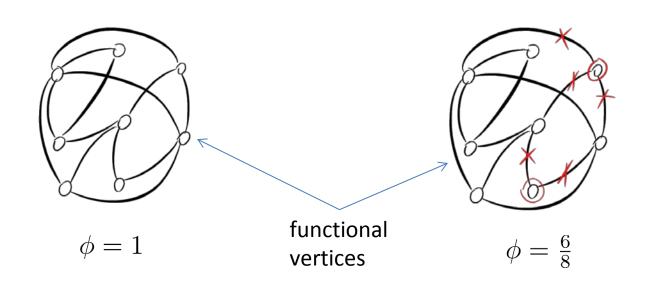
Percolation and Network Resilience

Percolation is a process of removing some fraction of network's nodes with adjacent edges. (more precisely site/link/cluster percolation)

- models real-life phenomena such as router failure, immunization of people, and disasters
- the process is parameterized by **occupation probability** ϕ
- **Percolation transition**: when ϕ is large there is a giant component but as ϕ is decreased then gc breaks into many small components or clusters (similar to phase transition in Poisson random graphs with gc \rightarrow sc)

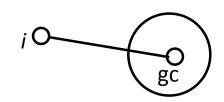


Percolation and Configuration Model

Consider a configuration model with

- degree distribution p_k
- occupation probability ϕ

consider node i which



- can belong to gc, i.e., connected to it through some $j \in N(i)$
- is not in gc (i.e., not connected to it via any of N(i)) define the avg probability of it u^k , where k = deg(i) and u is the same prob for one particular neighbor
- avg probability of not being in gc

generating function for the degree distribution

G

$$\sum_{k} p_{k} u^{k} = g_{0}(u), \text{ where } g_{0}(z) = \sum_{k}^{\infty} p_{k} z^{k} \text{ or } \Pr[i \in gc] = 1 - g_{0}(u)$$

• total fraction of nodes in gc when percolation is running

$$S = \phi(1 - g_0(u))$$

Let us calculate u, the probability that i is not connected to gc via a particular neighbor. Two cases:

- i is connected to j which is removed with prob $1-\phi$
- \bullet or j is not removed with prob ϕ but it is not in gc

$$\Pr[i \not\in gc \text{ via } j] = 1 - \phi + \phi u^k$$

j is on and its *k* neighbors are not in gc

Node j is reached by following an edge, so average probability

$$u = \sum_{k=0}^{\infty} q_k (1 - \phi + \phi u^k) = 1 - \phi + \phi + \sum_{k=0}^{\infty} q_k u^k = 1 - \phi + \phi g_1(u)$$

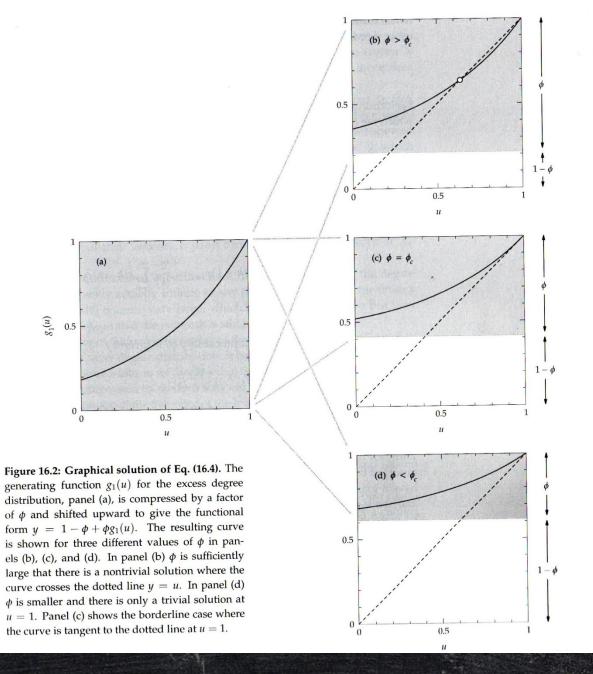
(see handout for graphical solution of the equation)

[Cohen, Erez, Ben-Avraham, Halvin]:
$$\phi_c = \frac{1}{g_1'(1)} = \frac{\langle k \rangle}{\langle k^2 \rangle - \langle k \rangle}$$

minimum fraction of nodes that must be occupied in configuration model for a gc to exist

Reminder: excess degree is the number of edges attached to a vertex other than the edge we arrived along.

$$q_k = \frac{(k+1) p_{k+1}}{\langle k \rangle}, \quad \sum_k q_k = 1, \quad g_1(z) = \sum_{k=0}^{\infty} q_k z^k$$



[Cohen, Erez, Ben-Avraham, Halvin]:
$$\phi_c = \frac{1}{g_1'(1)} = \frac{\langle k \rangle}{\langle k^2 \rangle - \langle k \rangle}$$

- In configuration model fixing low ϕ_c leads to gc, for example $\langle k^2 \rangle > \langle k \rangle$
- ullet Example: given Poisson degree distribution with c mean degree

$$p_k = e^{-c} \frac{c^k}{k!} \implies \langle k \rangle = c, \ \langle k^2 \rangle = c(c+1) \implies \phi_c = \frac{1}{c}$$

i.e., c = 4 means that $\frac{3}{4}$ vertices will fail before gc disappears.

[Cohen, Erez, Ben-Avraham, Halvin]: $\phi_c = \frac{1}{g_1'(1)} = \frac{\langle k \rangle}{\langle k^2 \rangle - \langle k \rangle}$

• Example: Exponential degree distribution $p_k = (1 - e^{-\lambda})e^{-\lambda k}, \lambda > 0$

$$g_0\left(z\right) = \frac{e^{\lambda}-1}{e^{\lambda}-z}, \ g_1\left(z\right) = \left(\frac{e^{\lambda}-1}{e^{\lambda}-z}\right)^2 \ \Rightarrow \\ u=1 \text{ is always a solution and } \\ u\left(e^{\lambda}-u\right)^2 - \left(1-\phi\right)\left(e^{\lambda}-u\right)^2 - \phi\left(e^{\lambda}-1\right)^2 = 0 \ \Rightarrow \\ u=1 \text{ is always a solution and } \\ \left(u-1\right) \text{ is always factor} \\ u=e^{\lambda}-\frac{1}{2}\phi-\sqrt{\frac{1}{4}\phi^2+\phi(e^{\lambda}-1)} \ \Rightarrow \\ S=\frac{3}{2}\phi-\sqrt{\frac{1}{4}\phi^2+\phi(e^{\lambda}-1)} \ \Rightarrow \\ \phi_c=\frac{1}{2}(e^{\lambda}-1)$$
 shold

size of gc

percolation threshold

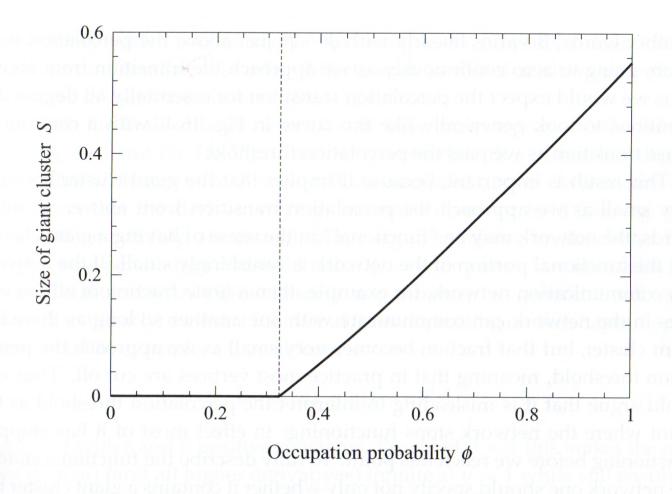


Figure 16.4: Size of the giant cluster for site percolation in the configuration model. The curve indicates the size of the giant cluster for a configuration model with an exponential degree distribution of the form (16.12) with $\lambda = \frac{1}{2}$, as given by Eq. (16.18). The dotted line indicates the position of the percolation transition, Eq. (16.20).

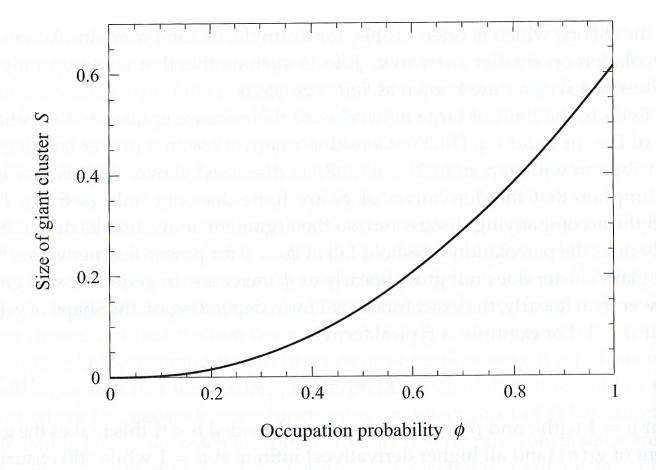


Figure 16.5: Size of the giant cluster for a network with power-law degree distribution. The size of the giant cluster for a scale-free configuration model network with exponent $\alpha = 2.5$, a typical value for real-world networks. Note the non-linear form of the curve near $\phi = 0$, which means that S, while technically non-zero, becomes very small in this regime. Contrast this figure with Fig. 16.4 for the giant cluster size in a network with an exponential degree distribution.

Geometric graph: percolation example

Non-uniform Removal of Vertices

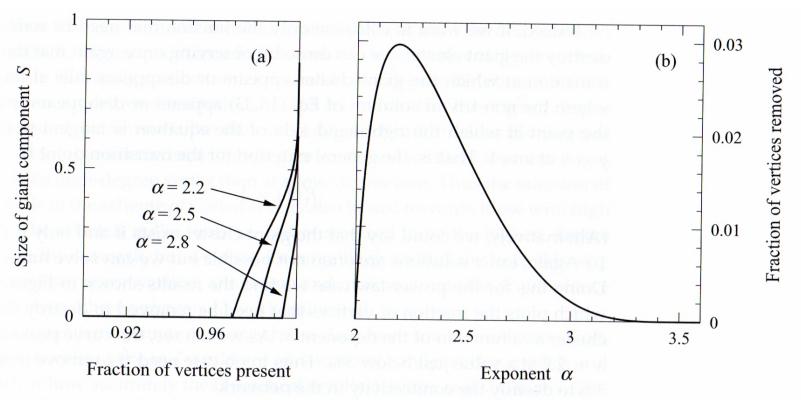


Figure 16.7: Removal of the highest-degree vertices in a scale-free network. (a) The size of the giant cluster in a configuration model network with a power-law degree distribution as vertices are removed in order of their degree, starting with the highest-degree vertices. Only a small fraction of the vertices need be removed to destroy the giant cluster completely. (b) The fraction of vertices that must be removed to destroy the giant cluster as a function of the exponent α of the power-law distribution. For no value of α does the fraction required exceed 3%.

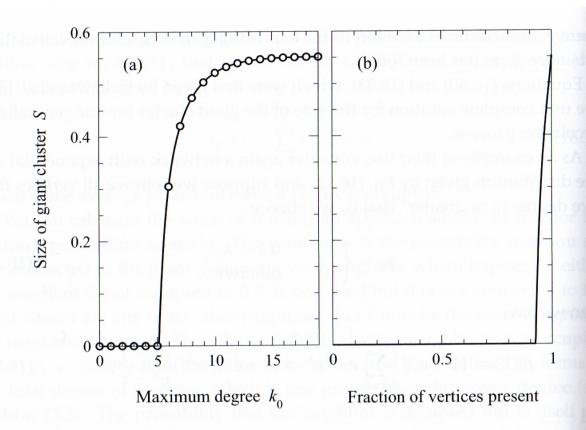


Figure 16.6: Size of the giant percolation cluster as the highest degree vertices in a network are removed. (a) The size of the giant cluster in a network with an exponential degree distribution $p_k \sim \mathrm{e}^{-\lambda k}$ with $\lambda = \frac{1}{2}$ as vertices are removed in order of degree, starting from those with the highest degree. The curve is shown as a function of the degree k_0 of the highest-degree vertex remaining in the network. Technically, since k_0 must be an integer, the plot is only valid at the integer points marked by the circles; the curves are just an aid to the eye. (b) The same data plotted now as a function of the fraction $\overline{\phi}$ of vertices remaining in the network.

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