

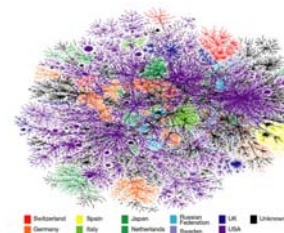
Large Complex Networks: Deterministic Models (Recursive Clique-Trees)



<http://www.caida.org/tools/visualization/plankton/>

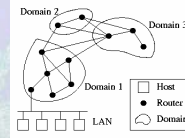
Francesc Comellas

Departament de Matemàtica Aplicada IV,
Universitat Politècnica de Catalunya, Barcelona
comellas@ma4.upc.edu



WWW

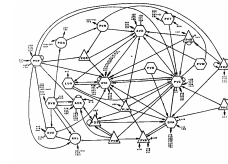
Internet



Air routes



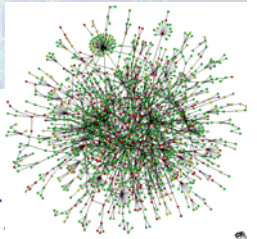
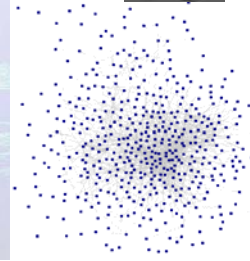
Power grid



C. Elegans



Erdős number



Proteins

Complex systems

Different elements (nodes)
Interaction among elements (links)

Complex networks

Mathematical model: **Graphs**

Real networks very often are

Large
Small-world
small diameter $\log(|V|)$, large clustering

Scale-free
power law degree distribution ("hubs")

Self-similar / fractal

Deterministic models

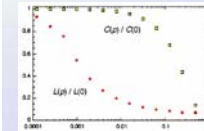
Based on cliques
(hierarchical graphs, recursive
clique-trees, Apollonian graphs)

Most "real" networks are

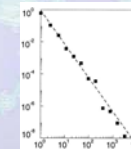
small-world

scale-free

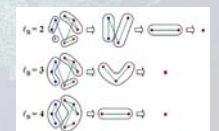
self-similar



Small diameter (logarithmic)
Milgram 1967
High clustering
Watts &
Strogatz 1998



Power law
(degrees)
Barabási &
Albert 1999



Fractal
Song, Havlin &
Makse
2005,2006

Main parameters (invariants)

Diameter - average distance

Degree

Δ degree.
 $P(k)$: Degree distribution.

Clustering

Are neighbours of a vertex also neighbours among them?

Small-world networks

small diameter (or average dist.)
high clustering

6 degrees of separation ! Stanley Milgram (1967)
160 letters Omaha -Nebraska- -> Boston



Small world phenomenon in social networks
What a **small-world** !

Structured graph

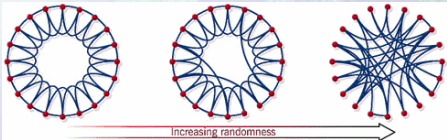
- high clustering
- large diameter
- regular

Small-world graph

- high clustering
- small diameter
- almost regular

Random graph

- small clustering
- small diameter

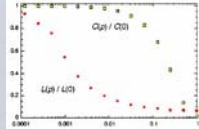


$|V|=1000$ $\Delta=10$
 $D=100$ $d=49.51$
 $C=0.67$

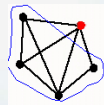
$|V|=1000$ $\Delta=8-13$
 $D=14$ $d=11.1$
 $C=0.63$

$|V|=1000$ $\Delta=5-18$
 $D=5$ $d=4.46$
 $C=0.01$

Watts & Strogatz,
 Collective dynamics of "small-world" networks,
 Nature 393, 440-442 (1998)



Network characteristics



Clustering $C(v) = \frac{\# \text{ of links among neighbors}}{n(n-1)/2}$

Diameter or Average distance

Maximum communication delay

Degree distribution

Resilience

Real life networks are clustered, large $C(p)$, but have small average distance $L(p)$. Very often they are also scale-free

	L	L _{rand}	C	C _{rand}	N
WWW	3.1	3.35	0.11	0.00023	153127
Actors	3.65	2.99	0.79	0.00027	225226
Power Grid	18.7	12.4	0.080	0.005	4914
C. Elegans	2.65	2.25	0.28	0.05	282



Erdős number

<http://www.oakland.edu/enp/>

1- 509
 2- 7494

N= 268,000 Jul 2004
 (connected component)

D=23 R=12 $D_{avg} = 7.64$
 $\delta=1$ $\Delta=509$ $\Delta_{avg} = 5.37$
 C = 0.14

Erdős number 0 --- 1 person
 Erdős number 1 --- 504 people
 Erdős number 2 --- 6593 people
 Erdős number 3 --- 33605 people
 Erdős number 4 --- 83642 people
 Erdős number 5 --- 87760 people
 Erdős number 6 --- 40014 people
 Erdős number 7 --- 11591 people
 Erdős number 8 --- 3146 people
 Erdős number 9 --- 819 people
 Erdős number 10 --- 244 people
 Erdős number 11 --- 68 people
 Erdős number 12 --- 23 people
 Erdős number 13 --- 5 people

(MathSciNet Jul 2004)

Notable Erdős coauthors:

- Frank Harary (257 coauthors)
- Noga Alon (143 coauthors)
- Saharon Shelah (136)
- Ronald Graham (120)
- Charles Colbourn (119)
- Daniel Kleitman (115)
- A. Odlyzko (104)

Erdős did not write a joint paper with his PhD advisor,
 Leopold Fejér

Some other Erdős coauthors

articles with Erdős

- András Sárközy 57
- András Hajnal 54
- Ralph Faudree 45
- Richard Schelp 38
- Vera Sós 34
- Alfréd Rényi 32
- Cecil C. Rousseau 32
- Pál Turán 30
- Endre Szemerédi 29
- Ronald Graham 27
- Stephan A. Burr 27
- Joel Spencer 23
- Carl Pomerance 21
- Miklos Simonovits 21
- Ernst Straus 20
- Melvyn Nathanson 19

- Richard Rado 18
- Jean Louis Nicolas 17
- Janos Pach 16
- Béla Bollobás 15
- Eric Milner 15
- John L. Selfridge 13
- Harold Davenport 7
- Nicolaas G. de Bruijn 6
- Ivan Niven 7
- Mark Kac 5
- Noga Alon 4
- Saharon Shelah 3
- Arthur H. Stone 3
- Gabor Szegő 2
- Alfred Tarski 2
- Frank Harary 2
- Irving Kaplansky 2
- Lee A. Rubel 2

Fields medals

Alain Connes	1982	France	3
William Thurston	1982	USA	3
Shing-Tung Yau	1982	China	2
Simon Donaldson	1986	Great Britain	4
Gerd Faltings	1986	Germany	4
Michael Freedman	1986	USA	3
Lars Ahlfors	1936	Finland	4
Jesse Douglas	1936	USA	4
Laurent Schwartz	1950	France	4
Atle Selberg	1950	Norway	4
Kunihiko Kodaira	1954	Japan	2
Jean-Pierre Serre	1954	France	3
Klaus Roth	1958	Germany	2
Rene Thom	1958	France	4
Lars Norstrand	1962	Sweden	2
John Milnor	1962	USA	3
Michael Atiyah	1966	Great Britain	4
Paul Cohen	1966	USA	5
Alexander Grothendieck	1966	Germany	5
Stephen Smale	1966	USA	4
Alan Baker	1970	Great Britain	2
Heisuke Hironaka	1970	Japan	4
Serge Novikov	1970	USSR	3
John G. Thompson	1970	USA	3
Enrico Bombieri	1974	Italy	2
David Mumford	1974	Great Britain	2
Pierre Deligne	1978	Belgium	3
Charles Fefferman	1978	USA	2
Gregori Margulis	1978	USSR	3
Daniel Quillen	1978	USA	4
Richard Borcherds	1998	S Afr/Gc Brtn	2
William T. Gowers	1998	Great Britain	4
Maxim L. Kontsevich	1998	Russia	4
Curtis McMullen	1998	USA	3
Vladimir Voevodsky	2002	Russia	4
Laurent Lafforgue	2002	France	inf
Andrei Okounkov	2006	USA	3
Terence Tao	2006	USA	3
Wendelin Werner	2006	France	3

Winning bid: US \$1,831.80
 Ended: Apr-30-04 09:58:51 PDT
 Start time: Apr-20-04 09:58:51 PDT
 History: 33 bids (US \$0.09 starting bid)
 Winning bidder: madd_greg (1)
 Item location: Ann Arbor, MI United States (Detroit)
 Ships to: Worldwide

Seller information: jaggedy (1,009 ★) m20
 Feedback Score: 1609
 Positive Feedback: 99.8%
 Member since Jul-31-99 in United States

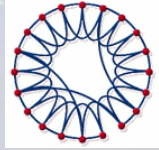
Pál Erdős (1913-1996)

Poisson distribution: $P(k)$ vs k with mean $\langle k \rangle$

Power law distribution: $P(k)$ vs k on a log-log scale

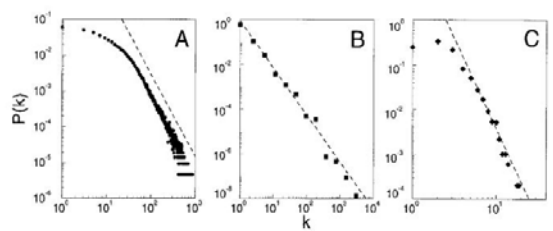
Scale-free networks

Scalability vs Fractality

1	1227	7	0	SWCirculant
2	1656	8	5	
3	1060	9	93	V =1000 Δ=8-13
4	401	10	806	
5	252	11	90	D = 14 d = 11.1
6	137	12	5	Small World
7	84	13	1	
8	46	14	0	C = 0.63
9	27			
10	26			
11	11			
12	5			
13	5			
14	3			
15	0			
16	0			
17	0			
18	1			
19	1			

Power grid
|V|=4491 δ=1 Δ=19
D = 46 d = 34.54
Small World
C = 0.08

A-L. Barabási i R. Albert,
Emergence of scaling in random networks.
Science 286, 509-510 (1999)



$$P(k) = k^{-\gamma}$$

A: actors	N=212.250	k=28.78	$\gamma=2.3$
B: WWW	N=325.729	k=5.46	$\gamma=2.67$
C: power grid	N= 4.94	k=2.67	$\gamma=4$

Real networks for which we know the topology:

$$P(k) \sim k^{-\gamma}$$

NON BIOLOGICAL	$\gamma > 2$
www (in)	$\gamma = 2.1$
actors	$\gamma = 2.3$
citations	$\gamma = 3$
power grid	$\gamma = 4$
BIOLOGICAL	$\gamma < 2$
yeast protein-protein net	$\gamma = 1.5, 1.6, 1.7, 2.5$
E. Coli metabolic net	$\gamma = 1.7, 2.2$
yeast gene expression net	$\gamma = 1.4-1.7$
gene functional interaction	$\gamma = 1.6$

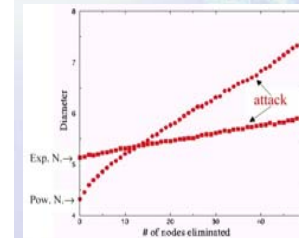
But, usual random models give: $P(k) \sim e^{-k}$

Interest on scale-free nets:

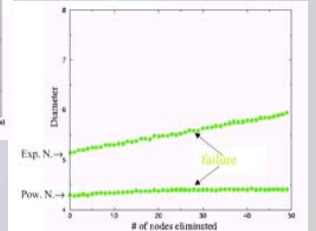
Resilience / Survival of the WWW

Albert, Jeong, Barabási
Nature 406, 378 (2000)

What happens when nodes fail randomly?



And when there are intentionate attacks to the best connected nodes?



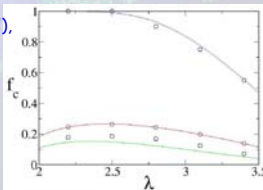
Epidemics spreading / "vaccination" WWW, social networks

R. Cohen, D. ben-Avraham, S. Havlin;
Efficient immunization of populations and computers
Phys. Rev. Lett. 91, 247901 (2003)

f_c threshold
 λ power law exponent
upper- totally random
lower- acquaintance immunisation (red),
double acq. imm. (green)

"vaccinate" high degree nodes !

method:
* select a node at random
* ask it to select a high degree node and immunize it



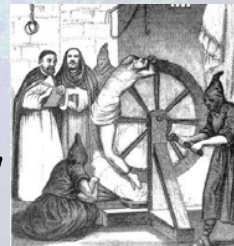
Search in power-law networks

Adamic, Lukose, Puniyani, Huberman;
Phys. Rev. E 64, 046135 (2001)

Even the Inquisition knew about scale-free networks!!

From random "vaccination"

Arnau d'Amaurí 1209. Besièrs
Caedite eos.
Robit enim Dominus qui sunt eius
Kill them all, God will know his own



to selection (see figure)

P. Ormerod, A.P. Roach;
The Medieval inquisition: scale-free networks and the suppression of heresy.
Physica A 339 (2004) 645-652

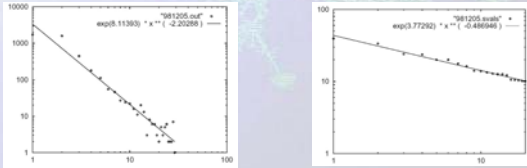
spectral properties

- Connectivity and vulnerability (diameter, cut sets, distances between subsets)
- Scalability, expansion (Cheeger constants)
- Routings (spanning trees)
- Load balancing
- Clustering (triangles)
- Reconstruction (Ipsen & Mikhaliiov, 2001)
- Dynamical aspects (interlacing theorem)

Experimental and simulation results

WWW / Internet eigenvalues

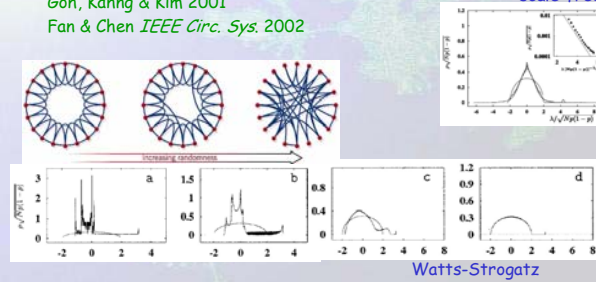
Faloutsos, Faloutsos & Faloutsos, 1999



Adjacency matrix eigenvalues (Watts & Strogatz model, Scale-free models)

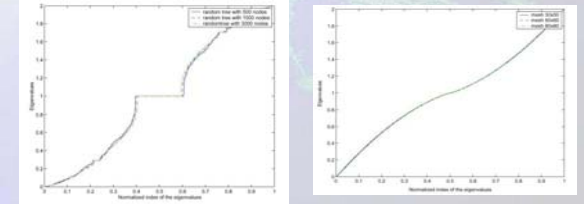
Farkas, Derenyi, Barabási & Vicsek *Phys. Rev E* 2001
Goh, Kahng & Kim 2001
Fan & Chen *IEEE Circ. Sys.* 2002

scale-free



Normalized Laplacian eigenvalues (meshes, random trees)

Vukadinovic, Huang, Erlebach 2002



How to model real networks ?

Erdős-Rényi,
Watts-Strogatz
Barabási-Albert

other models ?

Why appears a power law?

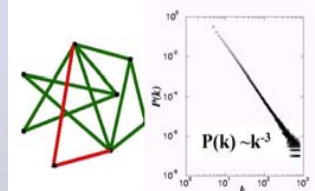
1. Networks grow continuously by addition of new nodes
2. Growth is NOT uniform: A new node will join, with high probability, an old well connected node

WWW: New documents point to "classic" references
Erdős: I would prefer to publish with a well known mathematician

"Standard" model: Barabási, Albert; *Science* 286, 509 (1999)

Preferential attachment : At each time unit a new node is added with m links which connect to existing nodes . The probability P to connect to a node i is proportional to its degree k_i ,

$$\Pi(k_i) = \frac{k_i}{\sum_j k_j}$$

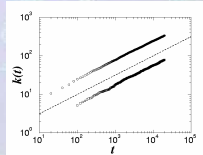


A.-L.Barabási, R. Albert, *Science* 286, 509 (1999)

Mean Field Theory

$$\frac{\partial k_i}{\partial t} \propto \Pi(k_i) = A \frac{k_i}{\sum_j k_j} = \frac{k_i}{2t}, \text{ with initial condition } k_i(t_0) = m$$

$$k_i(t) = m \sqrt{\frac{t}{t_0}}$$



$$P(k_i(t) < k) = P(t_i > \frac{m^2 t}{k^2}) = 1 - P(t_i \leq \frac{m^2 t}{k^2}) = 1 - \frac{m^2 t}{k^2(m_0 + t)}$$

$$\therefore P(k) = \frac{\partial P(k_i(t) < k)}{\partial k} = \frac{2m^2 t}{m_0 + t} \frac{1}{k^3} \sim k^{-3}$$

$\gamma = 3$

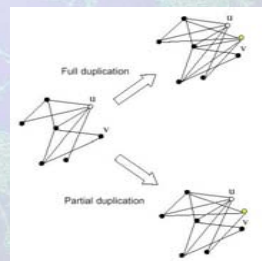
A.-L.Barabási, R. Albert and H. Jeong, *Physica A* 272, 173 (1999)

Duplication models:

Fan Chung, Lu, Dewey, Galas; (2002)

Nodes are duplicated together with all (or part) of their edges.

can produce $\gamma < 2$ as in biological networks
keep some network properties

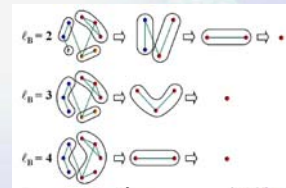
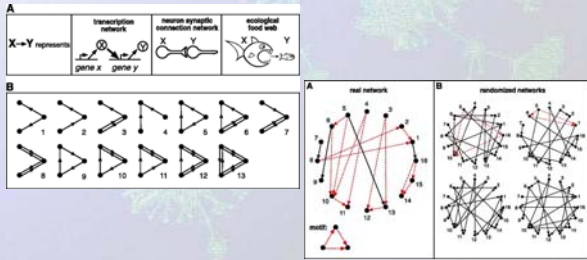


Fractal networks

Motifs, graphlets

Milo, Shen-Orr, Itzkoviz, Kashtan, Chkrovskii, Alon
Science 298, 824-827 (2002)

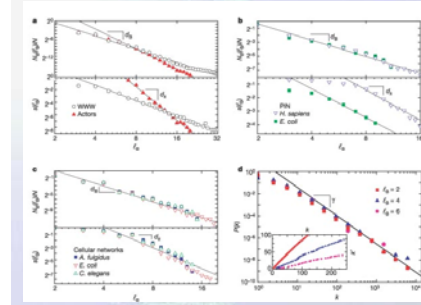
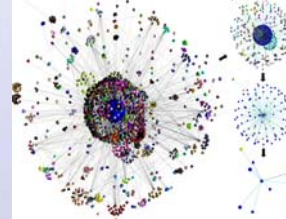
Pržulj, Corneil, Jurisica
Bioinformatics 20, 3508-3515 (2004)



Song, Havlin, Makse
Nature 433, 392-396 (2005)
Nature Physics 2, 275-281 (2006)

Self-similarity of complex networks

Origins of fractality in the growth of complex networks



$$N_B \approx \ell_B^{-d_B}$$

$$k \rightarrow k' = s(\ell_B)k$$

$$s(\ell_B) \approx \ell_B^{-d_k}$$

$$\gamma = 1 + d_B/d_k$$

WWW, protein interaction networks are fractal

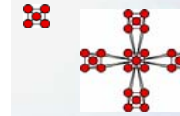
Internet (AS) is not fractal

Barabási-Albert is not fractal

Real complex networks: self-organized criticality (SOC) by some optimization process !!

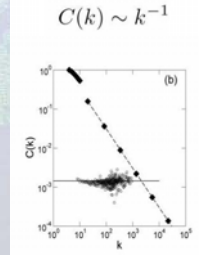
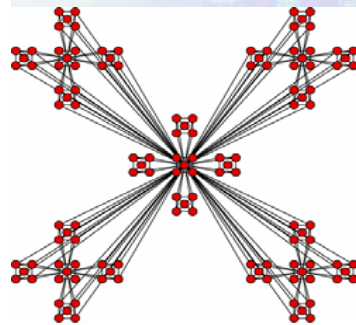
Cliques-trees, as deterministic models for real networks.

Hierarchical graphs
Recursive clique-trees
Apollonian graphs



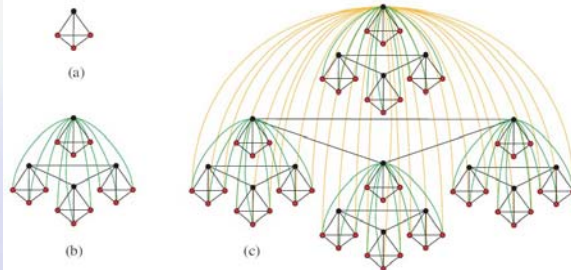
Hierarchical graphs

Ravasz, Barabasi, Hierarchical organization in complex networks
Phys. Rev. E (2003).



$$\gamma = 1 + (\ln 5 / \ln 4)$$

Real-life networks are fractal (Song, Havlin, Makse) but some fractal-looking graphs are not !



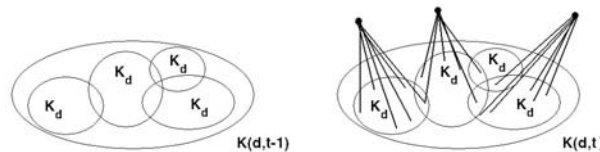
Barrière, Comellas, Dalfó (2006)

$$\gamma \approx 1 + \frac{\ln(d+1)}{\ln d}$$

Recursive clique-trees

SN Dorogovtsev, AV Gotsev, JFF Mendes., *Phys. Rev. E* (2002)

F. Comellas, Guillaume Fertin, André Raspaud, *Phys. Rev. E* (2004)



Recursive clique-trees

F. Comellas, G. Fertin, A. Raspaud, *Phys. Rev. E*

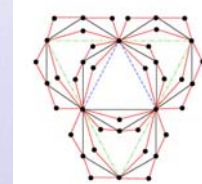
Initial graph: K_q - the complete graph with q vertices.

Operation: $t \rightarrow t+1$, obtain $K(q,t+1)$ from $K(q,t)$ by adding for every clique K_q of $K(q,t)$:

a: A new vertex u

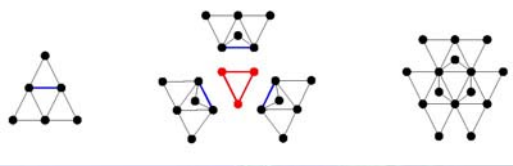
b: q edges joining u with the vertices of this clique

example $q=2$



example $q=3$

Recursive equivalent operation



•Order, size

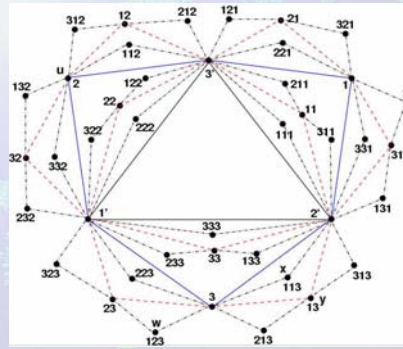
•Degree distribution $\gamma \approx 1 + \frac{\ln(d+1)}{\ln d}$ $2 < \gamma < 2.58496$

•Clustering $0.8 \leq C \leq 1$

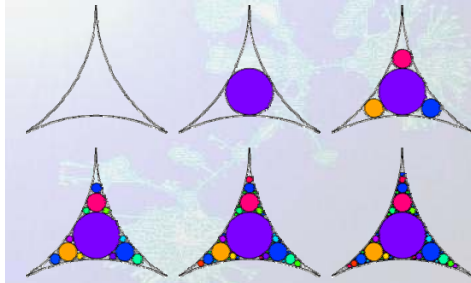
•Diameter logarithmic

Distance-labeling and routing (example q=2)

F.Comellas, G. Fertin, A. Raspaud, *Sirocco 2003*



Apollonian graphs

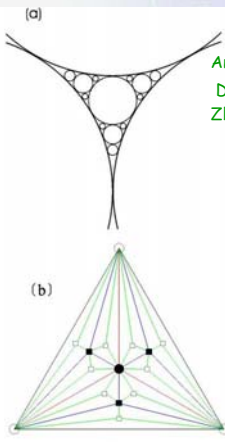


Apollonian packings

And let us not confine our cares
To simple circles, planes and spheres,
But rise to hyper flats and bends
Where kissing multiple appears,
In n-ic space the kissing pairs
Are hyperspheres, and Truth declares -
As n+2 such osculate
Each with an n+1 fold mate
The square of the sum of all the bends
Is n times the sum of their squares.

Thorold Gosset, The Kiss Precise, Nature 139 (1937) 62.

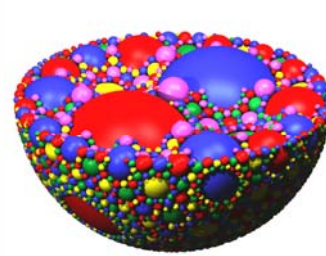
Apollonian graphs



Andrade et al. *Phys. Rev. Lett.* (2005)

Doye, Massen *Phys. Rev. E* (2005)

Zhang, Rong, Comellas, Fertin *J. Phys.A* (2006)



<http://graphics.ethz.ch/~peikert/personal/packing/images/apoll3d.png>

d=2

d=3

Step (t)	New edges	Number of K_{d+1}
0	$\frac{d(d+1)}{2}$	1
1	$d+1$	$d+1$
2	$(d+1)^2$	$(d+1)^2$
3	$(d+1)^3$	$(d+1)^3$
...
i	$(d+1)^i$	$(d+1)^i$
i+1	$(d+1)^{i+1}$	$(d+1)^{i+1}$
...

$$N_t = (d+1) + \sum_{j=0}^{t-1} (d+1)^j = \frac{(d+1)^t - 1}{d} + d + 1$$

$$|E|_t = \frac{d(d+1)}{2} + \sum_{j=1}^t (d+1)^j = \frac{d(d+1)}{2} + \frac{(d+1)^{t+1} - d - 1}{d} \quad (1)$$

TABLE II: Distribution of vertices and degrees for $A(d,t)$ at each step t .

Step(t)	Num. vert.	Degree
0	$d+1$	d
1	$d+1$	$d+1$
2	$d+1$	$d+1$
3	$d+1$	$d+1+d$
4	$d+1$	$(d+1) + (d+1)$
5	$d+1$	$d+1$
6	$d+1$	$d^2 + d + 1 + d$
7	$d+1$	$(d+1)d + (d+1) + (d+1)$
8	$d+1$	$(d+1) + (d+1)$
9	$(d+1)^2$	$d+1$
10	$(d+1)^2$	$d+1$
11	$(d+1)^2$	$d+1$
12	$(d+1)^2$	$d+1$
13	$(d+1)^2$	$d+1$
14	$(d+1)^2$	$d+1$
15	$(d+1)^2$	$d+1$
16	$(d+1)^2$	$d+1$
17	$(d+1)^2$	$d+1$
18	$(d+1)^2$	$d+1$
19	$(d+1)^2$	$d+1$
20	$(d+1)^2$	$d+1$
21	$(d+1)^2$	$d+1$
22	$(d+1)^2$	$d+1$
23	$(d+1)^2$	$d+1$
24	$(d+1)^2$	$d+1$
25	$(d+1)^2$	$d+1$
26	$(d+1)^2$	$d+1$
27	$(d+1)^2$	$d+1$
28	$(d+1)^2$	$d+1$
29	$(d+1)^2$	$d+1$
30	$(d+1)^2$	$d+1$
31	$(d+1)^2$	$d+1$
32	$(d+1)^2$	$d+1$
33	$(d+1)^2$	$d+1$
34	$(d+1)^2$	$d+1$
35	$(d+1)^2$	$d+1$
36	$(d+1)^2$	$d+1$
37	$(d+1)^2$	$d+1$
38	$(d+1)^2$	$d+1$
39	$(d+1)^2$	$d+1$
40	$(d+1)^2$	$d+1$
41	$(d+1)^2$	$d+1$
42	$(d+1)^2$	$d+1$
43	$(d+1)^2$	$d+1$
44	$(d+1)^2$	$d+1$
45	$(d+1)^2$	$d+1$
46	$(d+1)^2$	$d+1$
47	$(d+1)^2$	$d+1$
48	$(d+1)^2$	$d+1$
49	$(d+1)^2$	$d+1$
50	$(d+1)^2$	$d+1$
51	$(d+1)^2$	$d+1$
52	$(d+1)^2$	$d+1$
53	$(d+1)^2$	$d+1$
54	$(d+1)^2$	$d+1$
55	$(d+1)^2$	$d+1$
56	$(d+1)^2$	$d+1$
57	$(d+1)^2$	$d+1$
58	$(d+1)^2$	$d+1$
59	$(d+1)^2$	$d+1$
60	$(d+1)^2$	$d+1$
61	$(d+1)^2$	$d+1$
62	$(d+1)^2$	$d+1$
63	$(d+1)^2$	$d+1$
64	$(d+1)^2$	$d+1$
65	$(d+1)^2$	$d+1$
66	$(d+1)^2$	$d+1$
67	$(d+1)^2$	$d+1$
68	$(d+1)^2$	$d+1$
69	$(d+1)^2$	$d+1$
70	$(d+1)^2$	$d+1$
71	$(d+1)^2$	$d+1$
72	$(d+1)^2$	$d+1$
73	$(d+1)^2$	$d+1$
74	$(d+1)^2$	$d+1$
75	$(d+1)^2$	$d+1$
76	$(d+1)^2$	$d+1$
77	$(d+1)^2$	$d+1$
78	$(d+1)^2$	$d+1$
79	$(d+1)^2$	$d+1$
80	$(d+1)^2$	$d+1$
81	$(d+1)^2$	$d+1$
82	$(d+1)^2$	$d+1$
83	$(d+1)^2$	$d+1$
84	$(d+1)^2$	$d+1$
85	$(d+1)^2$	$d+1$
86	$(d+1)^2$	$d+1$
87	$(d+1)^2$	$d+1$
88	$(d+1)^2$	$d+1$
89	$(d+1)^2$	$d+1$
90	$(d+1)^2$	$d+1$
91	$(d+1)^2$	$d+1$
92	$(d+1)^2$	$d+1$
93	$(d+1)^2$	$d+1$
94	$(d+1)^2$	$d+1$
95	$(d+1)^2$	$d+1$
96	$(d+1)^2$	$d+1$
97	$(d+1)^2$	$d+1$
98	$(d+1)^2$	$d+1$
99	$(d+1)^2$	$d+1$
100	$(d+1)^2$	$d+1$

Discrete degree spectrum (with larger and larger jumps)

$$P_{cum}(k) \equiv \sum_{k' \geq k} N(k', t) / N_t \sim k^{1-\gamma}$$

$$\gamma \approx 1 + \frac{\ln(d+1)}{\ln d}$$

$$2 < \gamma < 2.58496$$

Random Apollonian graphs

Instead of adding simultaneously a new vertex to each clique (never used before), we add a unique vertex to a random clique.

Initially $A(d,0)$ is K_{d+2}

Step t choose clique K_{d+1} NEVER USED and add a node (and the corresponding edges)

Order increments by 1 at each step

$$N_t = t + d + 2$$

Degree distribution (self-averaging)

Given a vertex, when its degree increases by 1, the number of K_{d+1} which contains it increases by $d-1$. Thus, when the vertex attains degree k , the number of K_{d+1} is $(d+1) \cdot (k-d-1) \cdot (d-1) = (d-1) \cdot k_i - d^2 + d + 2$

$$\frac{\partial k_i}{\partial t} = \frac{(d-1)k_i - d^2 + d + 2}{dt + d + 2}$$

with initial condition $k_i(t_i) = d+1$ we obtain

$$k_i(t) = \frac{d^2 - d - 2}{d-1} + \frac{d+1}{d-1} \left(\frac{dt + d + 2}{dt_i + d + 2} \right)^{\frac{d-1}{d}}$$

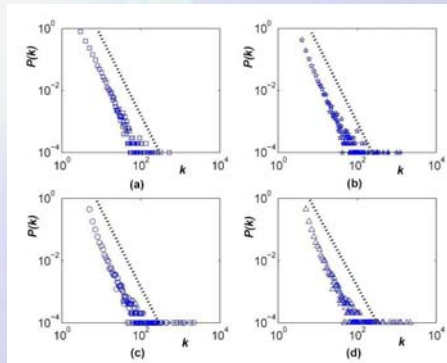
$$P(k_i(t) < k) = P\left(t_i > \frac{(dt + d + 2) \left(\frac{d+1}{d-1}\right)^{\frac{d}{d-1}} - d + 2}{d \left(k - \frac{d^2 - d - 2}{d-1}\right)^{\frac{d}{d-1}} - d + 2}\right)$$

$$P(k) = d(d+1)^{\frac{d}{d-1}} \left((d-1)k - (d^2 - d - 2) \right)^{\frac{1-2d}{d-1}}$$

If $k \gg d$ we have $P(k) \sim k^{-\gamma}$ with $\gamma(d) = \frac{2d-1}{d-1}$

$\gamma=3$ (for $d=2$, random seq.)

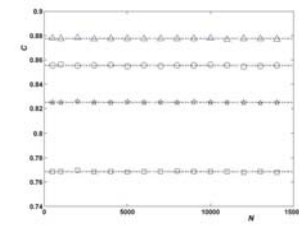
vs $\gamma=2.58496$ ($d=2$ parallel)



Degree distribution when $N=10000$, $d=2,3,4,5$

$$C(k) = \frac{\frac{d(d+1)}{2} + d(k-d-1)}{\frac{k(k-1)}{2}} = \frac{d(2k-d-1)}{k(k-1)} \quad \text{Clustering}$$

$$C = \int_{d+1}^{\infty} C(k)P(k)dk = \int_{d+1}^{\infty} \frac{d^2(2k-d-1)(d+1)^{\frac{d}{d-1}}}{k(k-1)} \left((d-1)k - (d^2-d-2) \right)^{\frac{1-2d}{d-1}} dk$$



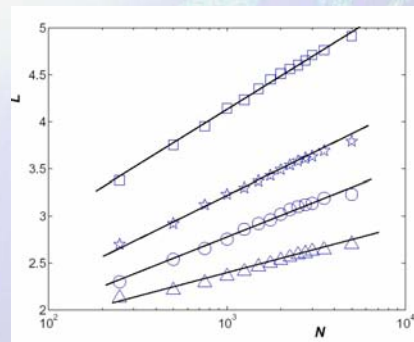
$$C = \frac{46}{3} - 36 \ln \frac{3}{2} = 0.7366$$

$$C = 18 + 36\sqrt{2} \arctan \sqrt{2} + \frac{9}{2}\pi - 18\sqrt{2}\pi = 0.8021$$

HDRAN clustering

$N=10000$, $d=2,3,4,5$

Average path length



- Zhongzhi Zhang, Lili Rong, F. Comellas, Guillaume Fertin, **High dimensional Apollonian networks** *J. Phys. A* (2006)
- Zhongzhi Zhang, Lili Rong, F. Comellas, **High dimensional random Apollonian networks** *Physica A* (2006),

Deterministic recursive clique-trees

	Adding at the same time a vertex to each d -clique with repetition	Adding at the same time a vertex to each d -clique without repetition
Case $d = 2$	<i>Pseudofractal scale-free</i> Dorogotsev, Goltsev, Mendes Phys.Rev.E 65 (2002) 066122	<i>Deterministic SW network</i> Zhang, Rong, Guo Physica A cond-mat/0503637
Case $d = 3$		<i>Apollonian network</i> Andrade, Herrmann, Andrade, Silva Phys.Rev.Lett. 94 (2005) 018702 Doye, Massen Phys. Rev. E 71 (2005) 016128.
General case $d = 2 \dots \infty$ (includes cases $d=2,3$)	<i>Recursive clique-trees</i> Comellas, Fertin, Raspaud Phys.Rev.E 69 (2004) 037104.	<i>High dimensional Apollonian network</i> Zhang, Comellas, Fertin, Rong J. Phys. A. 39 (2006) 1811 (introduced by Doye and Massen, Phys. Rev. E 71 (2005) 016128.)

Random recursive clique-trees

	Adding a single vertex to a random clique with repetition	Adding a single vertex to a random clique without repetition
Case $d = 2$		<i>Random SW network</i> Ozik, Hunt, Ott Phys.Rev.E 69 (2004) 02618
Case $d = 3$		<i>Random Apollonian network</i> Zhou, Yan, Wang Phys.Rev.E 71 (2005) 046141
General case $d = 2 \dots \infty$ (includes cases $d=2,3$)	<i>Random recursive clique-tree</i> see Appendix	<i>HD random Apollonian network</i> Zhang, Comellas, Rong Physica A. cond-mat/0502591

Deterministic vs Random

Graph family	$P(k)$ or γ -exponent	Clustering
Deterministic SW [78]	$2^{-\frac{k}{d}}$	$0.69 = \ln 2$
Random SW [77]	$\frac{3}{4} \left(\frac{2}{3}\right)^{-k}$	$0.65 (= \frac{3}{2} \ln 3 - 1)$
Apollonian [7,34]	$2.58 (= 1 + \frac{\ln 3}{\ln 2})$	0.83
Random Apollonian [82]	$\frac{3N-5}{N} \approx 3$	$0.74 (= \frac{46}{3} - 36 \ln \frac{3}{2})$
High-Dim. Apollonian [81]	$1 + \frac{\ln(d+1)}{\ln d}$ (2 to 2.58)	0.83 to 1
High-Dim. Random Apollonian [80]	$\frac{2d-1}{d-1}$ 2 to 3	0.74 to 1
Pseudo fractal scale-free [29]	$1 + \frac{\ln 3}{\ln 2} = 2.58$	$0.80 (= \frac{4}{5})$
Random pseudo fractal scale-free	$\frac{5}{2} = 2.5$	
Determ. recursive clique-trees [22]	$1 + \frac{\ln(d+1)}{\ln d}$ (2 to 2.58)	0.80 to 1
Random rec. clique-trees [see Appendix]	$\frac{2d-1}{d-1}$ (2 to 3)	0.74 to 1

Why the random approach produces a different distribution

F. Comellas, Hernan D. Rozenfeld, Daniel ben-Avraham
Synchronous and asynchronous recursive random scale-free nets
Phys. Rev. E (2005),

In many simulations choosing an edge might be biased.

It is not the same to choose edge e from $|E|$ edges than choose a node and then an adjacent node.

Present and future work in SW-SF networks

How to construct a better WWW (new topologies -Akamai) ?

How to analyse very large graphs ?

- mean field and other statistical methods
- fractal techniques
- spectral theory
- new invariants
- this workshop

How to deal with dynamical networks ?

New communication protocols

