Degree distribution in random Apollonian networks structures

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joint work with Michèle Soria

ALÉA 2007

Plan

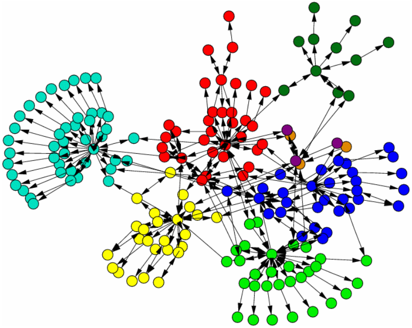
Real-life Networks

Application domains

- Computer Science
- Biology
- Sociology
- ...

Models

- Needed to simulate real-life networks
- Simple classes of random graphs not a good model



Web site

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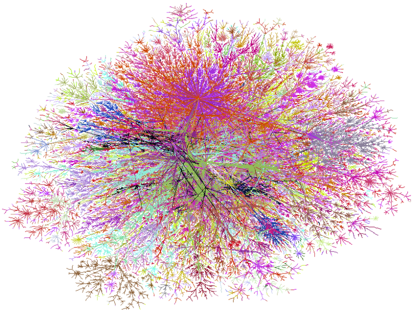

Internet

Real-life Networks

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- . . .

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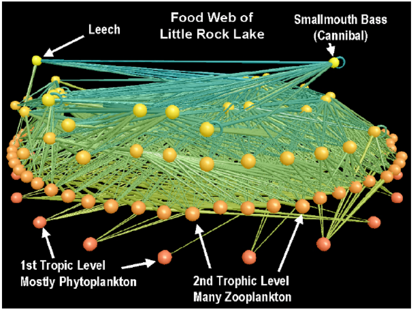

Food web

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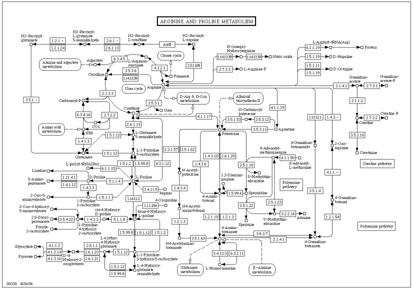

Metabolism

Real-life Networks

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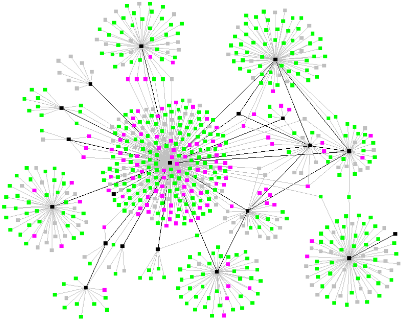

Contagion of diseases

Real-life Networks

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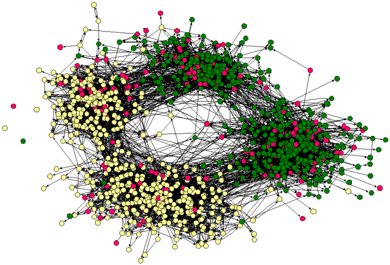

Friendship

Real-life Networks

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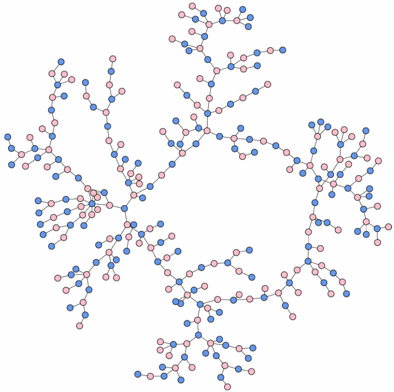


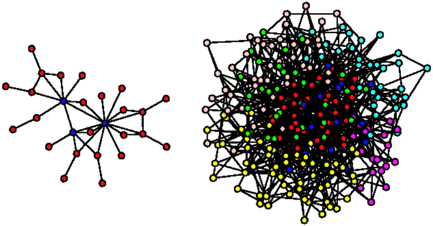
Dating

Distinctive properties of real-life graphs

- Number of edges
  - Of the same order as the number of vertices
- Connectivity
  - Strong (Giant component)
- Degree distribution
  - Heavy tailed (Power law, Scale-free)
- Mean distance
  - Small
- Clustering
  - Strong

Existing models

Scale-free networks

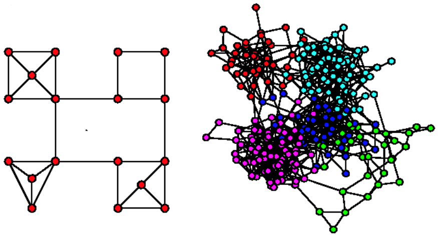


A.-L. Barabási & R. Albert
Emergence of scaling in random networks
Science **286**, 509 (1999)

- Number of edges
- Connectivity
- Degree distribution
- Mean distance
- Clustering

Existing models

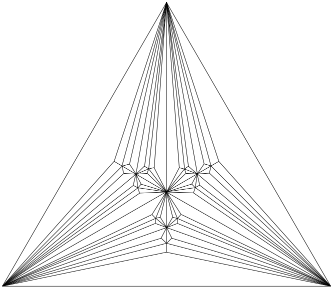
“Small world” networks



Watts D. J. & Strogatz S. H.
Collective dynamics of “small-world” networks
Nature **393**, 440 (1998)

- Number of edges
- Connectivity
- Degree distribution
- Mean distance
- Clustering

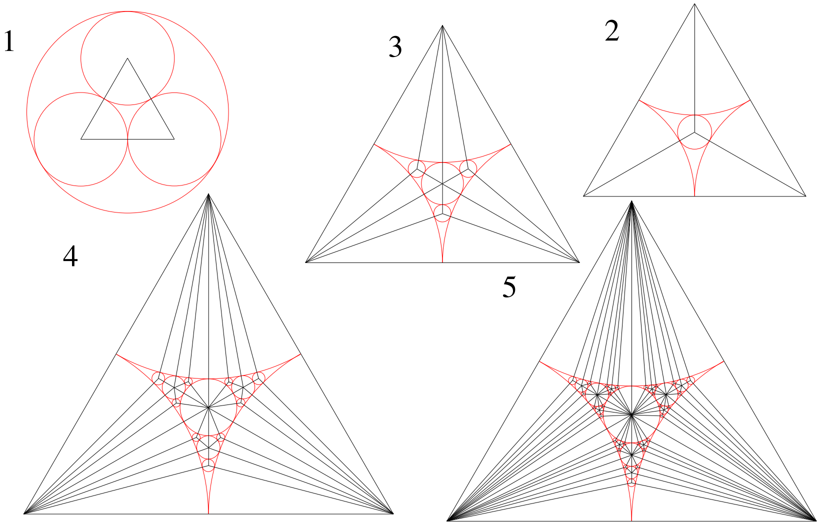
Apollonian networks



- All properties satisfied
- Inspired from the apollonian packings
- Model is deterministic

J. S. Andrade, Jr., H. J. Herrmann,
R. F. S. Andrade & L. R. da Silva
Apollonian Networks : Simultaneously Scale-Free, Small World, Euclidean, Space Filling, and with Matching Graphs
Phys. Rev. Lett. **94**, 018702 (2005)

Apollonian packings, Apollonian networks



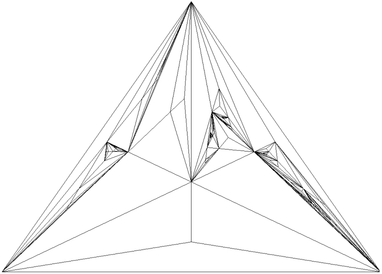
A randomized variation

random Apollonian networks

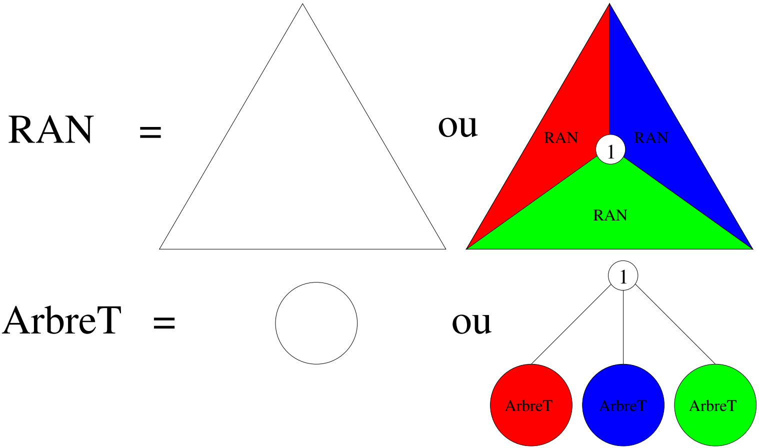
Tao Zhou, Gang Yan & Bing-Hong Wang
Maximal planar networks with large clustering coefficient and power-law degree distribution
Physical Review E **71**, 046141 (2005)

Algorithm

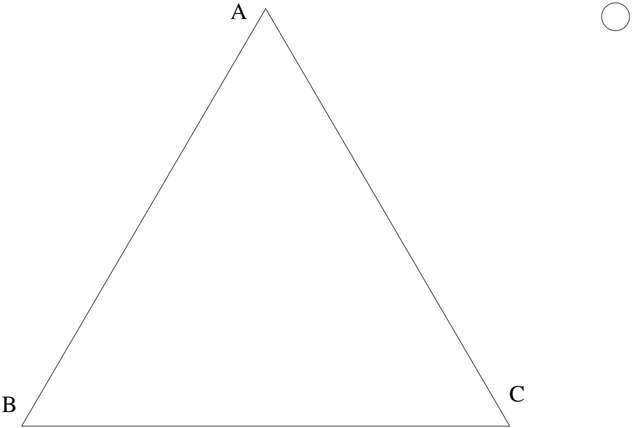
- Initial state : a triangle
- Iterative state : Choose a triangle and add to it a point and link it to the three vertices of the triangle



A bijection with ternary increasing trees



The bijection
Random Apollonian Networks $\leftrightarrow$ Ternary Inc. Trees



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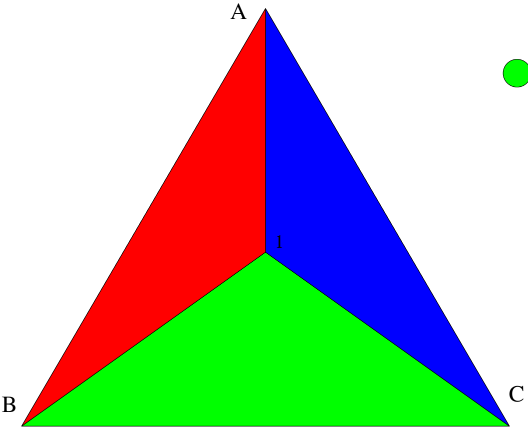

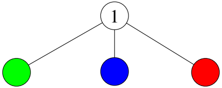

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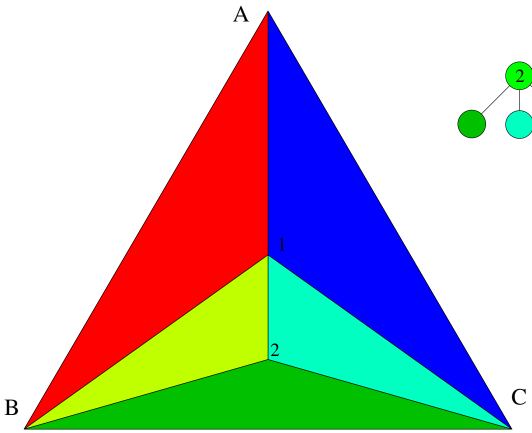

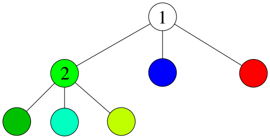

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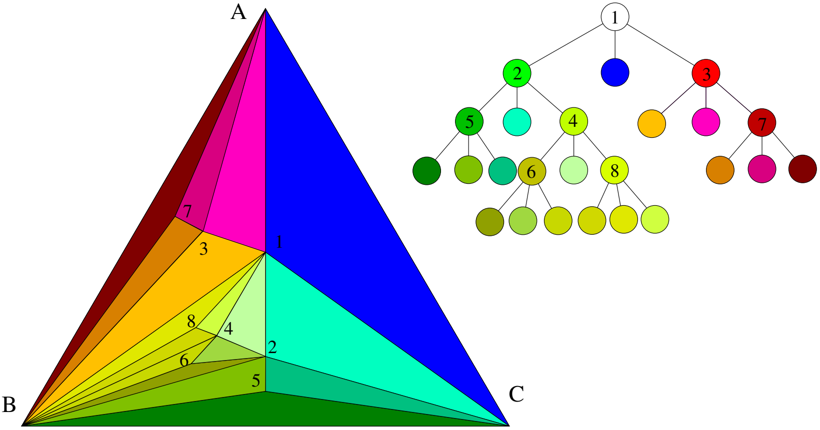


Random Apollonian network structures

Replace Ternary Increasing Trees with Ternary Trees

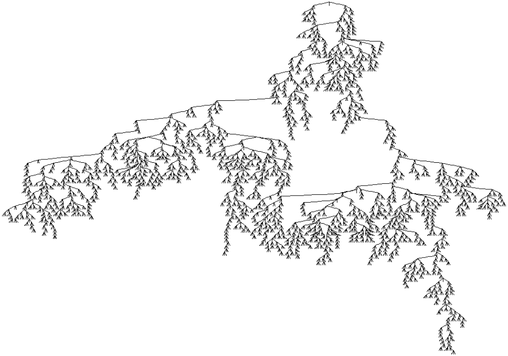
Properties

- Same bijection
- Same class of graphs
- Different probability distribution
- Similar properties
- Simple combinatorial description of the model

What for ?

- General methods for sampling
- Efficient generation (Boltzmann)
- Greater flexibility

Ternary tree generation using the Boltzmann model



The Boltzmann model

Specifiable combinatorial classes

- Basic operations :
  Union, **Product**, Sequence, Cycle, Set
- **Recursive definitions**

Properties

- Uniform generation
- Approximate size
- Efficiency

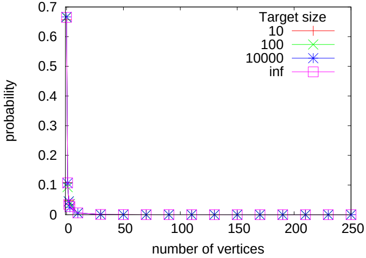
P. Duchon, P. Flajolet, G. Louchard, G. Schaeffer
Boltzmann samplers for the random generation of combinatorial structures

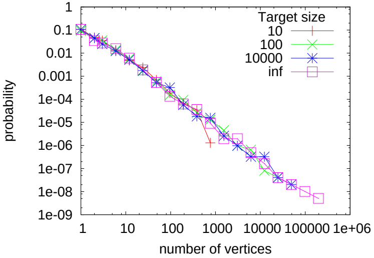
Algorithm for the generation of a ternary tree

$$T(z) = z + zT(z)^3$$

Algorithm : TernaryTree(p)

if rand(0..1) < p **then**
  Leaf
else
  Node(TernaryTree(p),TernaryTree(p),TernaryTree(p))
end if





Bivariate generating functions

$$C(z,u) = \sum_{n,k} C_{n,k} u^k z^n \qquad C(z) = \sum_n C_n z^n$$

Distribution of a parameter, fixed n

$$\Pr(\Omega_n = k) = \Pr\left(\Omega = k / N = n\right) = \frac{C_{n,k}}{C_n} = \frac{[z^n u^k] C(z,u)}{[z^n] C(z)}$$

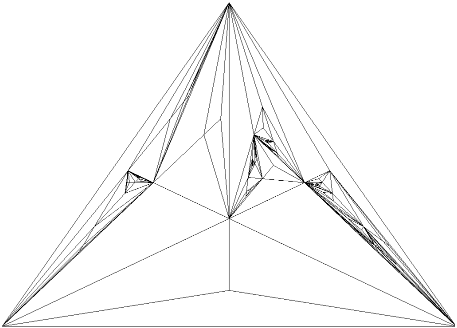
Distribution of a parameter under Boltzmann sampling

$$\Pr(\Omega = k) = \sum_n \Pr\left(\Omega = k / N = n\right) \times \Pr(N = n)$$

$$= \sum_n \frac{C_{n,k}}{C_n} \times \frac{C_n x^n}{C(x)} = \frac{\sum_n C_{n,k} x^n}{C(x)} = \frac{[u^k] C(x,u)}{C(x,1)}$$

subcritical $\quad \Pr(\Omega_n = k) \to Ck \Pr(\Omega = k)$ when $n \to \infty$

Back to the network properties



Properties of the generated networks

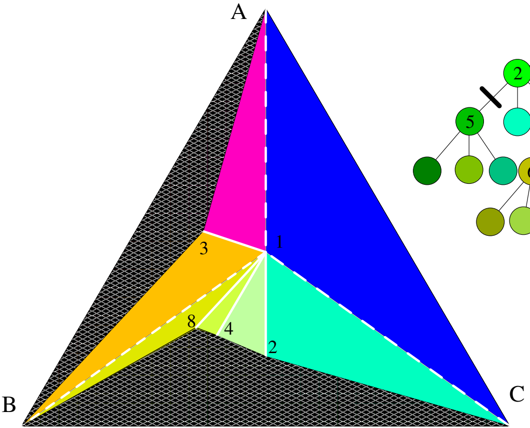
By construction :

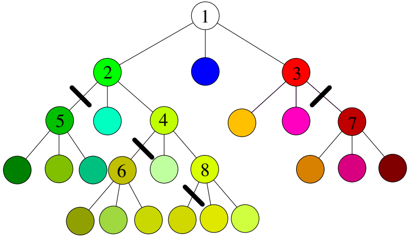
- Number of edges
  Equal to $3v - 6$, where v the number of vertices
- Connectivity
  A single component
- Mean degree
  ~ 6

Needing further investigation :

- Degree distribution
- Clustering
- Mean distance

Degree





u marks the neighbors

- of the center (root) : $RD(z,u) = zu^3T^3(z,u)$
- of an external node : $T(z,u) = 1 + zuT^2(z,u)T(z)$

Degree distribution

u marks the neighbors

- of the center (root) : $RD(z,u) = zu^3 T^3(z,u)$
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Proposition : Statistical properties

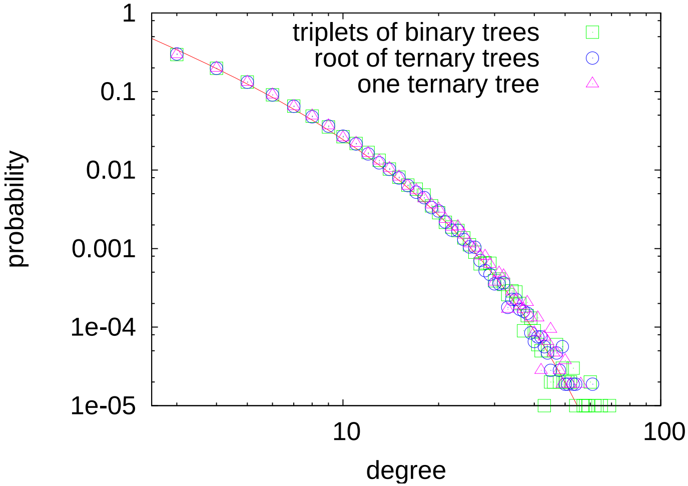
Same for :

- the set of all subtrees of a random tree
- a set of random trees independently generated with a Boltzmann sampler

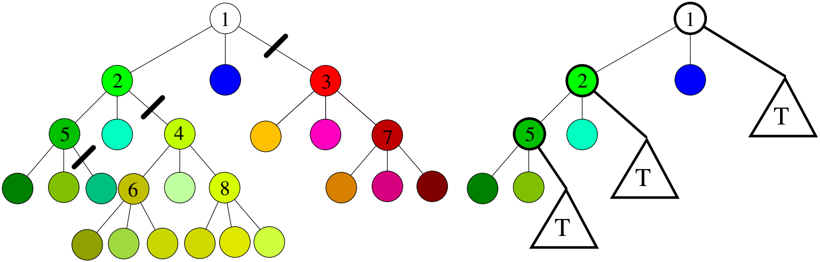
Degree distribution

Mean value 6 and a Catalan form for the pgf :

$$\Pr(D = 3 + k) = \frac{8}{9}\,\frac{1}{k+3}\binom{2k+2}{k} \sim C\,\left(\frac{8}{9}\right)^k\,(k+3)^{-3/2}$$

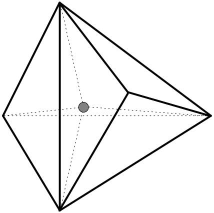


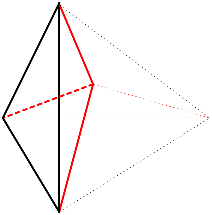
Sketch of proof



- Ternary trees marked for degree :
  $T(z,u) = 1 + zuT^2(z,u)T(z)$
- Simulated by binary trees :
  $T(z,u) = B(zuT(z))$, where $B(t) = \sum B_n t^n$
- Schema is subcritical : $\rho\tau < 1/4$
- $[u^k]B(zuT(z)) = \rho^k \tau^k \frac{1}{k+1}\binom{2k}{k}$

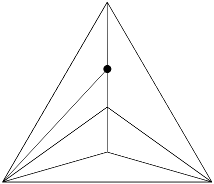
Conclusion

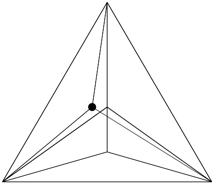




More flexibility :

Variants

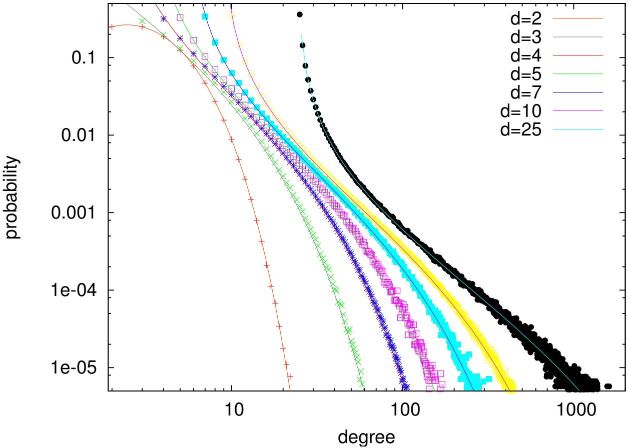




Higher dimension RANS

$$\Pr(D_d = d + k) \sim C\alpha^k \left(k + \frac{d}{d-2}\right)^{-\frac{3}{2}}$$

$$RD_d(z,u) = zu^d T_d^d(z,u) \qquad\qquad T_d(z,u) = T_{d-1}(uzT_d(z))$$

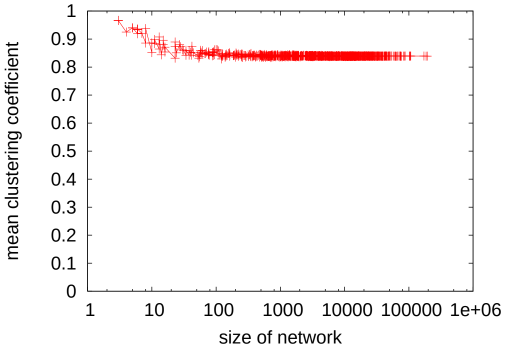


Clustering

Definition : Clustering coefficient of a vertex of degree k

$$C(k) = \frac{\text{number of links between neighbors}}{k(k-1)}$$

- $C(k) = 3\frac{2k-d-1}{k(k-1)}$
- Mean value over all vertices independent of size



Mean distance

Simulation confirms a small mean distance (order $\sqrt{N}$)

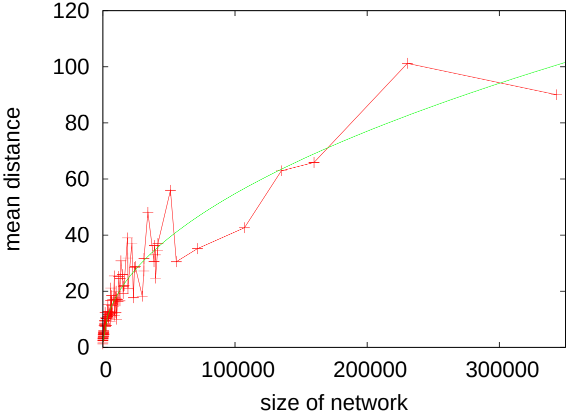


Image references

From http ://www-personal.umich.edu/ Emejn/networks/

Web site	M. E. J. Newman and M. Girvan, *Finding and evaluating community structure in networks*, Physical Review E 69, 026113 (2004).
Internet	Hal Burch and Bill Cheswick, Lumeta Corp.
Food web	Neo Martinez and Richard Williams.
Contagion of diseases	Valdis Krebs, www.orgnet.com.
Friendship	James Moody, *Race, school integration, and friendship segregation in America*, American Journal of Sociology 107, 679-716 (2001).
Dating	Data drawn from Peter S. Bearman, James Moody, and Katherine Stovel, *Chains of affection : The structure of adolescent romantic and sexual networks*, American Journal of Sociology 110, 44-91 (2004), image made by Mark Newman.

Scale-free & small world	E. Ravasz, A. L. Somera, D. A. Mongru, Z. N. Oltvai, A.-L. Barabási *Hierarchical Organization of Modularity in Metabolic Networks* Science 297, 1551 (2002).