Algorithmic and Economic Aspects of Networks Nicole Immorlica

Random Graphs

What is a random graph?

Erdos-Renyi Random Graphs

Specify number of vertices n edge probability p

For each pair of vertices i < j, create edge (i,j) w/prob. p G(n,p)

Erdos-Renyi Random Graphs

What does random graph G(n,p) look like? (as a function of p)

Random Graph Demo

http://ccl.northwestern.edu/netlogo/models/GiantComponent

Properties of G(n,p)

p < 1/n disconnected, small tree-like components

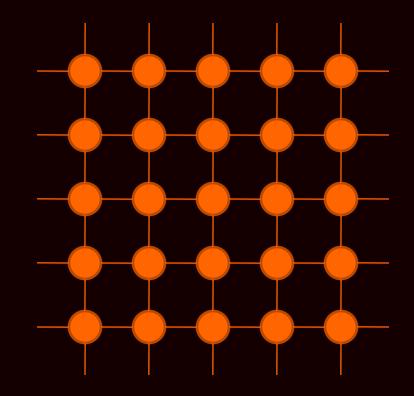
p > 1/n a giant component emerges
containing const. frac. of nodes

Proof Sketch

- 1. Percolation
- 2. Branching processes
- 3. Growing spanning trees

Percolation

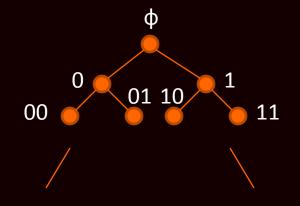
- 1. Infinite graph
- 2. Distinguished node i
- 3. Probability p
 - Each link gets ``open'' with probability p
- Q. What is size of component of i?



Percolation Demo

http://ccl.northwestern.edu/netlogo/models/Percolation

Percolation on Binary Trees



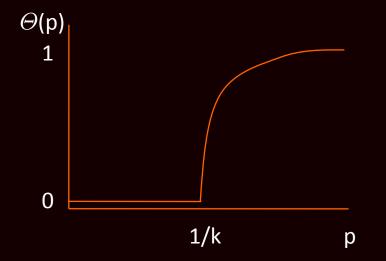
Def. Let $\Theta(p) = \Pr[\operatorname{comp}(\varphi) \text{ is infinite}]$. The critical threshold is $p_c = \sup \{ p \mid \Theta(p) = o \}$.

Critical Threshold

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Thm. Critical threshold of binary trees is $p_c = \frac{1}{2}$. Prf. On board.

Critical Threshold



Thm. Critical threshold of k-ary trees is $p_c = 1/k$.

Branching Processes

Node i has X_i children distributed as B(n,p): $Pr[X_i = k] = (n \text{ choose } k) p^k (1-p)^{(n-k)}$

O. What is probability species goes extinct? A. By percolation, if $p > (1+\epsilon)/n$, live forever.

Note extinction $\leftarrow \rightarrow$ Exists i, $X_1 + ... + X_i < i$.

Erdos-Renyi Random Graphs

We will prove (on board)

(1) If $p = (1-\epsilon)/n$, then there exists $c_1 s.t. Pr[G(n,p) has comp > c_1 log n]$ goes to zero

(2) If $p = (1+2\epsilon)/n$, then there exists c_2 s.t. Pr[G(n,p) has comp > c_2 n] goes to one

First show (on board)

(3) If $p = (1+2\epsilon)/n$, then there exists c_2 , c_3 s.t. Pr[G(n,p) has comp > c_2 n] > c_3

Emergence of Giant Component

Theorem. Let np = c < 1. For $G \in G(n, p)$, w.h.p. the size of the largest connected component is O(log n).

Theorem. Let np = c > 1. For $G \in G(n, p)$, w.h.p. G has a giant connected component of size (β + o(n))n for constant $\beta = \beta c$; w.h.p, the remaining components have size O(log n).

Application

Suppose ...

the world is connected by G(n,p) someone gets sick with a deadly disease all neighbors get infected unless immune a person is immune with probability q

O. How many people will die?

Analysis

- 1. Generate G(n,p)
- 2. Delete qn nodes uniformly at random
- 3. Identify component of initially infected individual

Analysis

Equivalently,

- 1. Generate G((1-q)n, p)
- 2. Identify component of initially infected individual

Analysis

By giant component threshold,

- $p(1-q)n < 1 \rightarrow disease dies$
- $p(1-q)n > 1 \rightarrow$ we die

E.g., if everyone has 50 friends on average, need prob. of immunity = 49/50 to survive!

Summary

Random graphs G(n, c/n) for c > 1 have ...

- 🗸 unique giant component
- ✓ small (logarithmic) diameter
- X low clustering coefficient (= p)
- K Bernoulli degree distribution

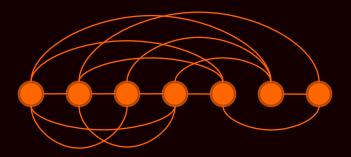
A model that better mimics reality?

In real life

Friends come and go over time.

Growing Random Graphs

On the first day, God created m+1 nodes who were all friends



And on the (m+i)'th day, He created a new node (m+i) with m random friends

Mean Field Approximation

Estimate distribution of random variables by distribution of *expectations*.

E.g., degree dist. of growing random graph?

Degree Distribution

$F_t(d) = 1 - \exp[-(d - m)/m]$ (on board)

This is exponential, but social networks tend to look more like power-law deg. distributions...

In real life

The rich get richer

... much faster than the poor.

Preferential Attachment

Start: m+1 nodes all connected

Time t > m: a new node t with m friends distributed according to degree

> Pr[link to j] = m x deg(j) / \sum deg(.) = m x deg(j) / (2mt) = deg(j) / (2t)

Degree Distribution

Cumulative dist.: $F_t(d) = 1 - m^2/d^2$ Density function: $f_t(d) = 2m^2/d^3$

(heuristic analysis on board, for precise analysis, see Bollobas et al)

A power-law!

Assignment:

- Readings:
 - Social and Economic Networks, Chapters 4 & 5
 - M. Mitzenmacher. A brief history of generative models for power law and lognormal distributions. Internet Mathematics 1, No 2, 226-251, 2005.
 - D.J. Watts, and S.H. Strogatz. *Collective dynamics of small-world networks*. Nature 393, 440-442, 1998.
- Reactions:
 - Reaction paper to one of research papers, or a research paper of your choice
- Presentation volunteer?