Network formation models

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Network Science



Network models

Empirical network features:

- Power-law (heavy-tailed) degree distribution
- Small average distance (graph diameter)
- Large clustering coefficient (transitivity)

Generative models:

- Random graph model (Erdos & Renyi, 1959)
- Preferential attachment model (Barabasi & Albert, 1999)
- "Small world" model (Watts & Strogatz, 1998)

Random graph model

Graph $G\{E, V\}$, nodes n = |V|, edges m = |E| Erdos and Renyi, 1959.

- Random graph models
 - $G_{n,m}$, a randomly selected graph from the set of $C_N^m graphs$, $N = \frac{n(n-1)}{2}$, with n nodes and m edges
 - $G_{n,p}$, each pair out of $N = \frac{n(n-1)}{2}$ pairs of nodes is connected with probability p, m random number

$$\langle m \rangle = p \frac{n(n-1)}{2}$$

$$\langle k \rangle = \frac{1}{n} \sum_{i} k_{i} = \frac{2\langle m \rangle}{n} = p (n-1) \approx pn$$

$$\rho = \frac{\langle m \rangle}{n(n-1)/2} = p$$

Random graph model

• Probability that *i*-th node has a degree $k_i = k$

$$P(k_i = k) = P(k) = C_{n-1}^k p^k (1-p)^{n-1-k}$$

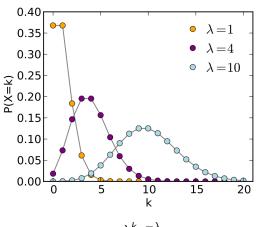
(Bernoulli distribution) p^k - probability that connects to k nodes (has k-edges) $(1-p)^{n-k-1}$ - probability that does not connect to any other node C_{n-1}^k - number of ways to select k nodes out of all to connect to

• Limiting case of Bernoulli distribution, when $n \to \infty$ at fixed $\langle k \rangle = pn = \lambda$

$$P(k) = \frac{\langle k \rangle^k e^{-\langle k \rangle}}{k!} = \frac{\lambda^k e^{-\lambda}}{k!}$$

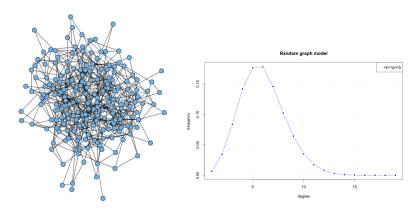
(Poisson distribution)

Poisson Distribution



$$P(k_i = k) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad \lambda = pn$$

Random graph

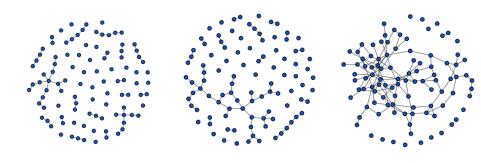


$$\langle k \rangle = pn = 5$$

Random graph model

Consider $G_{n,p}$ as a function of p

- p=0, empty graph $\langle k \rangle = 0$
- p=1, complete (full) graph $\langle k \rangle = n-1$
- ullet n_G -largest connected component, $s=rac{n_G}{n}$



Phase transition

Let u – fraction of nodes that do not belong to GCC. The probability that a node does not belong to GCC

$$u = \frac{n - n_G}{n} = P(k = 0) + P(k = 1) \cdot u + P(k = 2) \cdot u^2 + P(k = 3) \cdot u^3 \dots =$$

$$= \sum_{k=0}^{\infty} P(k) u^k = \sum_{k=0}^{\infty} \frac{\lambda^k e^{-\lambda}}{k!} u^k = e^{-\lambda} e^{\lambda u} = e^{\lambda(u-1)} e^{\lambda(u-1)}$$

Let s -fraction of nodes belonging to GCC (size of GCC)

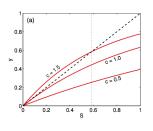
$$s = 1 - u$$

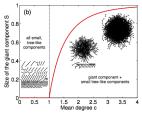
$$1 - s = e^{-\lambda s}$$

when $\lambda \to \infty$, $s \to 1$ when $\lambda \to 0$, $s \to 0$ $\lambda = pn = \langle k \rangle$

Phase transition

$$s = 1 - e^{-\lambda s}$$





non-zero solution exists when (at s = 0):

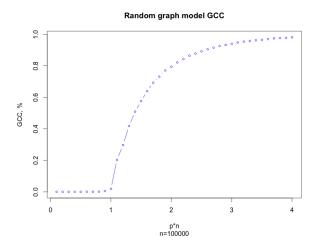
$$\lambda e^{-\lambda s} > 1$$

critical value:

$$\lambda_c = 1$$

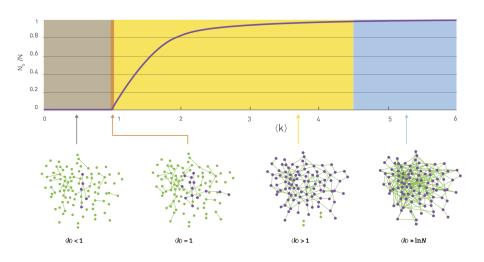
$$\lambda_c = \langle k \rangle = p_c n = 1, \quad p_c = \frac{1}{n}$$

Numerical simulations



$$\langle k \rangle = pn$$

evolution of random network



from A-L. Barabasi, 2016

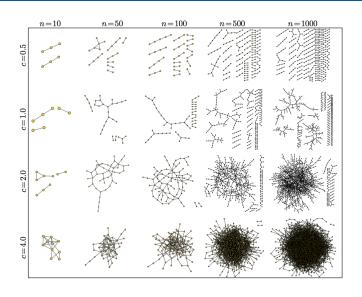
Phase transition

Graph G(n,p), for $n \to \infty$, critical value $p_c = 1/n$

- Subcritical regime: $p < p_c$, $\langle k \rangle < 1$ there is no components with more than $O(\ln n)$ nodes, largest component is a tree
- Critical point: $p=p_c$, $\langle k \rangle=1$ the largest component has $O(n^{2/3})$ nodes
- Supercritical regime: $p>p_c$, $\langle k \rangle>1$ gigantic component has all $O((p-p_c)n)$ nodes
- Connected regime: $p >> \ln n/n$, $\langle k \rangle > \ln n$ gigantic component has all O(n) nodes

Critical value: $\langle k \rangle = p_c n = 1$ - on average one neighbor for a node

Numerical simulation

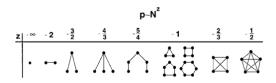


Clauset, 2014

Threshold probabilities

Graph G(n, p)

Threshold probabilities when different subgraphs of k-nodes and l-edges appear in a random graph $p_s \sim n^{-k/l}$



When $p > p_s$:

- $p_s \sim n^{-k/(k-1)}$, having a tree with k nodes
- $p_s \sim n^{-1}$, having a cycle with k nodes
- $p_s \sim n^{-2/(k-1)}$, complete subgraph with k nodes

Barabasi, 2002

Clustering coefficient

 Clustering coefficient (probability that two neighbors link to each other):

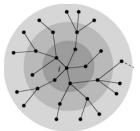
$$C_i(k) = \frac{\text{\#of links between NN}}{\text{\#max number of links NN}} = \frac{pk(k-1)/2}{k(k-1)/2} = p$$

$$C = p = \frac{\langle k \rangle}{n}$$

• when $n \to \infty$, $C \to 0$

Graph diameter

• G(n, p) is locally tree-like (GCC) (no loops; low clustering coefficient)



• on average, the number of nodes d steps away from a node

$$n = 1 + \langle k \rangle + \langle k \rangle^2 + ... \langle k \rangle^D = \frac{\langle k \rangle^{D+1} - 1}{\langle k \rangle - 1} \approx \langle k \rangle^D$$

• in GCC, around p_c , $\langle k \rangle^D \sim n$,

$$D \sim \frac{\ln n}{\ln \langle k \rangle}$$

Random graph model

• Node degree distribution function - Poisson:

$$P(k) = \frac{\lambda^k e^{-\lambda}}{k!}, \quad \lambda = pn = \langle k \rangle$$

• Average path length:

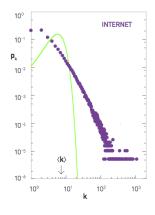
$$\langle L \rangle = \frac{\ln n}{\ln \langle k \rangle}$$

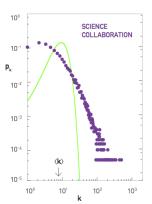
Clustering coefficient:

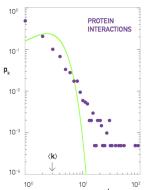
$$C=\frac{\langle k\rangle}{n}$$

Real networks

Degree distribution in real networks







Configuration model

- Random graph with n nodes with a given degree sequence: $D = \{k_1, k_2, k_3...k_n\}$ and $m = 1/2 \sum_i k_i$ edges.
- Construct by randomly matching two stubs and connecting them by an edge.



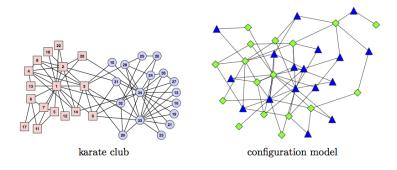
- Can contain self loops and multiple edges
- Probability that two nodes i and j are connected

$$p_{ij} = \frac{k_i k_j}{2m-1}$$

• Will be a simple graph for special "graphical degree sequence"

Configuration model

Can be used as a "null model" for comparative network analysis



Clauset, 2014

Motivation of Evolutional Random Graph Model

Most of the networks we study are evolving over time, they expand by adding new nodes:

- Citation networks
- Collaboration networks
- Web
- Social networks

Preferential attachment model

Barabasi and Albert, 1999

Dynamic growth model

Start at t = 0 with n_0 nodes and some edges $m_0 \ge n_0$

Growth

At each time step add a new node with m edges ($m \le n_0$), connecting to m nodes already in network $k_i(i) = m$

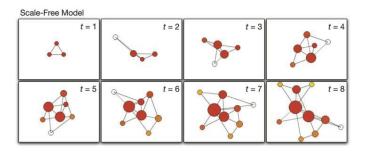
Preferential attachment

The probability of linking to existing node i is proportional to the node degree k_i

$$\Pi(k_i) = \frac{k_i}{\sum_i k_i}$$

after t timesteps: $t + n_0$ nodes, $mt + m_0$ edges

Preferential attachment model



Barabasi, 1999

Preferential attachment

Continues approximation: continues time, real variable node degree $\langle k_i(t) \rangle$ - expected value over multiple realizations Time-dependent degree of a single node:

$$k_i(t + \delta t) = k_i(t) + m\Pi(k_i)\delta t$$

$$\frac{dk_i(t)}{dt} = m\Pi(k_i) = m\frac{k_i}{\sum_i k_i} = \frac{mk_i}{2mt}$$
$$\frac{dk_i(t)}{dt} = \frac{k_i(t)}{2t}$$

initial conditions: $k_i(t = i) = m$

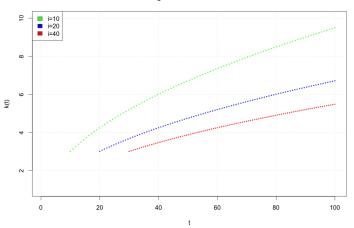
$$\int_{m}^{k_{i}(t)} \frac{dk_{i}}{k_{i}} = \int_{i}^{t} \frac{dt}{2t}$$

Solution:

$$k_i(t) = m\left(\frac{t}{i}\right)^{1/2}$$

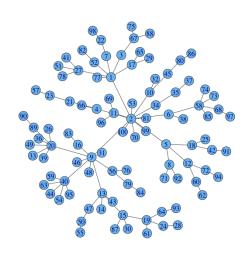
Preferential attachement

Node degree k as function of time t



$$k_i(t) = m\left(\frac{t}{i}\right)^{1/2}$$

Preferential attachement



Preferential attachment

Time evolution of a node degree

$$k_i(t) = m\left(\frac{t}{i}\right)^{1/2}$$

Nodes with $k_i(t) \leq k$:

$$m\left(\frac{t}{i}\right)^{1/2} \le k$$
$$i \ge \frac{m^2}{k^2}t$$

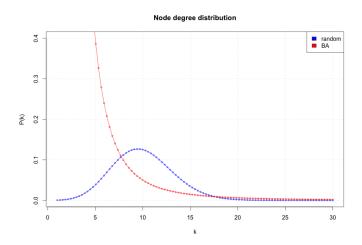
Probability of randomly selected node to have $k' \leq k$ (fraction of nodes with $k' \leq k$)

$$F(k) = P(k' \le k) = \frac{n_0 + t - m^2 t/k^2}{n_0 + t} \approx 1 - \frac{m^2}{k^2}$$

Distribution function:

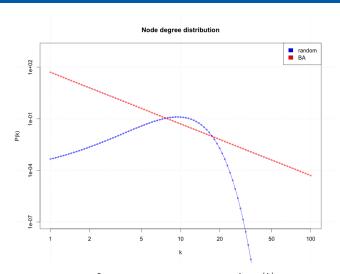
$$P(k) = \frac{d}{dk}F(k) = \frac{2m^2}{k^3}$$

Preferential attachment vs random graph



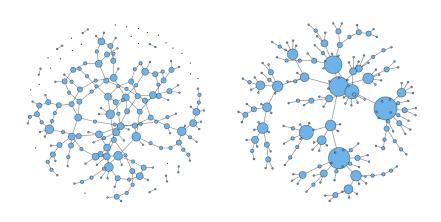
$$BA: P(k) = \frac{2m^2}{k^3}, \qquad ER: P(k) = \frac{\langle k \rangle^k e^{-\langle k \rangle}}{k!}, \quad \langle k \rangle = pn$$

Preferential attachment vs random graph

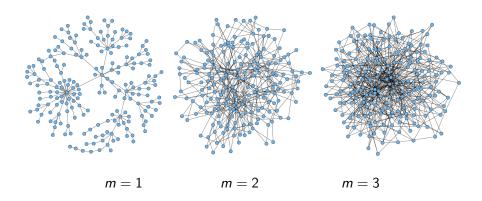


$$BA: P(k) = \frac{2m^2}{k^3}, \qquad ER: P(k) = \frac{\langle k \rangle^k e^{-\langle k \rangle}}{k!}, \quad \langle k \rangle = pn$$

Preferential attachment vs random graph



Preferential attachment model



Growing random graph

Growth

At each time step add a new node with m edges ($m \le n_0$), connecting to m nodes already in network $k_i(i) = m$

Preferential attachment Uniformly at randomThe probability of linking to existing node *i* is

$$\Pi(k_i) = \frac{1}{n_0 + t - 1}$$

Node degree growth:

$$k_i(t) = m\left(1 + \log\left(\frac{t}{i}\right)\right)$$

Node degree distribution function:

$$P(k) = \frac{e}{m} \exp\left(-\frac{k}{m}\right)$$

Preferential attachment

Power law distribution function:

$$P(k) = \frac{2m^2}{k^3}$$

• Average path length (analytical result) :

$$\langle L \rangle \sim \log(N)/\log(\log(N))$$

Clustering coefficient (numerical result):

$$C \sim N^{-0.75}$$

Many more models

Some other models that produce scale-free networks:

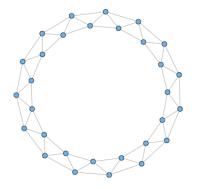
- Non-linear preferential attachment
- Link selection model
- Copying model
- Cost-optimization model
- ...

Historical note

- Polya urn model, George Polya, 1923
- Yule process, Udny Yule, 1925
- Distribution of wealth, Herbert Simon, 1955
- Evolution of citation networks, cumulative advantage, Derek de Solla Price, 1976
- Preferential attachment network model, Barabasi and Albert, 1999

Small world

Motivation: keep high clustering, get small diameter



Clustering coefficient C = 1/2Graph diameter d = 8

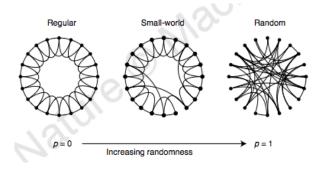
Small world

Watts and Strogatz, 1998

Single parameter model, interpolation between regular lattice and random graph

- start with regular lattice with n nodes, k edges per vertex (node degree), k << n
- randomly connect with other nodes with probability p, forms pnk/2 "long distance" connections from total of nk/2 edges
- p = 0 regular lattice, p = 1 random graph

Small world



Watts, 1998

Small world model

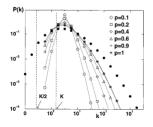
Node degree distribution:

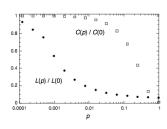
Poisson like

- Ave. path length $\langle L(p) \rangle$: $p \to 0$, ring lattice, $\langle L(0) \rangle = 2n/k$ $p \to 1$, random graph, $\langle L(1) \rangle = \log(n)/\log(k)$
- Clustering coefficient C(p):

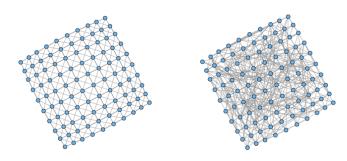
$$p \rightarrow 0$$
, ring lattice, $C(0) = 3/4 = const$

 $p \rightarrow 1$, random graph, C(1) = k/n





Small world model



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20% rewiring: ave. path length = 3.58 \rightarrow ave. path length = 2.32 clust. coeff = 0.49 \rightarrow clust. coeff = 0.19
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Model comparison

	Random	BA model	WS model	Empirical networks
P(k)	$\frac{\lambda^k e^{-\lambda}}{k!}$	k^{-3}	poisson like	power law
C	$\langle k \rangle / N$	$N^{-0.75}$	const	large
$\langle L \rangle$	$\frac{\log(N)}{\log(\langle k \rangle)}$	$\frac{\log(N)}{\log\log(N)}$	log(N)	small

References

- On random graphs I, P. Erdos and A. Renyi, Publicationes Mathematicae 6, 290–297 (1959).
- On the evolution of random graphs, P. Erdos and A. Renyi,
 Publication of the Mathematical Institute of the Hungarian Academy of Sciences, 17-61 (1960)
- Emergence of Scaling in Random Networks, A.L. Barabasi and R. Albert, Science 286, 509-512, 1999
- Collective dynamics of small-world networks. Duncan J. Watts and Steven H. Strogatz. Nature 393 (6684): 440-442, 1998