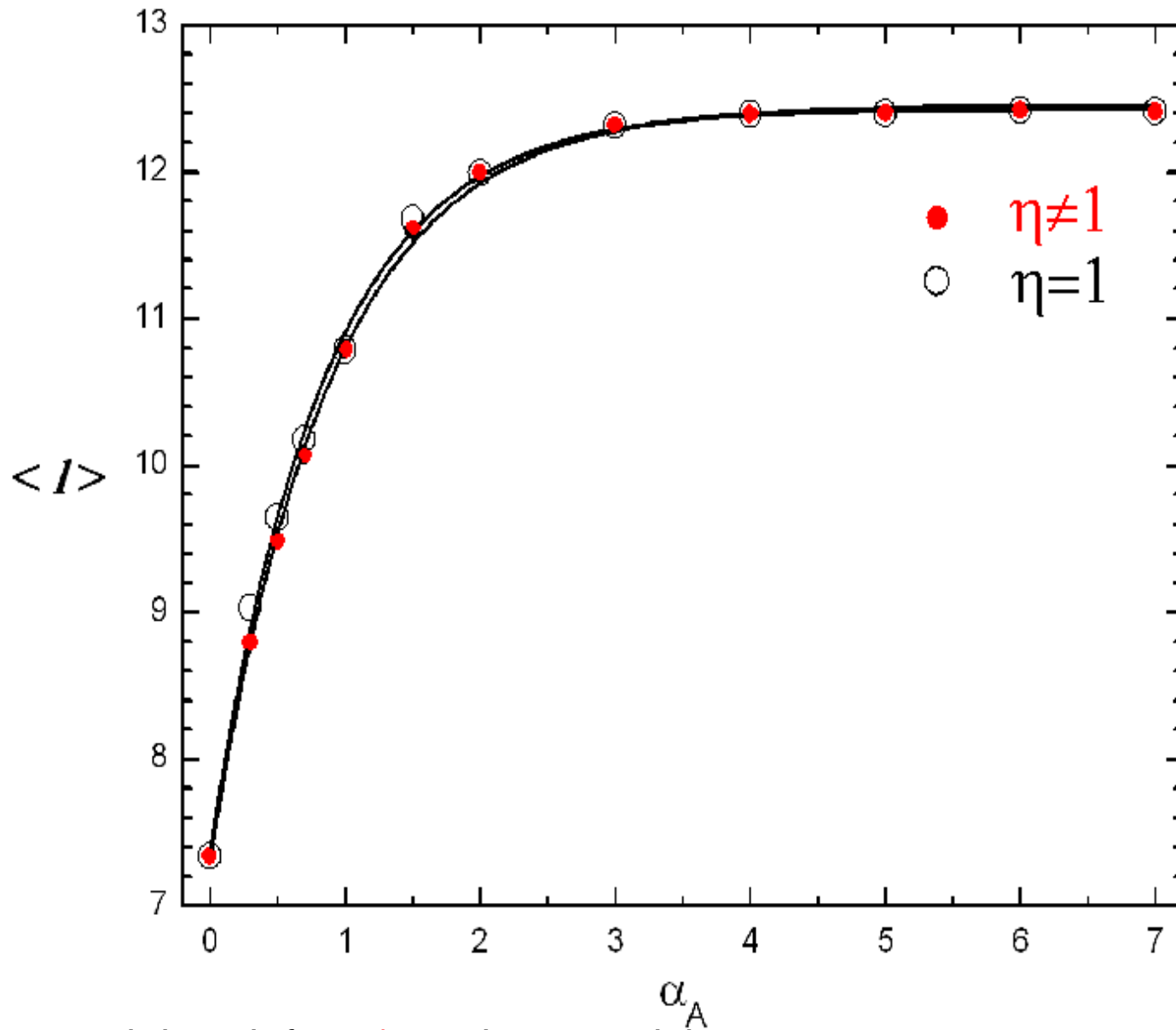


Temporal dependence of the average connectivity for $\eta \neq 1$, in 2000 samples.

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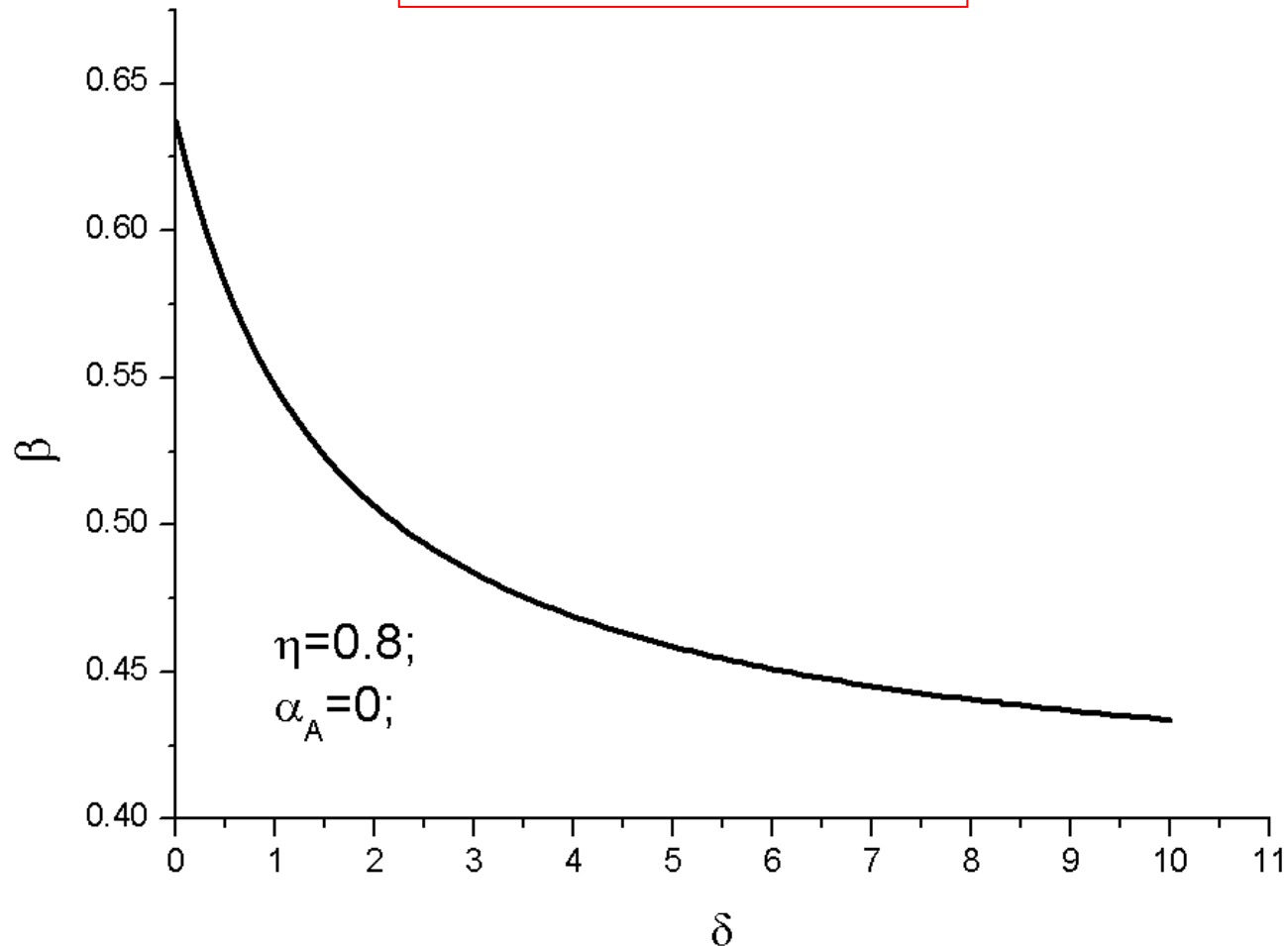


Average connectivity exponent for α_A values relative to measures on node $i = 50$.



Average path length for $\eta \neq 1$ and $\eta = 1$ models.

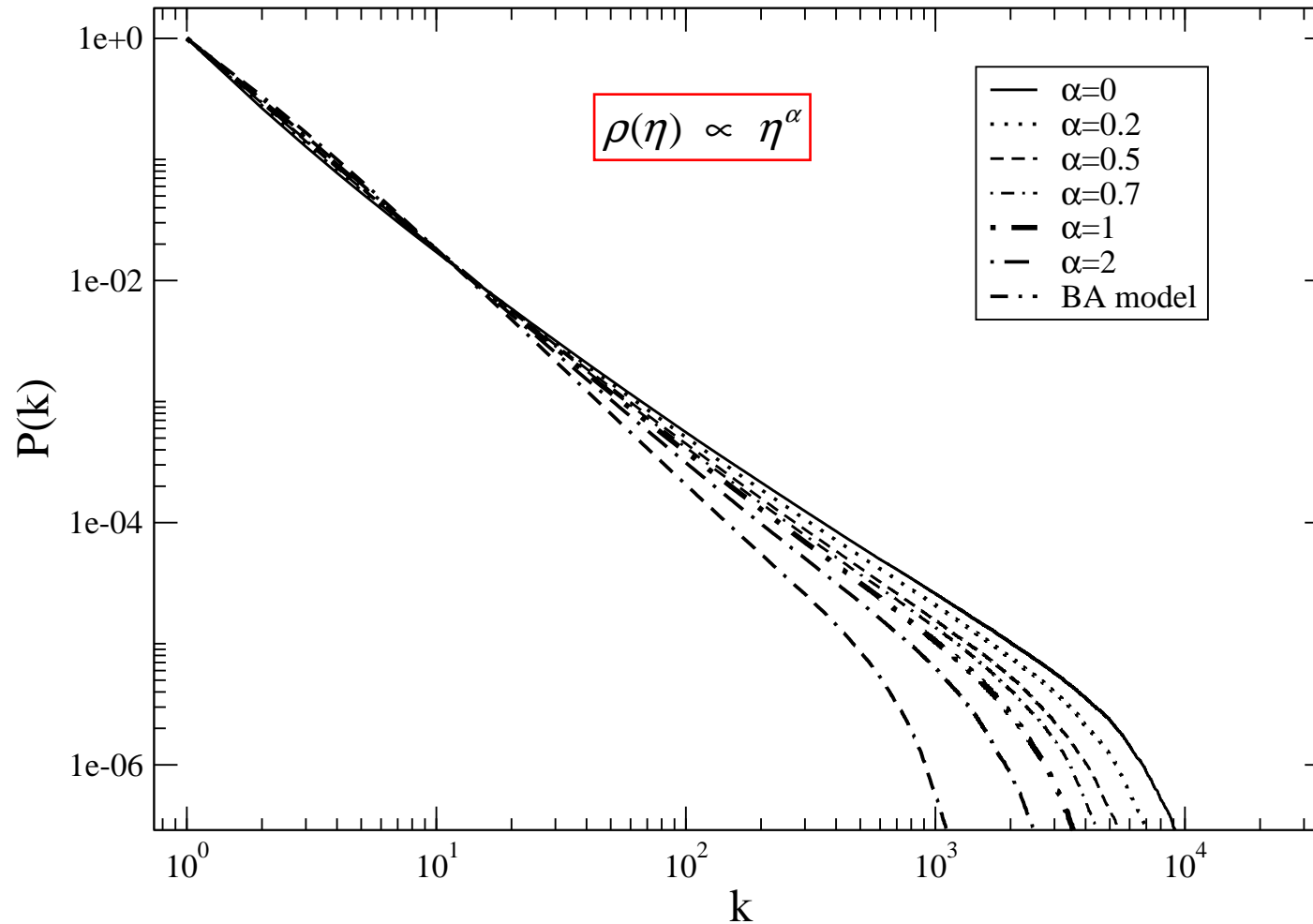
$\alpha_A \neq 0$ in progress



$$\rho(\eta) \propto \eta^\delta$$

Average connectivity exponent for δ values relative to measures on node $i = 20$.

Model with Fitness Power-law Distributed



$$2.25 < \gamma < 3$$

Generalized Model: Fitness and Euclidean Distance Power-law Distributed

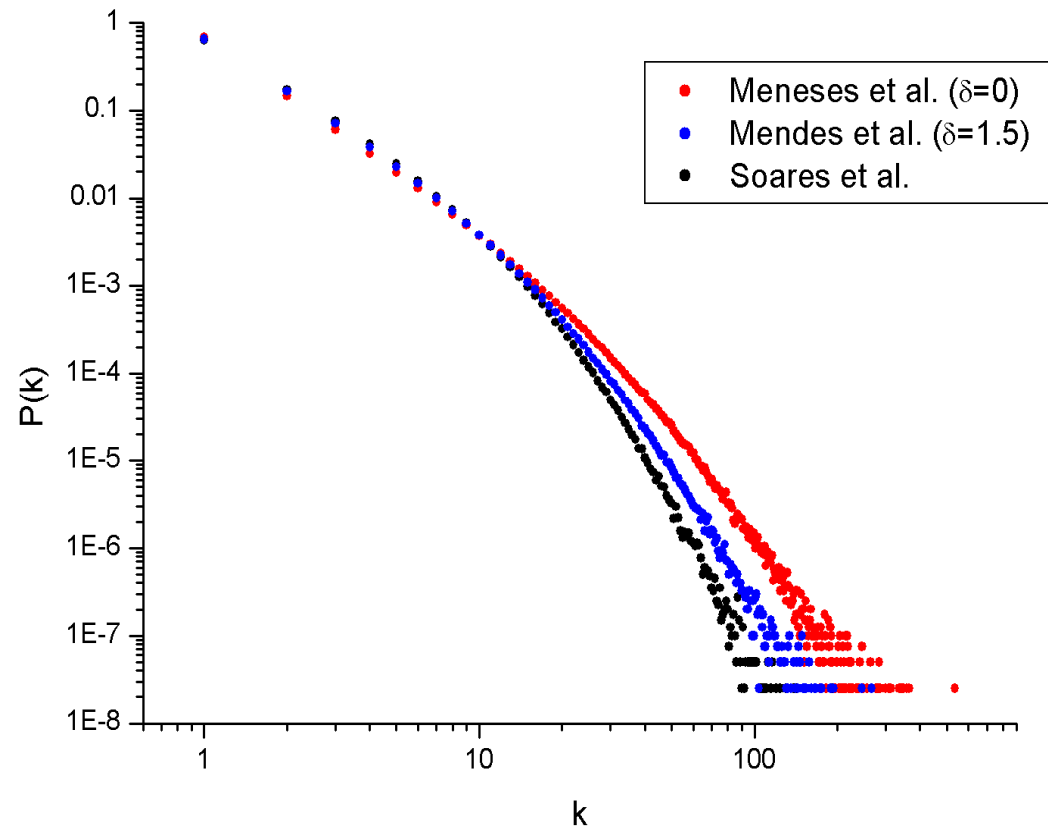
- Inspired in the works of Meneses et al;
- Mendes et al;

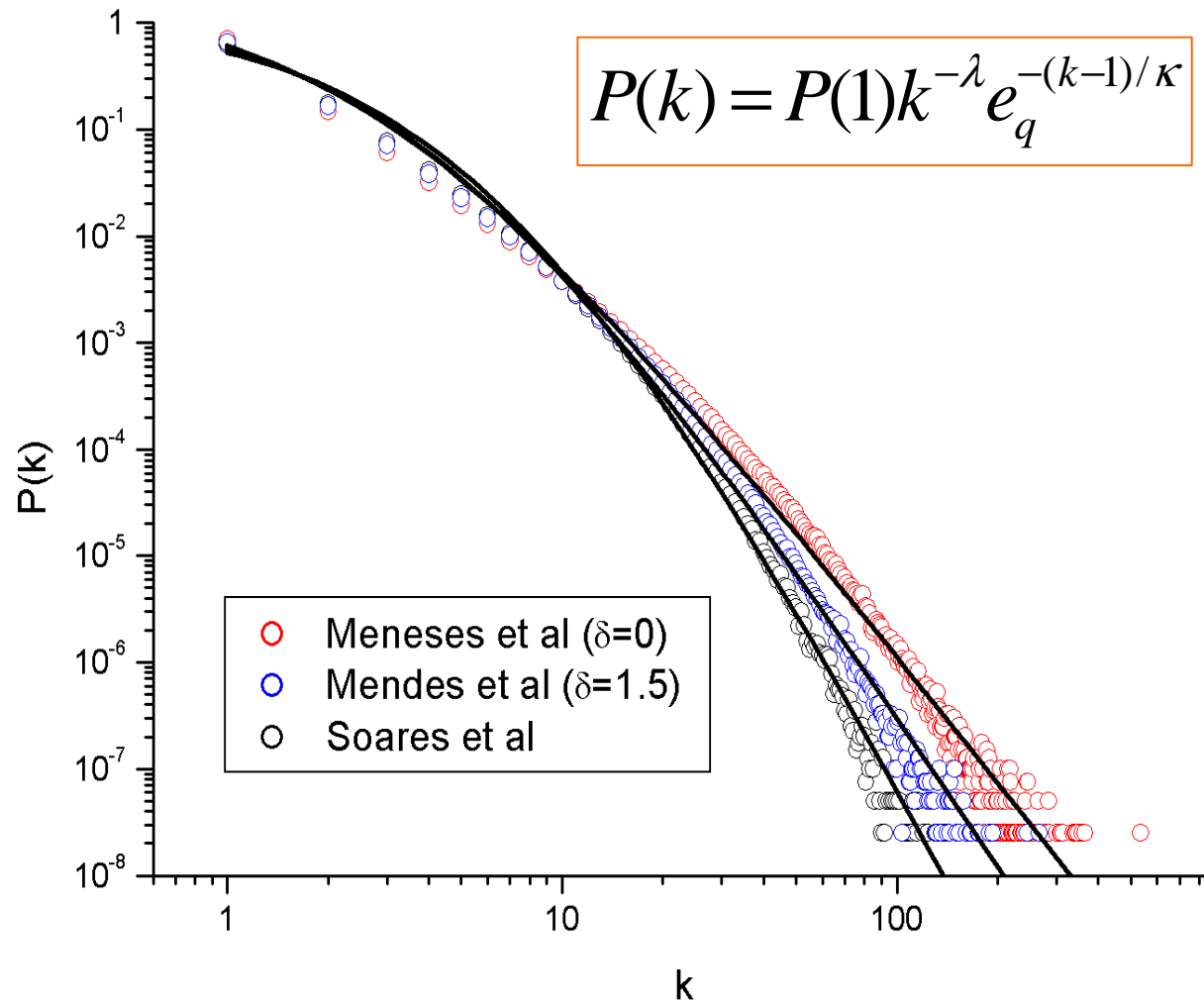
$$\Pi(k_i) = \frac{k_i \eta_i / r_i^{\alpha_A}}{\sum_{j=1}^{N'} k_j \eta_j / r_j^{\alpha_A}} \quad (08)$$

with

$$\rho(\eta) \propto \eta^\delta$$

$$\delta \geq 0$$

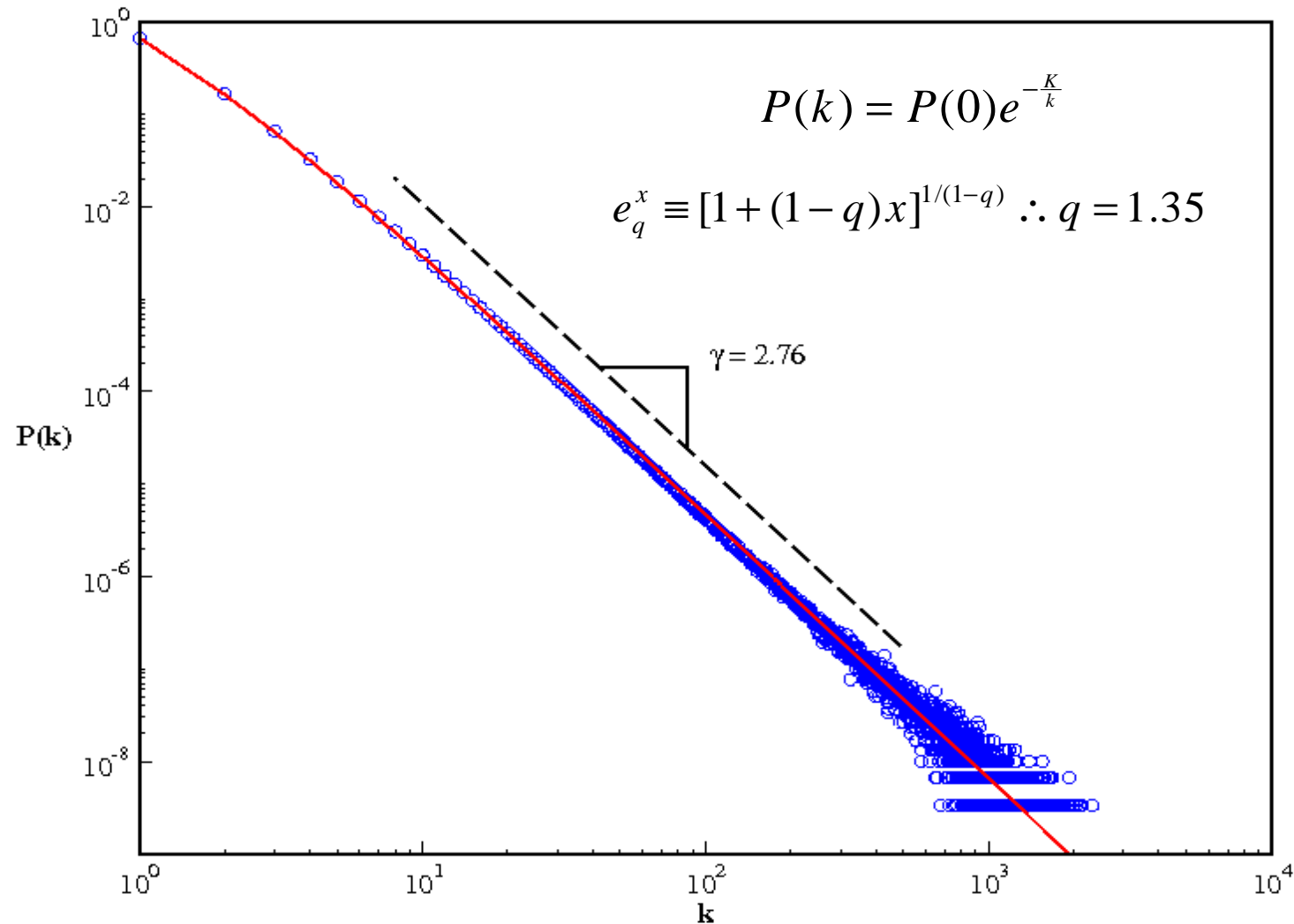




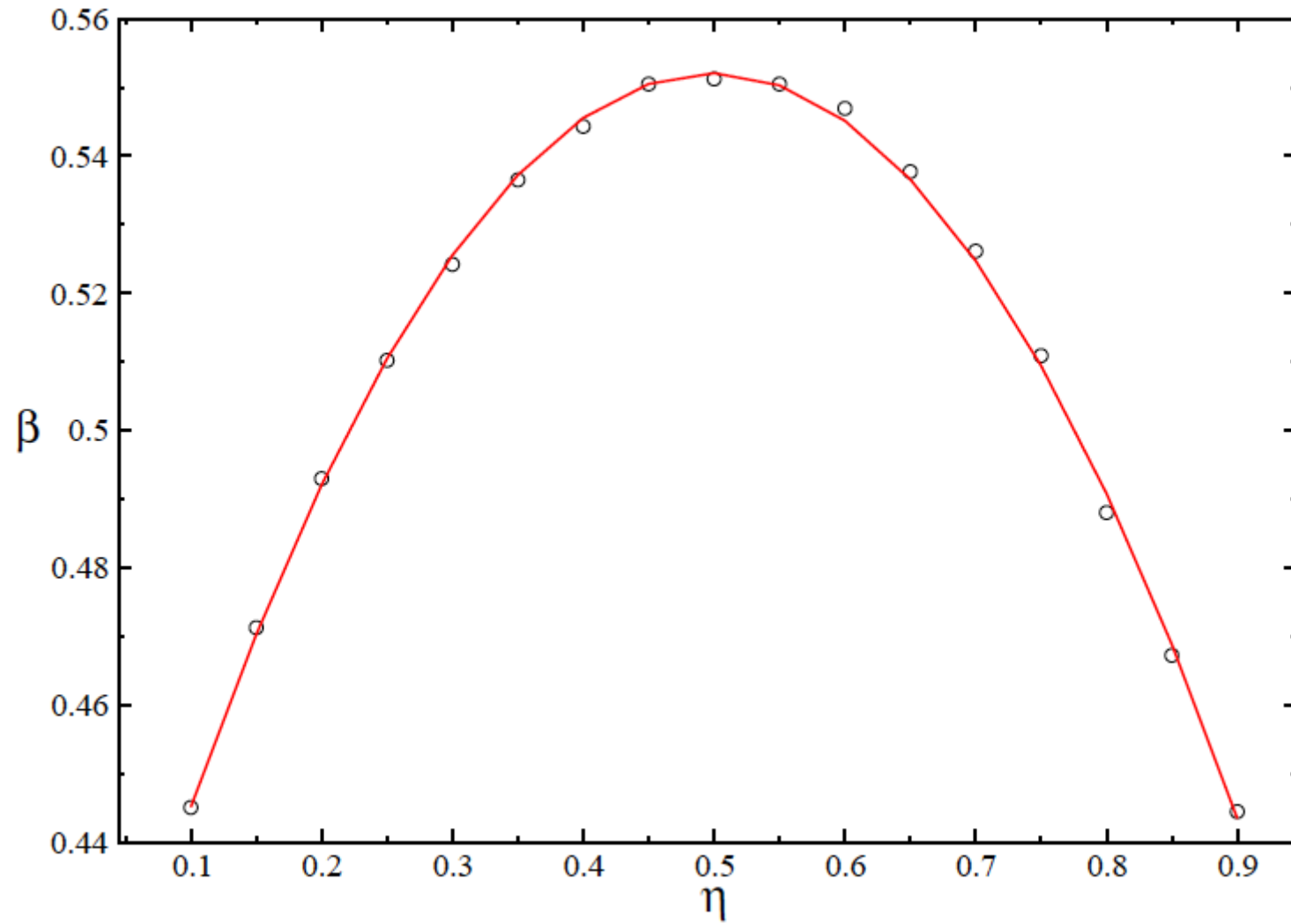
Conectivity distribution $P(k)$ for $\alpha_A=2$ for **Meneses et al**, **Mendes et al** and Soares et al models. The symbols are numerical results and continuous lines are the best fits in according to $P(k)$.

Acquaintance Model

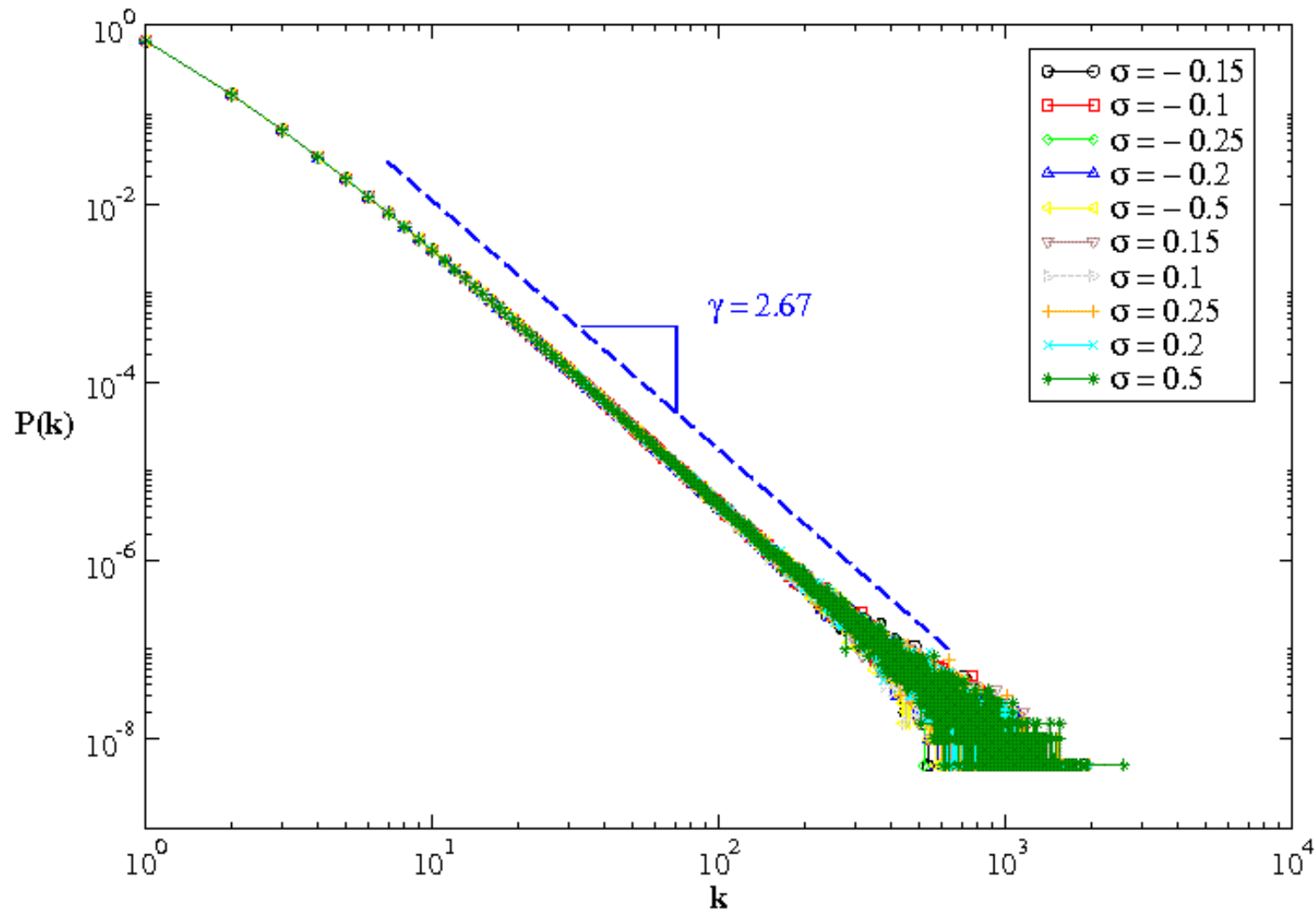
$$\Pi_{i \rightarrow j} \propto k_j \left(1 - |\eta_i - \eta_j|\right)$$



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General Acquaintance Model $\Pi_{i \rightarrow j} \propto k_j \left(1 - |\eta_i - \eta_j|^\sigma\right)$



Summary

(a)

- The present models

MODEL	CONNECTIVITY	FITNESS	METRIC	AFFINITY
Barabási-Albert	YES	NO	NO	?
Bianconi et al	YES	UNIFORM	NO	?
Soares et al	YES	NO	POWER-LAW	?
Meneses et al	YES	UNIFORM	POWER-LAW	?
Mendes et al	YES	POWER-LAW	POWER-LAW	?
Almeida et al	YES	?	?	YES
Santos et al	YES	?	?	YES

Summary

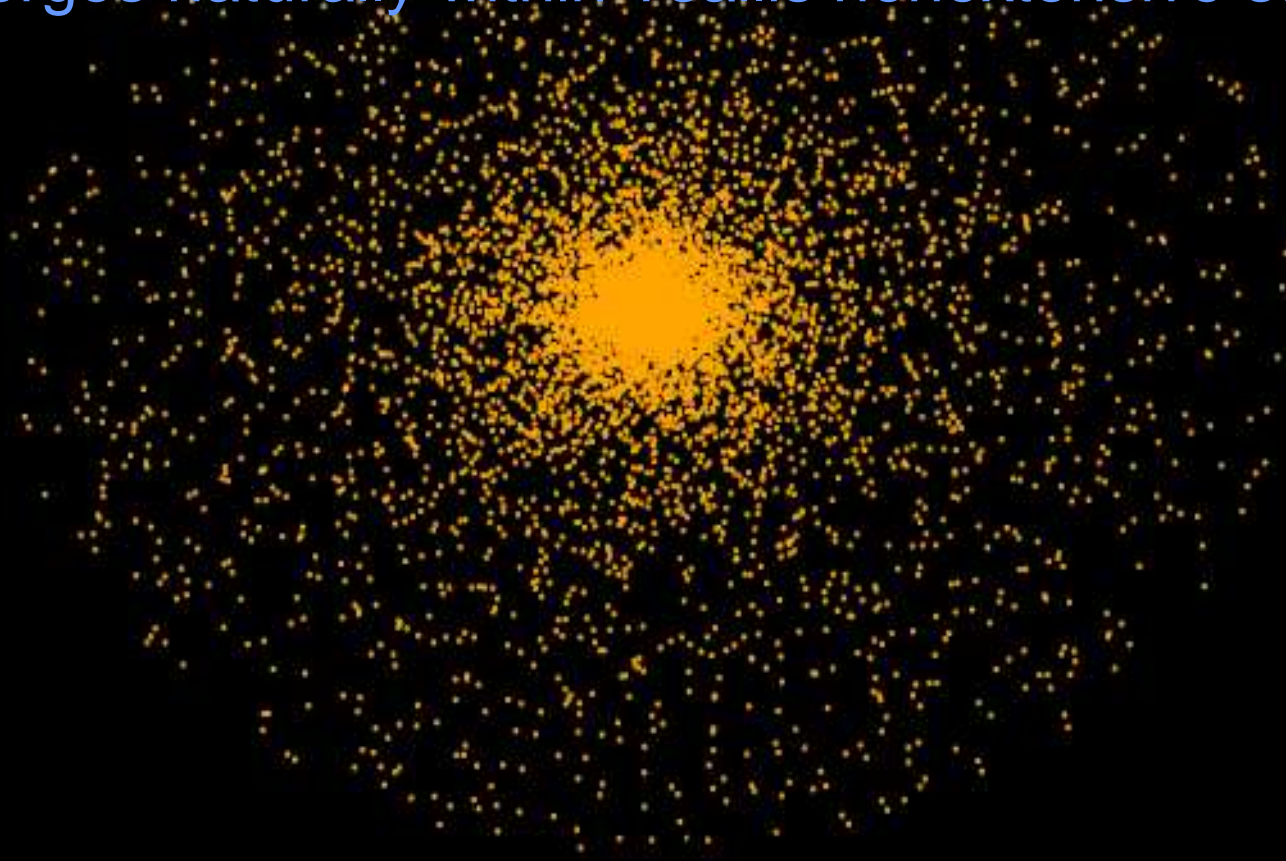
(b)

- We study the effect of competition between the relevant variables: connectivity k , fitness η and metrics r .
- The fitness may give the possibility to the younger nodes to compete equally with the older ones, when the younger node gets a high fitness.
- By including metrics favors the linking between first neighbors.
- The average connectivity $\langle k_i \rangle$ is appreciably influenced by metrics and by fitness, while the average path length $\langle l \rangle$ keeps approximatively the same.

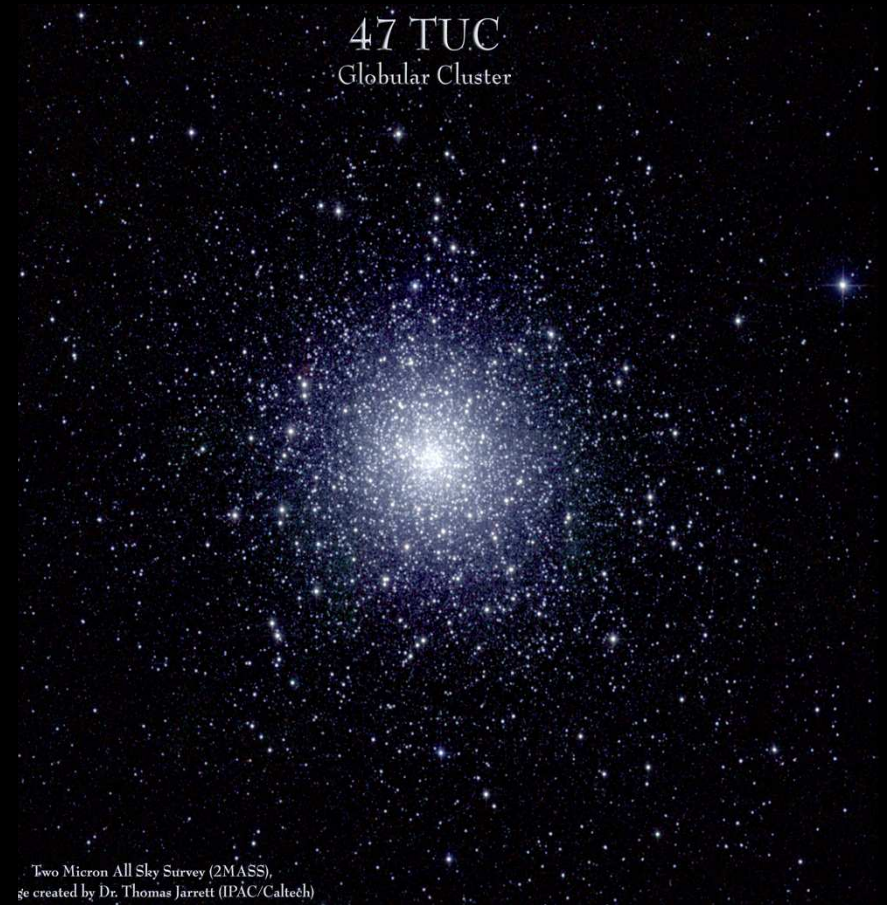
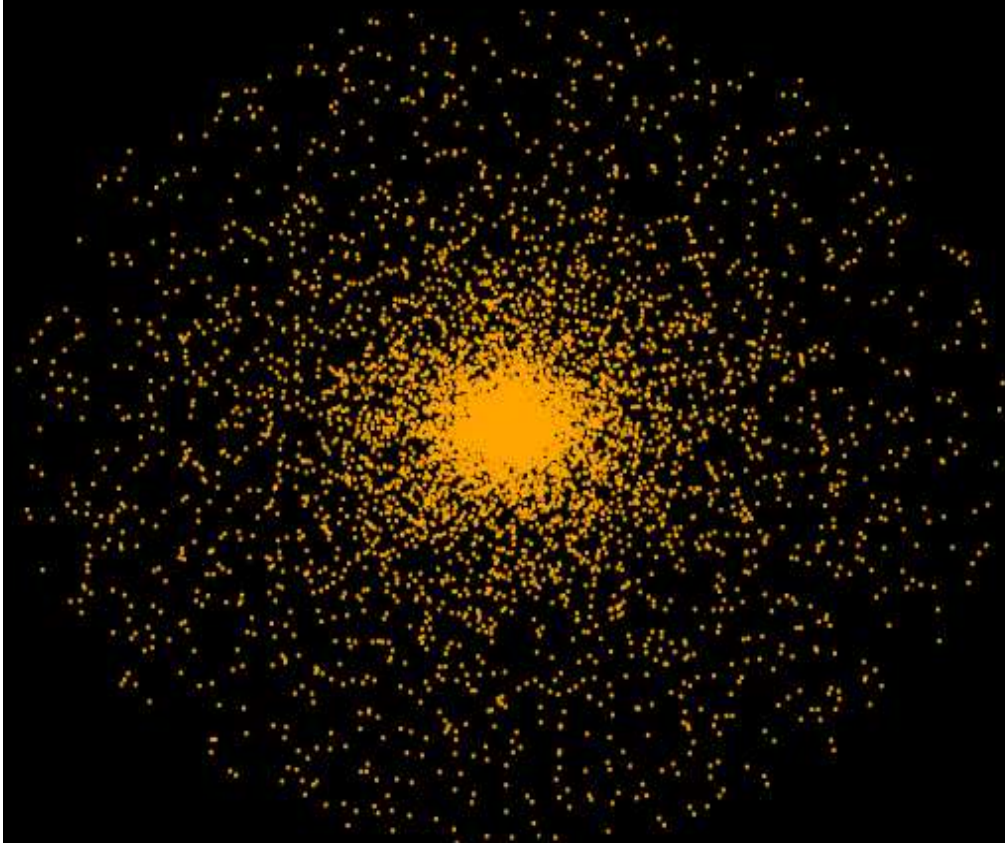
Summary

(c)

- The degree distribution $P(k)$ of the present generalized model appears to be the q -exponential function that emerges naturally within Tsallis nonextensive statistics.



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References

1. D. J. B. Soares, C. Tsallis, A. M. Mariz, and L. R. da Silva.
“Preferential Attachment Growth Model and Nonextensive Statistical Mechanics”
Europhysics Letters **70**, 70 (2005)
2. J. S. Andrade Jr., H. J. Herrmann, R.F. Andrade and L. R. da Silva.
Apollonian Networks: Simultaneously Scale-free, Small World, Euclidean, Space Filling and with Matching Graphs.
Physical Review Letters **94**, 018702 (2005).
3. M. D. de Meneses, Sharon D. da Cunha, D.J.B. Soares and L. R. da Silva.
“Preferential Attachment Scale-free Growth Model with Random Fitness and Connection with Tsallis Statistics”
Progress of Theoretical Physics Supplement **162** 131 (2006)
4. D. J. B. Soares, J. S. Andrade Jr., H. J. Herrmann and L. R. da Silva
“Three Dimension Apollonian Networks”
International Journal of Modern Physics **17** 1219 (2006)
5. P. G. Lind, L. R. da Silva, J. S. Andrade Jr. and H. J. Herrmann.
“The Spread of Gossip in American Schools”
Europhysics Letters **78**, 68005 (2007)

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References

6. P. G. Lind, L. R. da Silva, J. S. Andrade Jr. and H. J. Herrmann.

“Spreading Gossip in Social Networks”

Physical Review E **76**, 036117 (2007)

7. S. B. Jácume, A. A. Moreira, J. S. Andrade L. R. da Silva and H. J. Herrmann

“Decomposition in Scale-free Networks”

Submetido para publicação em Physica A

8. G. A. Mendes and L. R. da Silva

Generating more realistic complex networks from power-law distribution of fitness

9. A. M. Filho, U. L. Fulco and L. R. da Silva

Epidemic Diffusion System in Scale-free Networks

10. M. L. de Almeida (Dissertação de Mestrado)

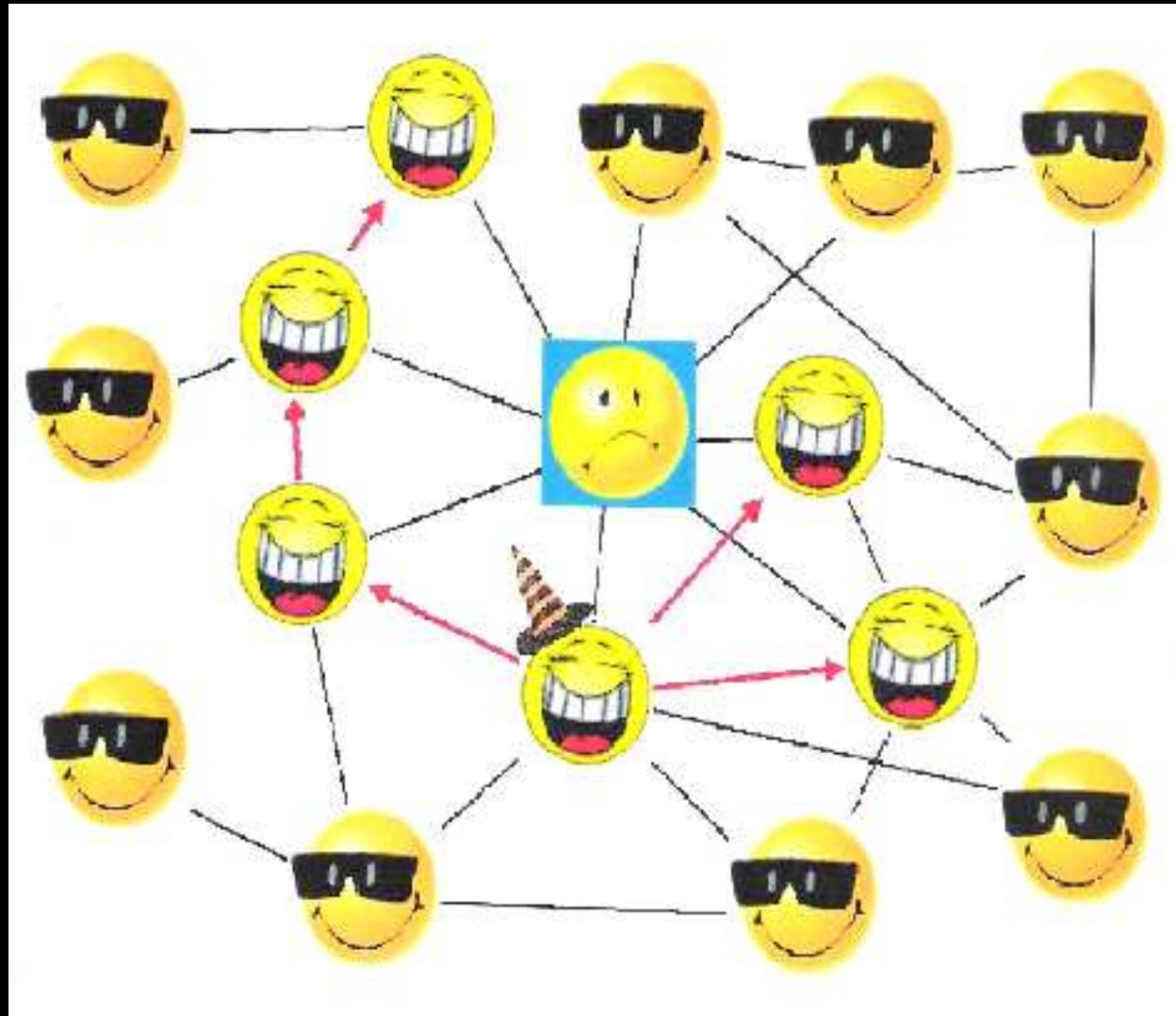
Dinâmica e Estrutura de Redes Complexas no Modelo de Afinidade

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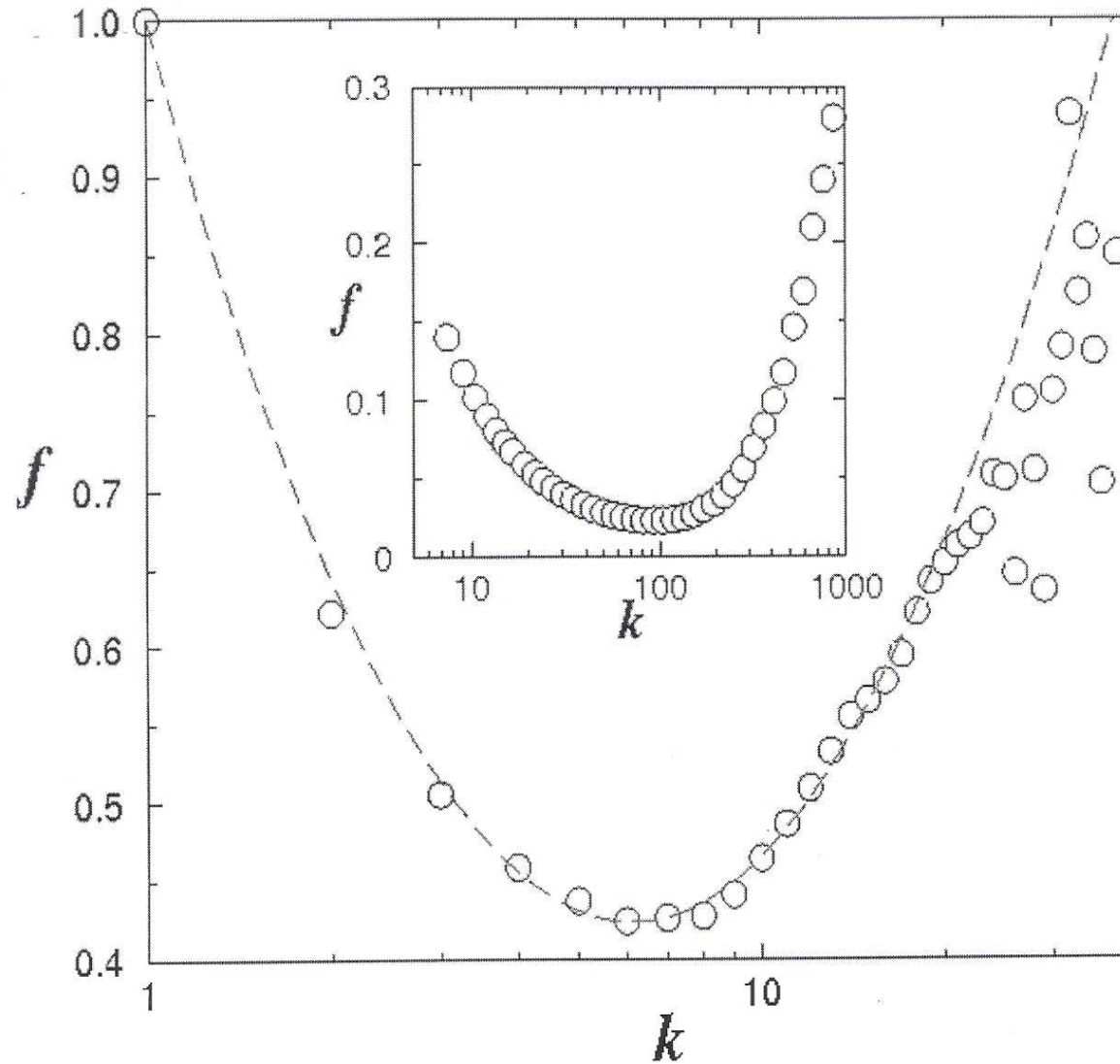
A large, dense field of small, bright yellow and orange particles arranged in a circular pattern on a black background. The particles are most concentrated in the center, creating a bright, glowing core that fades towards the edges. The overall appearance is that of a complex system or a large-scale simulation.

Thank you very much

6. How gossip propagates



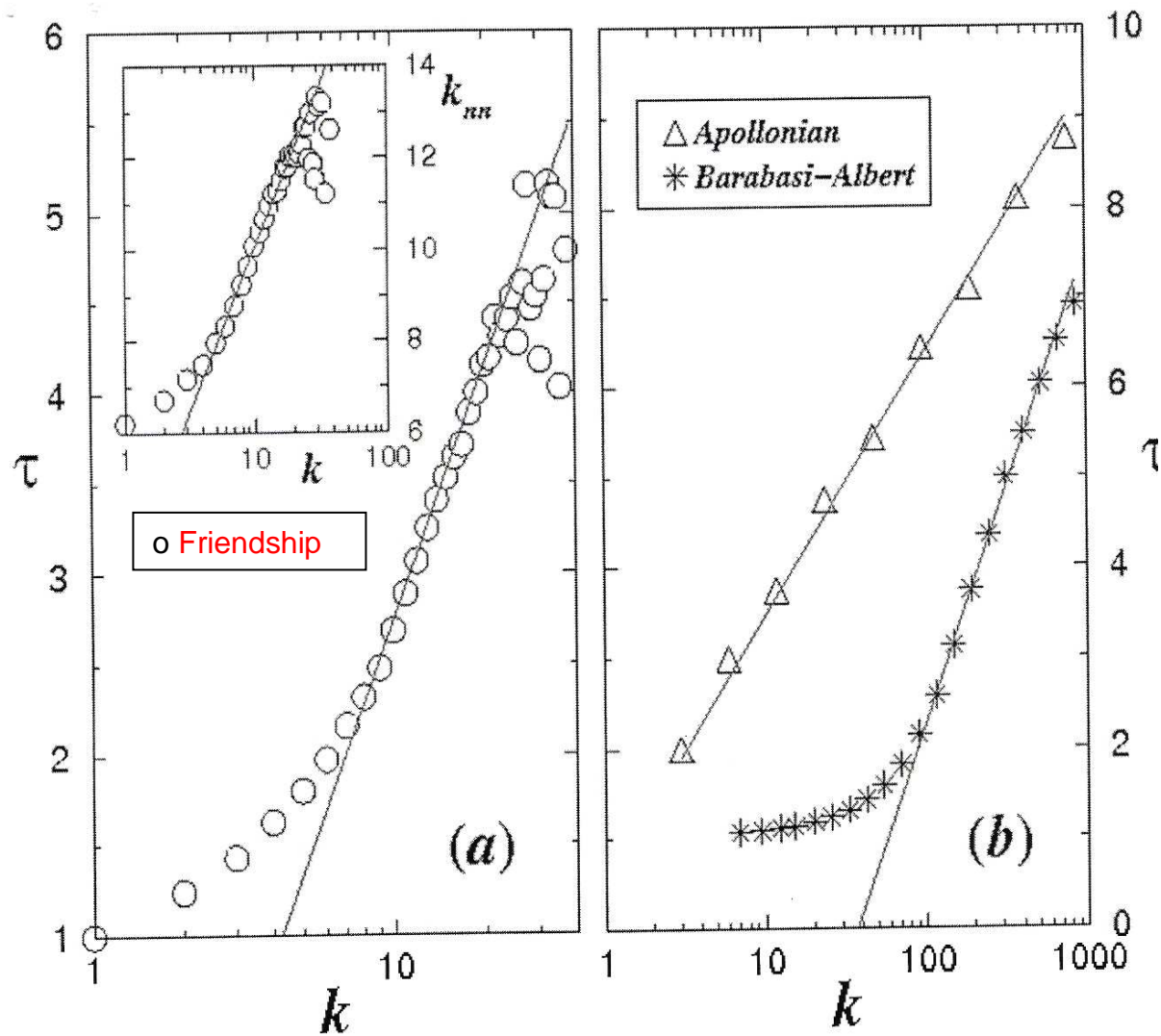
Spread Factor ($f=n_f/k$)



U.S Friend Schools

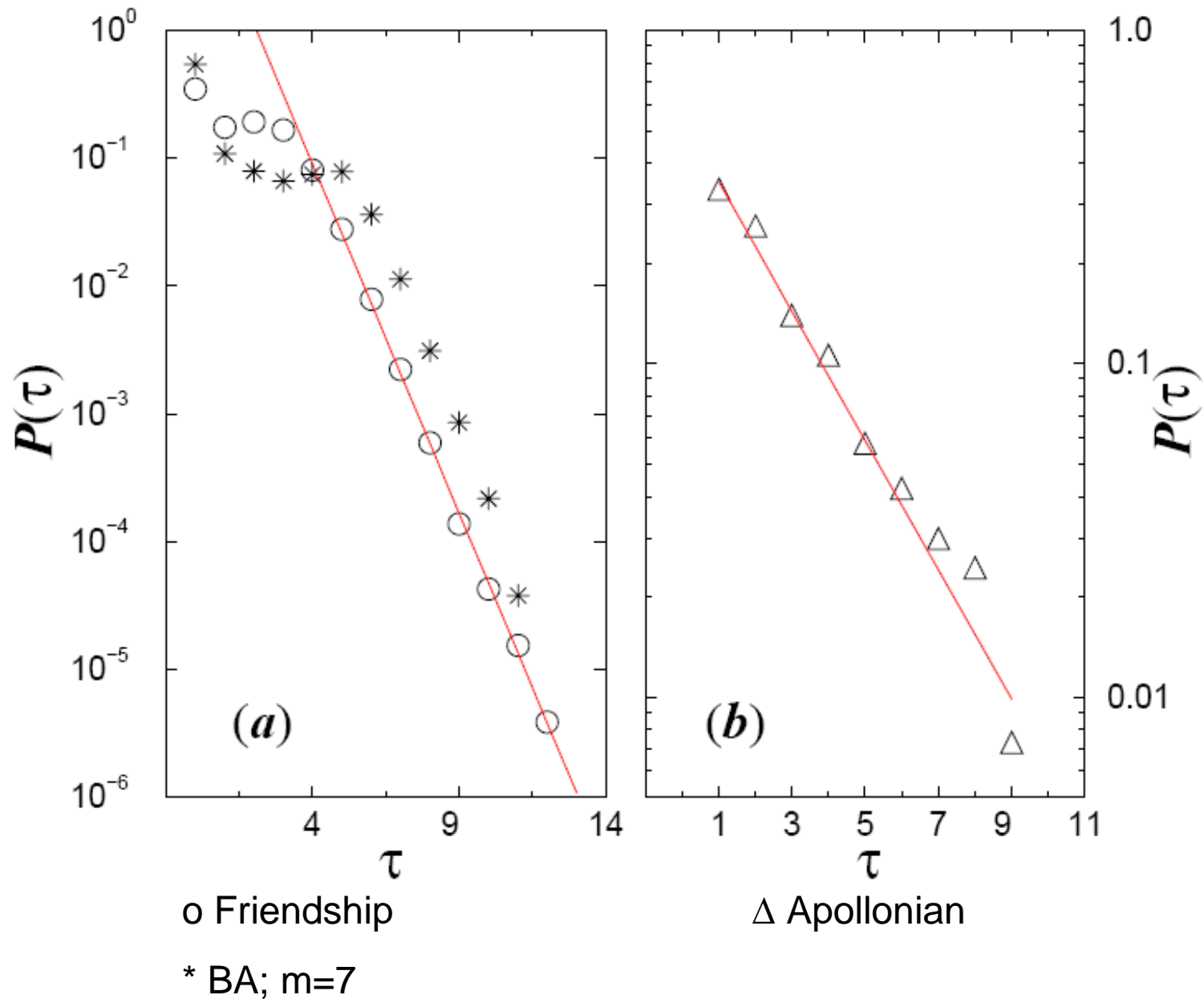
Inset graph: Barabási-Albert

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$$\tau = A + B \log(k)$$

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GAS-LIKE (NODE COLLAPSING) NETWORK:

S. Thurner and C. T., Europhys Lett **72**, 197 (2005)

Number N of nodes fixed (*chemostat*); $i=1, 2, \dots, N$

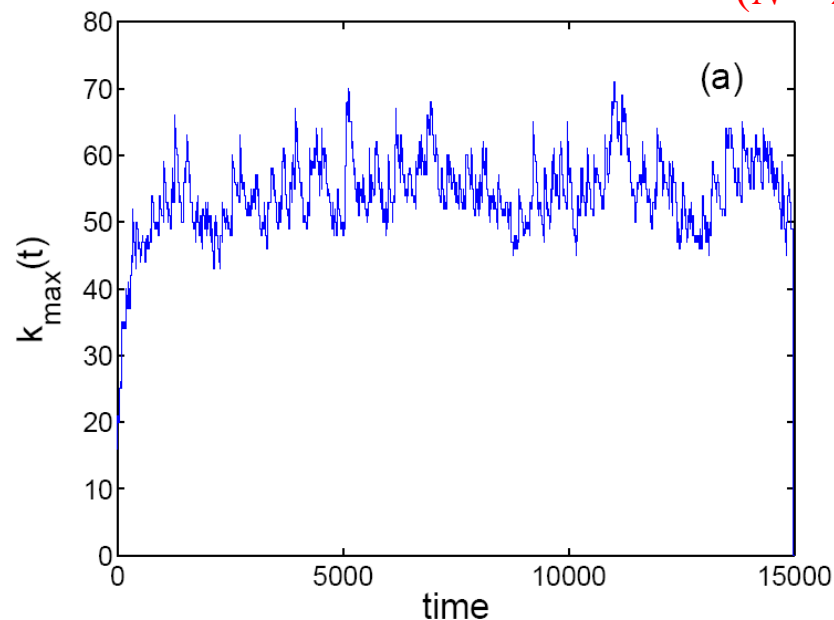
Merging probability $p_{ij} \propto \frac{1}{d_{ij}^\alpha}$ ($\alpha \geq 0$)

$d_{ij} \equiv$ shortest path (chemical distance) connecting nodes i and j on the network

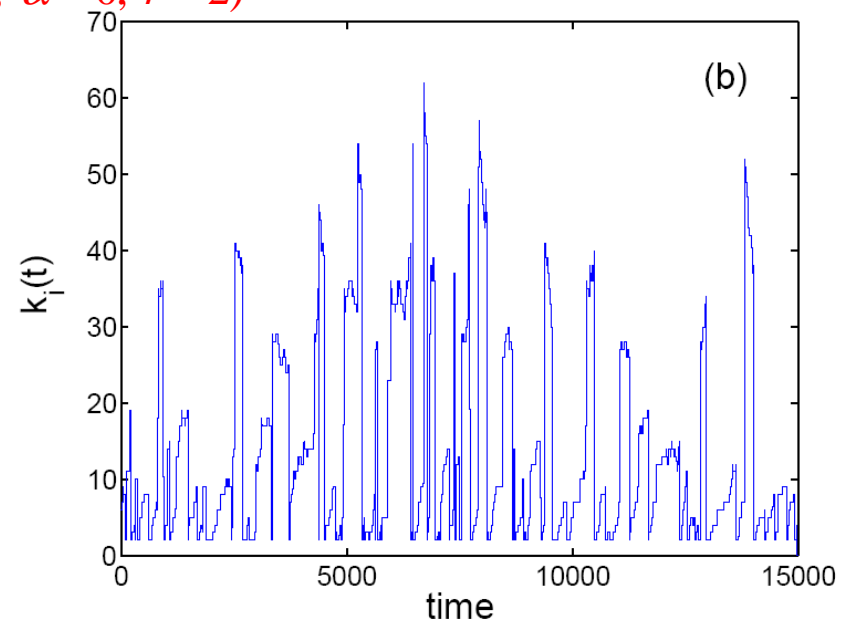
$\alpha = 0$ and $\alpha \rightarrow \infty$ recover the *random* and the *neighbor* schemes respectively

(Kim, Trusina, Minnhagen and Sneppen, *Eur. Phys. J. B* 43 (2005) 369)

($N = 2^7$; $\alpha = 0$; $r = 2$)

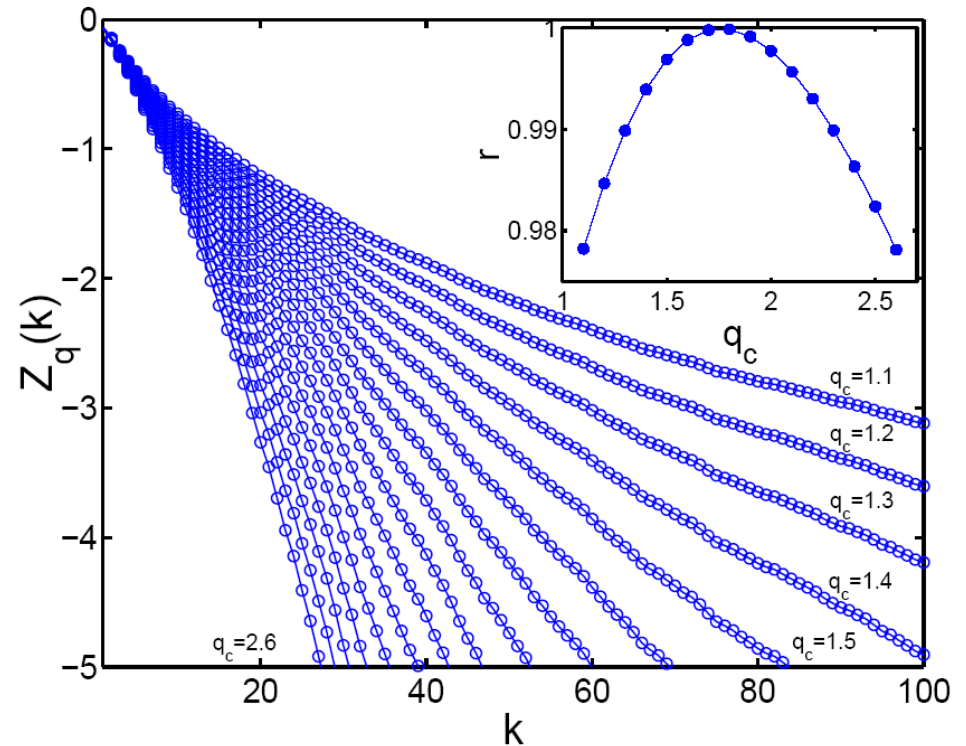
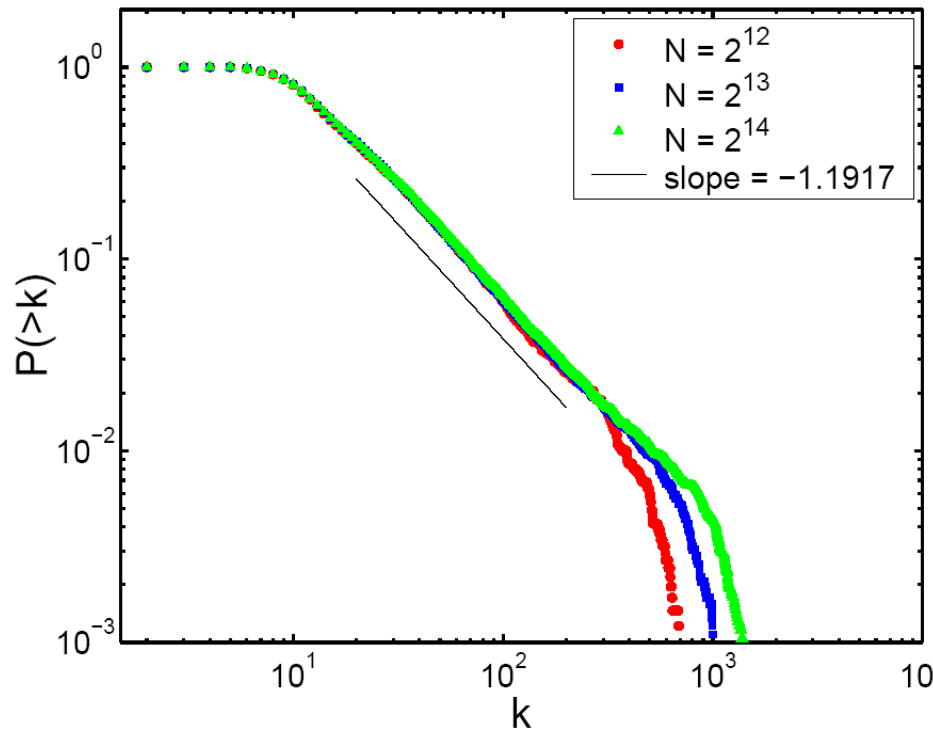


Degree of the most connected node



Degree of a randomly chosen node

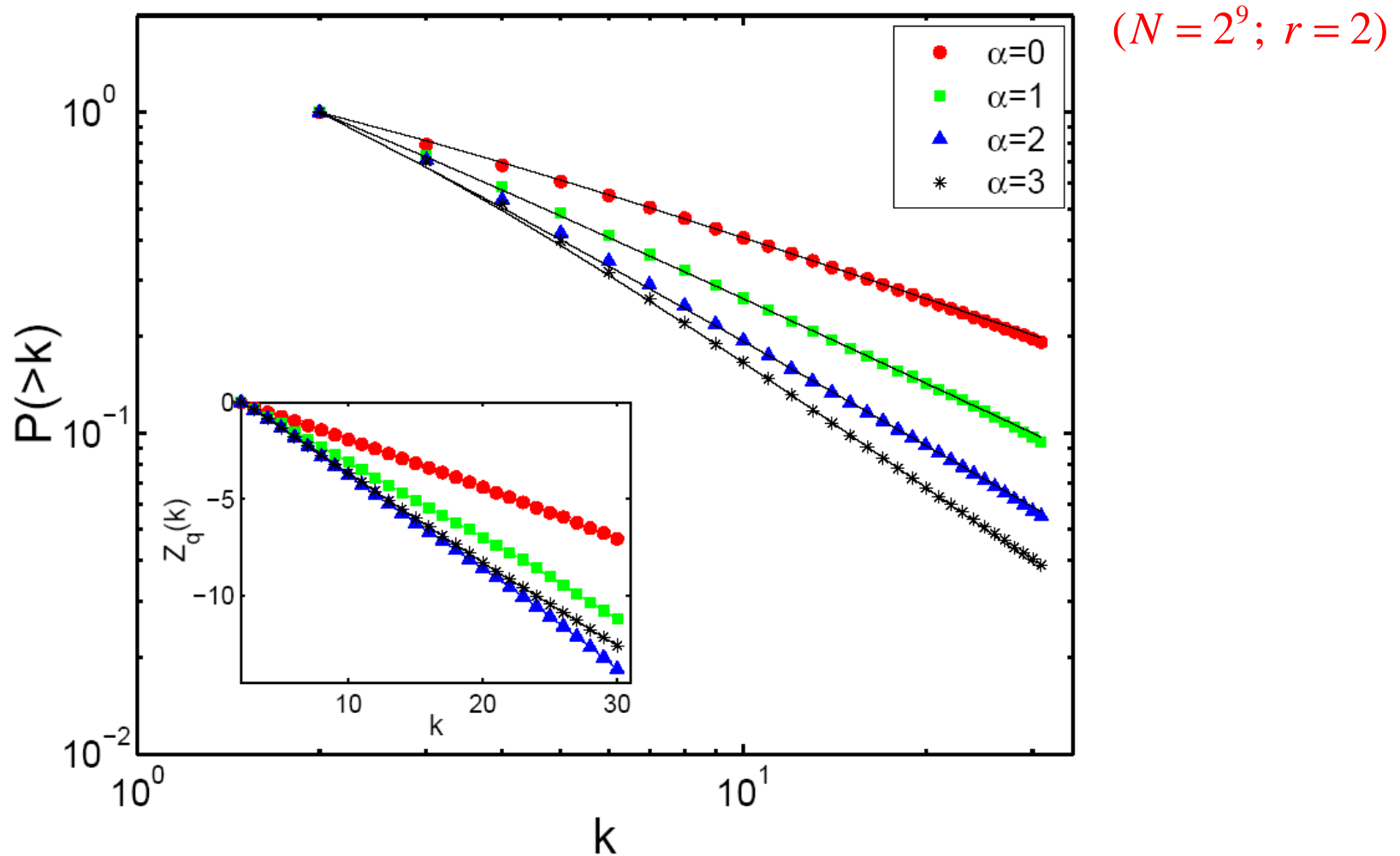
$(\alpha \rightarrow \infty ; \langle r \rangle = 8)$



$$Z_q(k) \equiv \ln_q [P(>k)] \equiv \frac{[P(>k)]^{1-q} - 1}{1-q}$$

$(\text{optimal } q_c = 1.84)$

S. Thurner and C. T., Europhys Lett 72, 197 (2005)

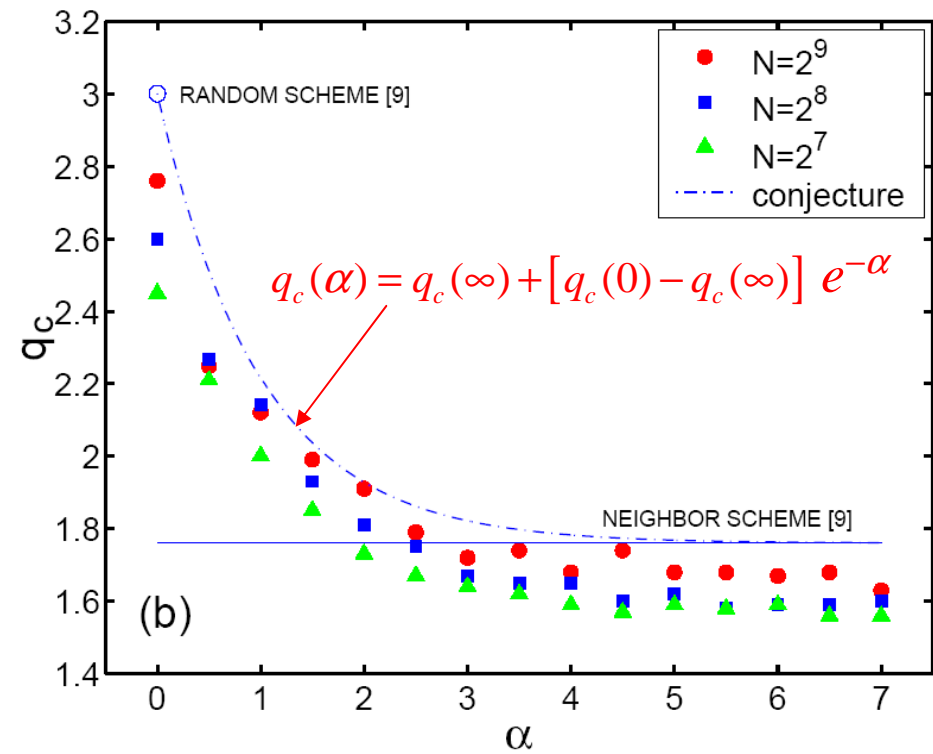
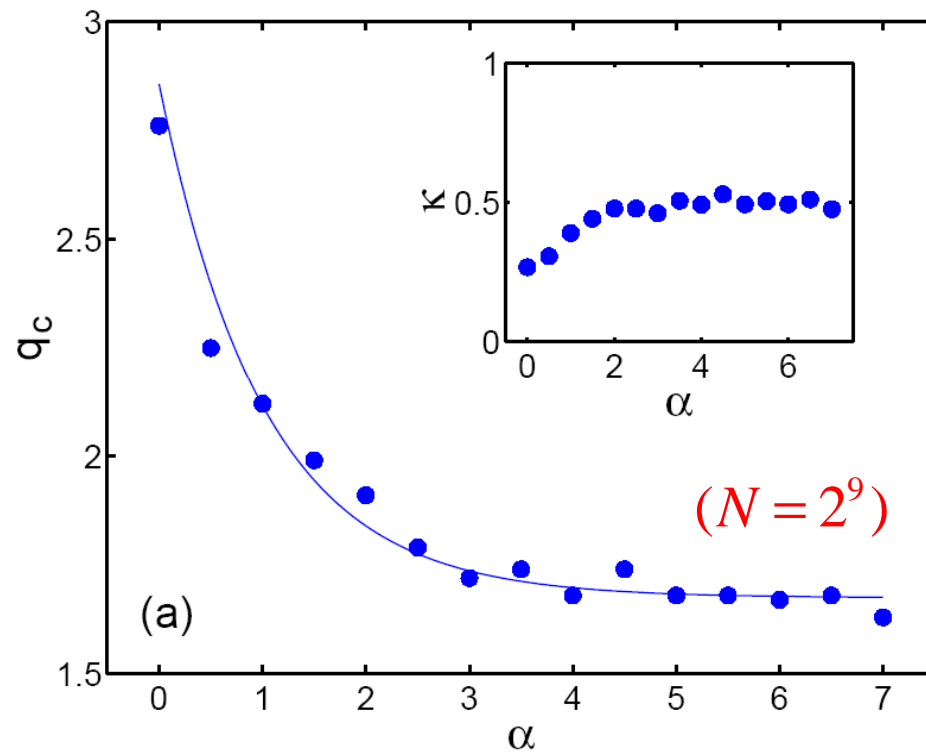


$$P(\geq k) = e_{q_c}^{- (k-2)/\kappa} \quad (k = 2, 3, 4, \dots)$$

linear correlation $\in [0.999901, 0.999976]$

S. Thurner and C. T., Europhys Lett 72, 197 (2005)

$(r = 2)$



S. Thurner and C. T., Europhys Lett 72, 197 (2005)