



Aalto University
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CS-E5745 Mathematical Methods for Network Science

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Generating functions and their use in networks

- ▶ Learning goals this week:
 - ▶ Recap of probability generating functions from last week
 - ▶ Learn how to use PGFs to solve component size distributions in networks
- ▶ We will be following the Section 13 in Newman: *Networks, An Introduction*

PGFs from last week

- ▶ Definition:

$$g(z) = p(0) + p(1)z + p(2)z^2 \dots = \sum_{k=0}^{\infty} p(k)z^k \quad (1)$$

- ▶ $p(k)$ can be extracted through derivation:

$$p(k) = \left[\frac{1}{k!} \frac{d^k}{dz^k} g(z) \right]_{z=0} \quad (2)$$

- ▶ Moments can also be calculated through derivation:

$$\langle X^m \rangle = \left[z \frac{d}{dz} \dots z \frac{d}{dz} g(z) \right]_{z=1} = \left[\left(z \frac{d}{dz} \right)^m g(z) \right]_{z=1} \quad (3)$$

PGFs from last week

- Sums of independent RVs

$$g_{X_1+X_2}(z) = g_{X_1}(z) * g_{X_2}(z) \quad (4)$$

$$g_{\sum_{i=1}^N X_i}(z) = [g_{X_i}(z)]^N \quad (5)$$

$$g_{X_1+c}(z) = g_{X_1}(z) * z^c \quad (6)$$

- If N is also a RV in $S = \sum_{i=1}^N X_i$:

$$g_S(z) = g_N(g_{X_i}(z)) \quad (7)$$

Notation for networks (from Newman)

- ▶ For the degree distribution $p(k)$:

$$g_0(z) = \sum_{k=0}^{\infty} p(k)z^k$$

- ▶ For the excess degree distribution $q(k)$:

$$g_1(z) = \sum_{k=0}^{\infty} q(k)z^k$$

- ▶ These two are related:

$$g_1(z) = \frac{1}{\langle k \rangle} \frac{d}{dz} g_0(z)$$

Solving the Galton-Watson process for networks

- ▶ Last week we derived the equations for Galton-Watson processes

$$\begin{aligned}g_{K_1}(z) &= g_0(z) \\g_{K_d}(z) &= g_{K_{d-1}}(g_1(z))\end{aligned}$$

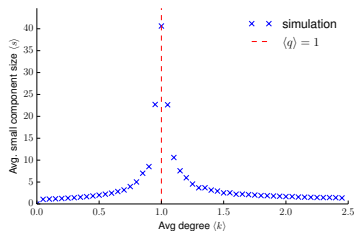
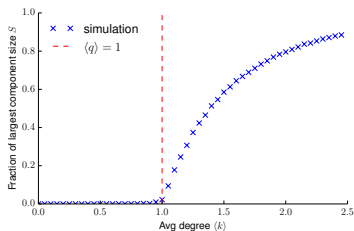
- ▶ Solution for the expected value:

$$\langle K_d \rangle = \langle q \rangle^{d-1} \langle k \rangle = \left(\frac{\langle k^2 \rangle - \langle k \rangle}{\langle k \rangle} \right)^{d-1} \langle k \rangle$$

- ▶ A criterion for the percolation threshold!

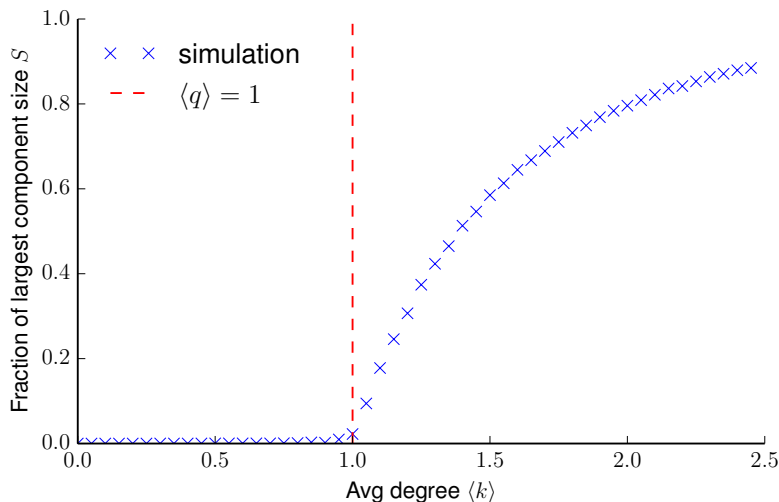
Solving the Galton-Watson process for networks

- ▶ We have the percolation threshold, but we are still missing
 - ▶ Shape of the relative giant size curve ($\sum_d K_d$ in the giant)
 - ▶ Size distributions of small components ($\sum_d K_d$ when not in the giant)
- ▶ Note: GW process approximation for networks only works when d is small



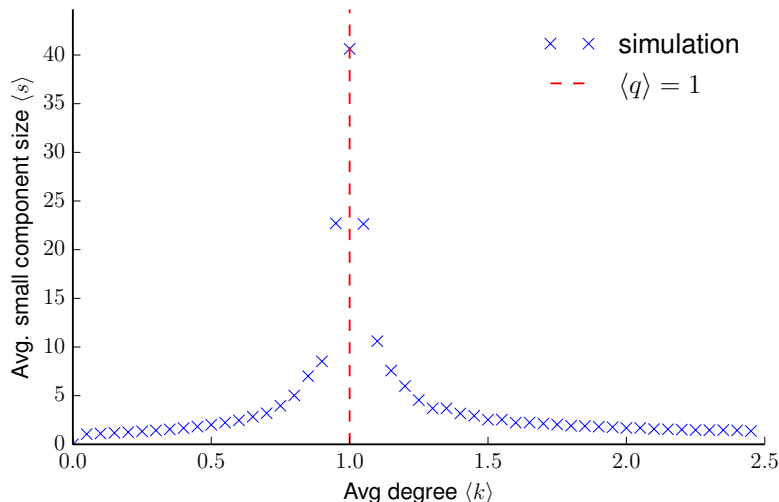
Giant component size in ER networks

- ▶ Largest component size in ER networks with $N = 10^5$



Expected small component size in ER networks

- ▶ Expected size of components other than the largest component in ER networks with $N = 10^5$



Uniqueness of the giant component

- ▶ Assume that there are two large components in ER networks with $S_1, S_2 > 0$ proportion of all nodes in them
- ▶ Number of nodes: $S_1 N$ and $S_2 N$
- ▶ Possible edges between the components:
 $S_1 N S_2 N = S_1 S_2 N^2$
- ▶ Probability that no two pairs are connected:
 $q = (1 - p)^{S_1 S_2 N^2} = (1 - \frac{\langle k \rangle}{N-1})^{S_1 S_2 N^2}$
- ▶ In “thermodynamic limit” ($N \rightarrow \infty$, s.t. $\langle k \rangle$ constant):

$$q = q_0 e^{-\langle k \rangle S_1 S_2 N} \rightarrow 0$$

Solving for giant component size

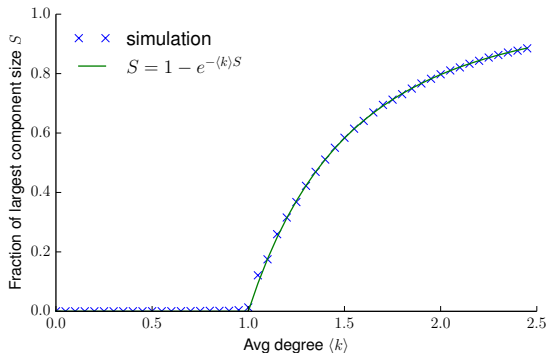
- ▶ Idea: Instead of the GW process, we write down “self-consistency equations”
 - ▶ Write down the probability random node not being in giant u as a function of u
 - ▶ The fixed point where $u = f(u)$ gives us u
 - ▶ The probability of random node in giant $S = 1 - u$

Solving for giant component size for ER networks

- ▶ Definition: u is the probability that (uniformly) randomly selected node doesn't belong to the giant component
 - ▶ Note that for Poisson degree distributions: $p(k) = q(k)$
- ▶ If node i doesn't belong to the giant then for every other node j :
 - ▶ There is no connection between i and j (probability $1 - p$),
or
 - ▶ There is connection, but j doesn't belong to the giant (probability pu)
- ▶ In total the following must hold $u = [(1 - p) + pu]^{N-1}$
 - ▶ $u = e^{-\langle k \rangle(1-u)}$, when $N \rightarrow \infty$

Solving for giant component size for ER networks

- Shape of the relative giant size curve for ER network (with $N = 10^5$ nodes)



Solving for giant component size for configuration model (1/3)

- ▶ Definition: u is the probability that following an edge (and removing it) doesn't lead to the giant component
- ▶ If node i doesn't belong to the giant then none of the neighbors j belong to the giant
 - ▶ The number of new neighbors the node i has k is distributed as the excess degree $q(k)$
 - ▶ Probability that none of the k neighbors is in the giant is u^k
- ▶ In total the following must hold $u = \sum_{k=0}^{\infty} q(k)u^k$

Solving for giant component size for configuration model (2/3)

- ▶ In total the following must hold $u = \sum_{k=0}^{\infty} q(k)u^k$
 - ▶ Using the definition of the PGF [Eq. (1)]: $u = g_1(u)$
 - ▶ Probability of uniformly randomly selected node not being in the giant is $\sum_{k=0}^{\infty} p(k)u^k = g_0(u)$
- ▶ Returns the previous result for ER networks: $u = e^{-\langle k \rangle(1-u)}$

Solving for giant component size for configuration model (3/3)

- ▶ We can also solve the component size distributions of non-giant components
- ▶ Write equations for component size distributions using component size distributions
 - ▶ Leads to nice equations when done using PGF's

Solving for component size distributions

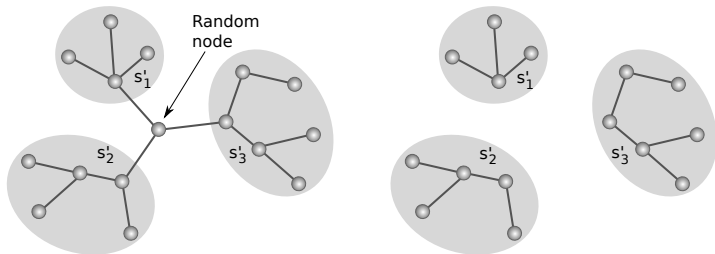
- ▶ Separate giant component size and small component size distribution
- ▶ S : Probability that uniformly randomly selected node belongs to the giant component
- ▶ RV s : size of non-giant component where uniformly randomly selected node belongs to
- ▶ Notation from Newman:
 - ▶ $P(s = k) = \pi_k$
 - ▶ $h_0(z) = g_s(z)$
- ▶ Sum of the above probabilities is one: $h_0(1) + S = 1$

Solving for component size distributions

- ▶ RV s' : Follow a link and remove it. s' is the size of the non-giant component after that link is removed (“excess small component size”)
- ▶ Notation from Newman:
 - ▶ $P(s' = k) = \rho_k$
 - ▶ $h_1(z) = g_{s'}(z)$

Solving for component size distributions

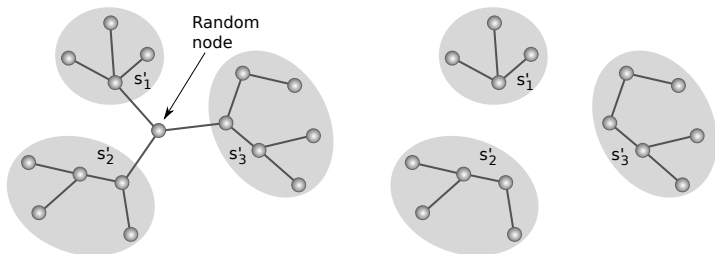
► $s = 1 + s'_1 + s'_2 + s'_3$



Solving for component size distributions

- ▶ RVs s , s' , and the first neighborhood size k_1 are related

$$s = 1 + \sum_{k=0}^{K_1} s'$$

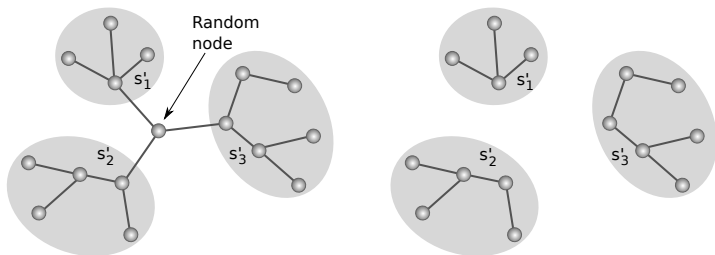


Solving for component size distributions

- ▶ Notation from Newman: $h_0(z) = g_s(z)$ and $h_1(z) = g_{s'}(z)$
- ▶ Using GF properties

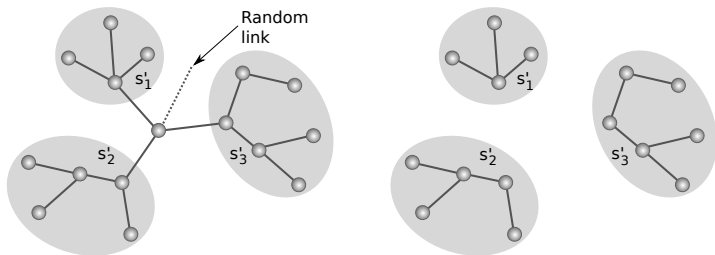
$$s = 1 + \sum_{k=0}^{K_1} s' \iff$$

$$h_0(z) = zg_0(h_1(z))$$



Solving for component size distributions

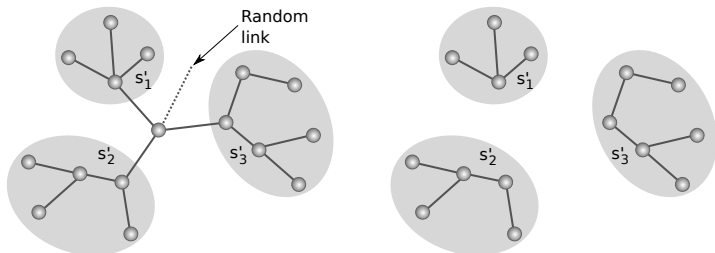
► $s' = 1 + s'_1 + s'_2 + s'_3$



Solving for component size distributions

- ▶ s' (h_1) and the excess degree $K_{1,i}$ (g_1) are related

$$s' = 1 + \sum_{k=0}^{K_{1,i}} s'$$

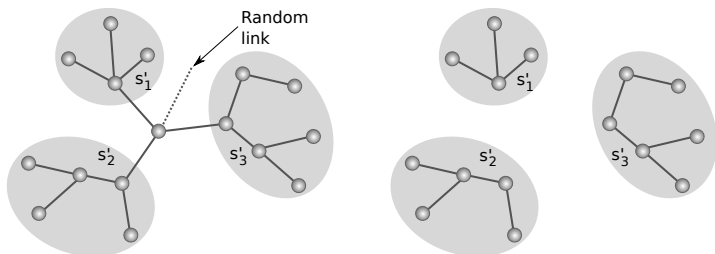


Solving for component size distributions

- ▶ Using GF properties

$$s' = 1 + \sum_{k=0}^{K_{1,i}} s' \iff$$

$$h_1(z) = zg_1(h_1(z))$$



Solving for component size distributions

- ▶ In total we have

$$h_0(z) = zg_0(h_1(z)) \quad (8)$$

$$h_1(z) = zg_1(h_1(z)) \quad (9)$$

- ▶ Solving $h_0(z)$ possible but not easy
- ▶ Solving $h_0(1)$ easier
 - ▶ Remember: $h_0(1) + S = 1$

Solving for component size distributions

- ▶ Using $h_0(1) + S = 1$ and Eq. (8):

$$S = 1 - h_0(1) = 1 - g_0(h_1(1))$$

- ▶ The value $h_0(1)$ can be solved from Eq. (9):

$$h_1(1) = g_1(h_1(1))$$

- ▶ Using Newmans notation: $u = h_1(1)$ (=probability that following a link does not lead to the giant component):

$$S = 1 - g_0(u) \tag{10}$$

$$u = g_1(u) \tag{11}$$

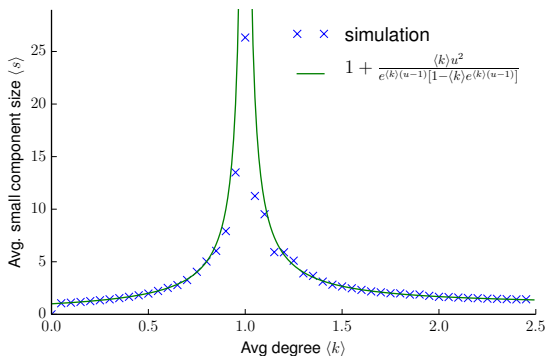
Solving for component size distributions

- ▶ With some (home)work one can use previous equations to solve for the expected small component size

$$\langle s \rangle = 1 + \frac{g'_0(1)u^2}{g_0(u)[1 - g'_1(u)]} \quad (12)$$

Solving for giant component size for ER networks

- Shape of the expected size of the small components in ER network (with $N = 10^5$ nodes)



Solving for component size distributions

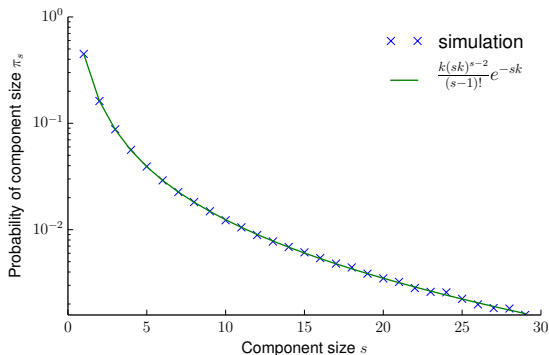
- ▶ With even more work one can find a formula for the whole small component size distribution!
 - ▶ See Newman pp. 468–469

$$\pi_s = \frac{\langle k \rangle}{(s-1)!} \left[\frac{d^{s-2}}{dz^{s-2}} [g_1(z)]^s \right]_{z=0} \quad (13)$$

- ▶ Note that the formula works only for $s > 1$ and $\pi_1 = \rho_0$

Solving for component size distribution for ER networks

- ▶ Component sizes in ER network (with $N = 10^7$ nodes and $\langle k \rangle = 0.8$)



Summary on component sizes

- ▶ We can now solve:
 - ▶ Threshold for connectivity (percolation threshold)
 - ▶ The (expected) number of nodes d steps away in BFS process
 - ▶ The size of the giant component
 - ▶ Component size distribution
- ▶ ... with some limitations:
 - ▶ Equations are derived for large configuration models (but might give good approximations for real networks)
 - ▶ We can write the equations but they might not have closed form solutions
 - ▶ Some equations do not work for the critical point (one can also solve for this separately)

Extensions

- ▶ Similar ideas (PGF's, self-consistency equations ...) can be used for many extensions
 - ▶ Networks with degree correlations
 - ▶ Networks with triangles
 - ▶ Mutual connectivity
 - ▶ ...
- ▶ Some of these extensions are done in the possible projects

Percolation problems

- ▶ In percolation problems one removes nodes or edges and calculates the connected components
- ▶ Equivalent to finding component size distributions with altered degree distributions! (if done for configuration model or ER networks)
 - ▶ If K and K' are RV's of the original and the modified degree distribution, then percolation methods are given by
$$P(K' = k | K = m)$$
 - ▶ $P(K' = k) = \sum_{m=0}^{\infty} P(K' = k | K = m) * P(K = m)$
 - ▶ $g_{K'}(z) = \sum_{k=0}^{\infty} \sum_{m=0}^{\infty} P(K' = k | K = m) * P(K = m) z^k$
 - ▶ Some projects might include percolation done in this way