

# Random Graphs with Bounded Degrees

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Talk, paper available from: <http://cnls.lanl.gov/~ebn>

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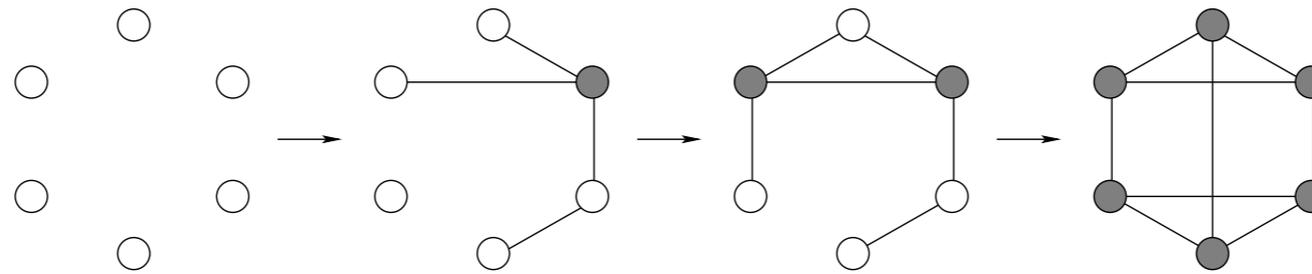
- **Aperitif: Introduction**
  - Motivation
  - Problem: Giant component in a regular random graph
  - Tutorial: Classical Random Graphs
- **Main Course: Regular Random Graphs**
  - Degree distribution
  - Emergence of Giant Component
  - Finite-size scaling Laws
- **Dessert: Shuffling Algorithms and Rings**

# Motivation

In many network problems, the degree is bounded

- Social network: bounded number of friends Wasserman 88
- Power grids: transmission lines
- Communication networks: cellphone towers
- Computer networks Peleg 88
- Physics: bounded number of neighbors in a bead pack Girvan 10
- Chemistry: bounded number of chemical bonds in branched polymers Ballinska 91

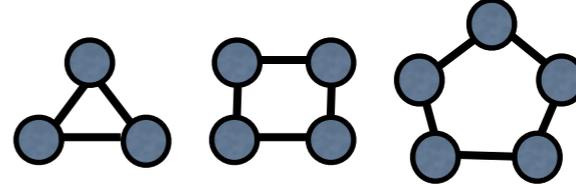
# Problem: Generating Regular Random Graphs



- Initial state: regular random graph (degree = 0 )
- Define two classes of nodes
  - Active nodes: degree  $< d$
  - Inactive nodes: degree =  $d$
- Sequential linking
  - Pick two active nodes
  - Draw a link
- Final state: regular random graph (degree =  $d$  )

# Emergence of the Giant Component

✓  $d=1$  microscopic graphs, dimers 

✓  $d=2$  mesoscopic graphs, rings   $N_k = k^{-1}$

?  $d>2$  one macroscopic graph = “giant component”

- *Dwarf component* phase: microscopic graphs only
- *Giant component* phase: one macroscopic component coexists with many microscopic graphs

## Question

How many links (per node) are needed for the giant component to emerge?

Answer

0.577200

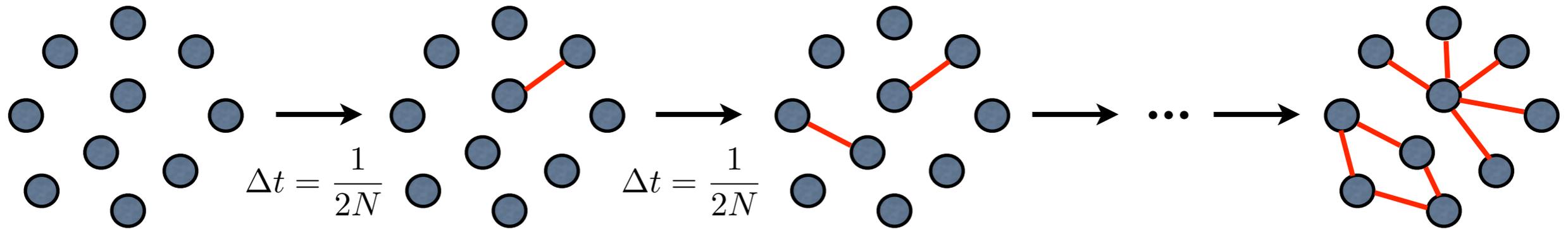
(when  $d=3$ )

# The kinetic approach

- Implicitly take the infinite system size limit
- Implicitly take an average over all realizations of the stochastic process
- Introduce the notion of continuous time variable
- For evolving graphs, time=number of links per node
- Describe the time evolution of probability distributions through differential equations
- Heavily used in physics, chemistry, biology

**Discrete mathematics gone continuous!**

# Evolving Classical Random Graphs



- Initial state:  $N$  isolated nodes
- Dynamical linking
  1. Pick 2 nodes at random
  2. Connect the 2 nodes with a link
  3. Augment time  $t \rightarrow t + \frac{1}{2N}$
- Each node experiences one linking event per unit time

# Degree Distribution

- Distribution of nodes with degree  $j$  at time  $t$  is  $n_j(t)$
- Linking Process is very simple

$$j \rightarrow j + 1$$

- Linear evolution equation

$$\frac{dn_j}{dt} = n_{j-1} - n_j$$

- Initial condition: all nodes are isolated

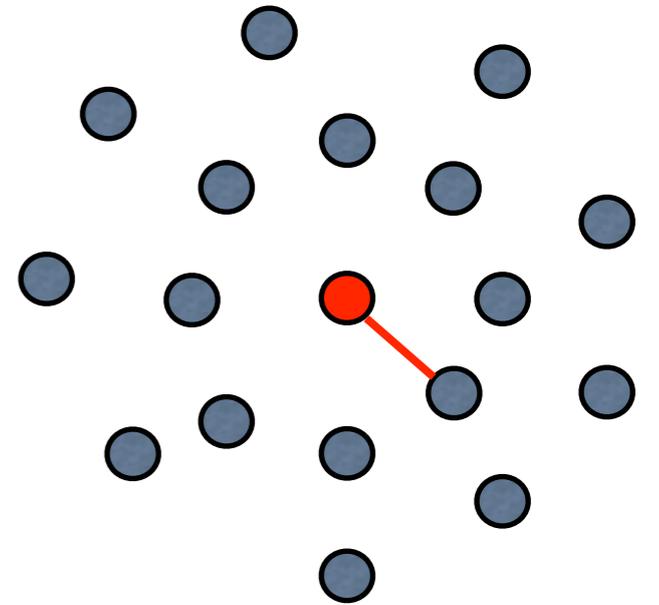
$$n_j(t = 0) = \delta_{j,0}$$

- Degree distribution is Poissonian

$$n_j(t) = \frac{t^j}{j!} e^{-t}$$

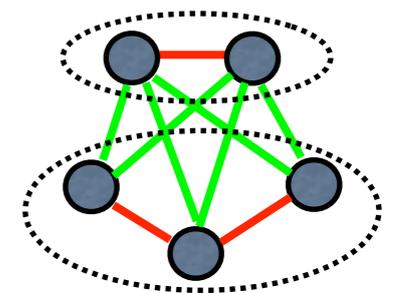
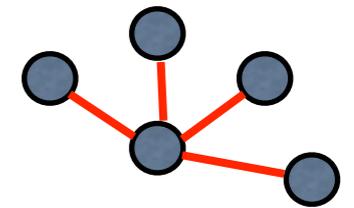
- Average degree characterizes the entire distribution

Random Process, Random Distribution



# Aggregation Process

- Cluster = a connected graph component
- Aggregation rate = product of cluster sizes



$$K_{ij} = ij$$

- Master equation

$$\frac{dc_k}{dt} = \frac{1}{2} \sum_{i+j=k} ij c_i c_j - k c_k$$

$$c_k(t=0) = \delta_{k,1}$$

- Cluster size density

$$c_k(t) = \frac{1}{k \cdot k!} (kt)^{k-1} e^{-kt}$$

- Divergent second moment = emergence of giant component

$$M_2 = \sum_k k^2 c_k$$

$$M_2 = (1-t)^{-1}$$

$$t_g = 1$$

# Detecting the giant component

- Master equation

$$\frac{dc_k}{dt} = \frac{1}{2} \sum_{i+j=k} ij c_i c_j - k c_k \quad c_k(t=0) = \delta_{k,1}$$

- Moments of the size distribution

$$M_n(t) = \sum_k k^n c_k(t)$$

- First moment is conserved

$$\frac{dM_1}{dt} = M_2(M_1 - 1) = 0 \quad \text{when} \quad M_n(t=0) = 1$$

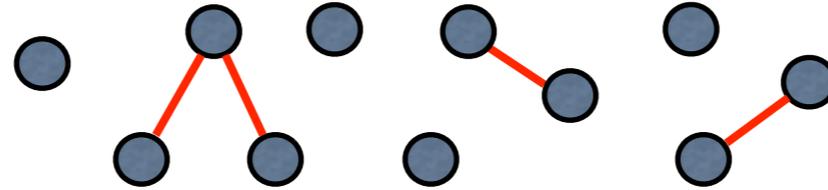
- Second moment obeys closed equation

$$\frac{dM_2}{dt} = M_2^2 \quad M_2(0) = 1$$

- Finite-time singularity signals emergence of infinite cluster

$$M_2 = (1 - t)^{-1}$$

# Dwarf Component Phase ( $t < 1$ )



- Microscopic clusters, tree structure
- Cluster size distribution contains entire mass

$$M(t) = \sum_{k=1}^{\infty} k c_k = 1$$

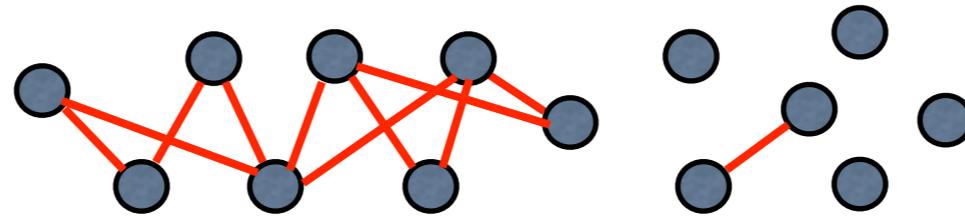
- Typical cluster size diverges near percolation point

$$k_* \sim (1 - t)^{-2}$$

- Critical size distribution has power law tail

$$c_k(1) \simeq \frac{1}{\sqrt{2\pi}} k^{-5/2}$$

# Giant Component Phase ( $t > 1$ )



- Macroscopic component exist, complex structure
- Cluster size distribution contains fraction of mass

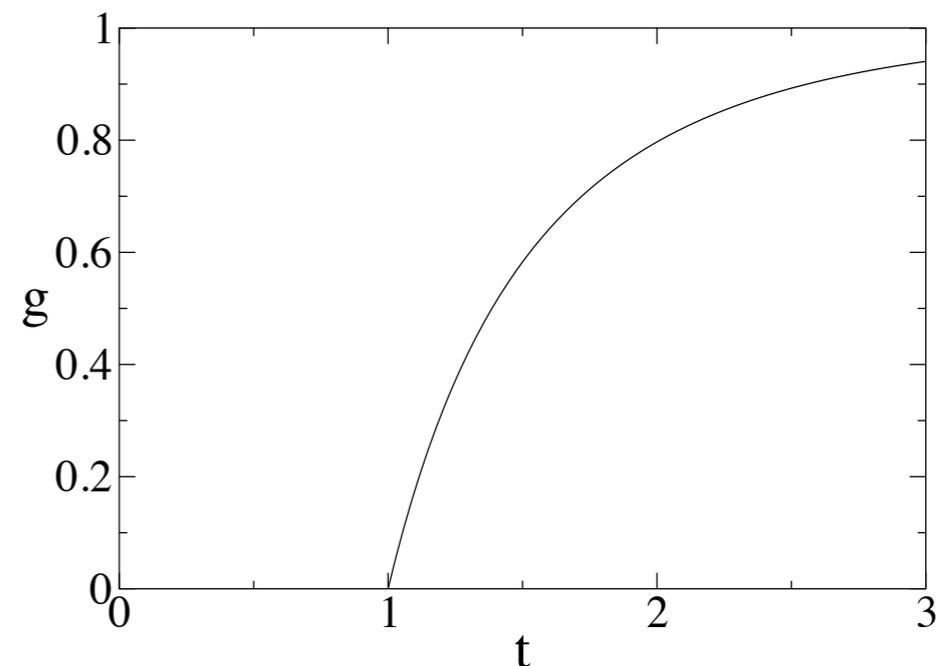
$$M(t) = \sum_{k=1}^{\infty} k c_k = 1 - g$$

- Giant component accounts for “missing” mass

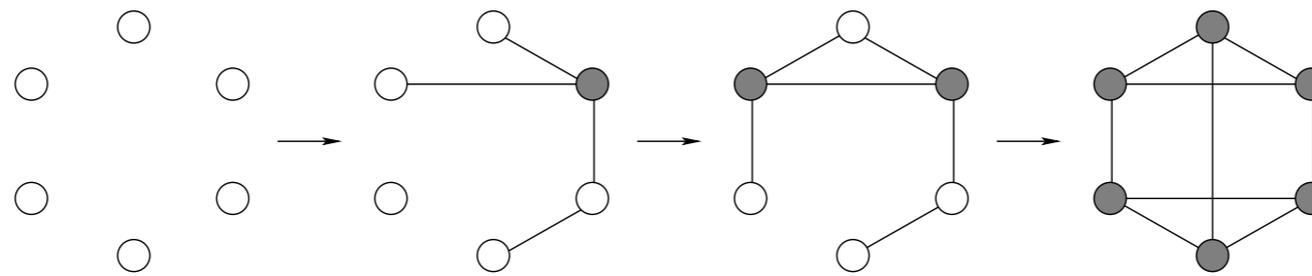
$$g = 1 - e^{-gt}$$

- Giant component takes over entire system

$$g \rightarrow 1$$



# Generating a Regular Random Graph



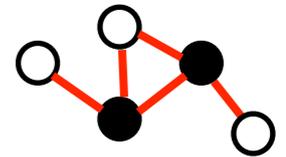
- Initial state: regular random graph (degree = 0 )
- Define two classes of nodes
  - Active nodes: degree  $< d$
  - Inactive nodes: degree =  $d$
- Sequential linking
  - Pick two active nodes
  - Draw a link
- Final state: regular random graph (degree =  $d$  )

# Degree Distribution

- Distribution of nodes with degree  $j$  is  $n_j$
- Density of active nodes  $\nu = n_0 + n_1 + \dots + n_{d-1}$   $\nu = 1 - n_d$

- Linking Process

$$(i, j) \rightarrow (i + 1, j + 1) \quad i, j < d$$



- Active nodes control linking process, effectively linear evolution equation

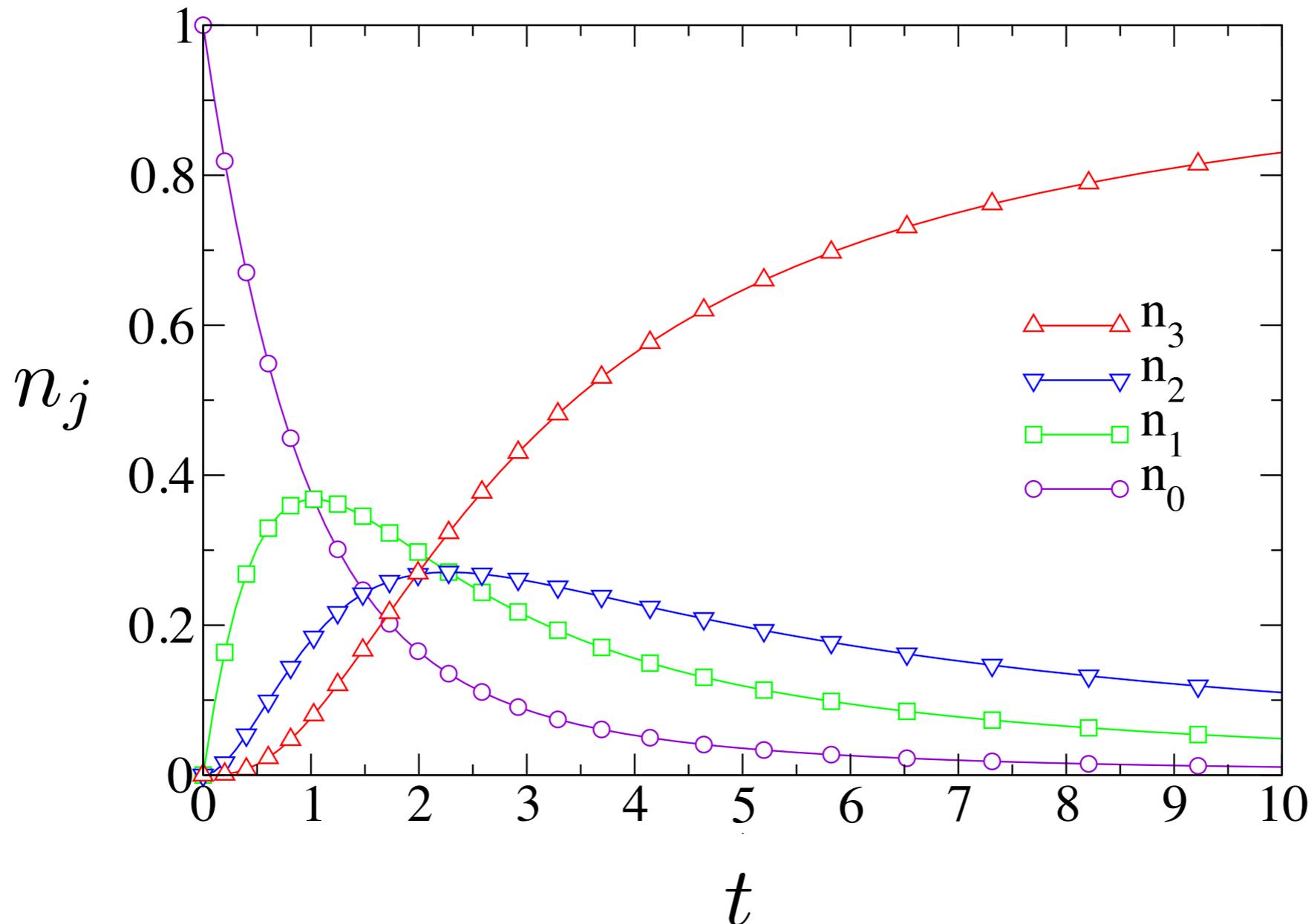
$$\frac{dn_j}{dt} = \nu (n_{j-1} - n_j) \quad \xrightarrow{\tau = \int_0^t dt' \nu(t')} \quad \frac{dn_j}{d\tau} = n_{j-1} - n_j$$

- Solve using an effective time variable

$$n_j = \frac{\tau^j}{j!} e^{-\tau} \quad j < d$$

Truncated Poisson Distribution

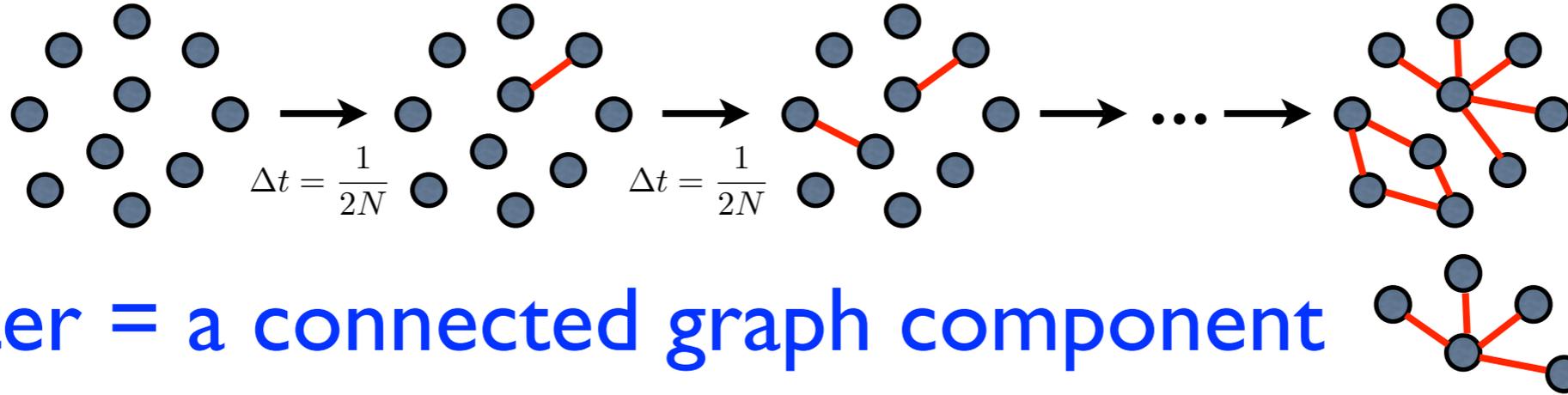
# Degree Distribution



**Isolated nodes dominate initially**  
**All nodes become inactive eventually**

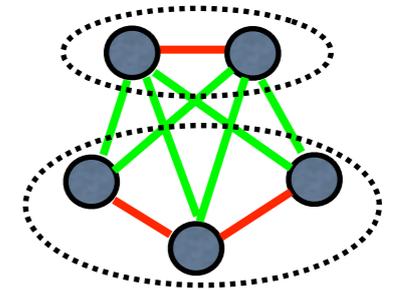
# Unbounded Random Graphs

Erdos-Renyi



- Cluster = a connected graph component
- Links involving two separate components lead to merger
- Aggregation rate = product of cluster sizes

$$K_{ij} = ij$$



- Master equation for size distribution

$$\frac{dc_k}{dt} = \frac{1}{2} \sum_{i+j=k} ij c_i c_j - k c_k$$

$$c_k(t=0) = \delta_{k,1}$$

- Master equation for generating function

$$\frac{\partial \mathcal{C}}{\partial t} + x \frac{\partial \mathcal{C}}{\partial x} = \frac{1}{2} \left( x \frac{\partial \mathcal{C}}{\partial x} \right)^2$$

$$\mathcal{C}(x, t) = \sum_k c_k(t) x^k$$

# Hamilton-Jacobi Theory I

- Master equation is a first-order PDE

$$\frac{\partial \mathcal{C}}{\partial t} + x \frac{\partial \mathcal{C}}{\partial x} = \frac{1}{2} \left( x \frac{\partial \mathcal{C}}{\partial x} \right)^2 \quad \mathcal{C}(x, 0) = x$$

- Recognize as a Hamilton-Jacobi equation

$$\frac{\partial \mathcal{C}(x, t)}{\partial t} + H(x, p) = 0$$

- By identifying “momentum” and “Hamiltonian”

$$p = \frac{\partial \mathcal{C}}{\partial x} \quad \text{and} \quad H = xp - \frac{1}{2}(xp)^2$$

- Hamilton-Jacobi equations generate two coupled ODEs

$$\frac{dx}{dt} = \frac{\partial H}{\partial p}, \quad \frac{dp}{dt} = -\frac{\partial H}{\partial x} \quad \Longrightarrow \quad \frac{dx}{dt} = x(1 - xp), \quad \frac{dp}{dt} = -p(1 - xp)$$

$x(0) = 1 - g \quad p(0) = 1$

Initial coordinate unknown, final coordinate known!

Hamiltonian is a conserved quantity

# Solution I

- Coordinate and momentum are immediate

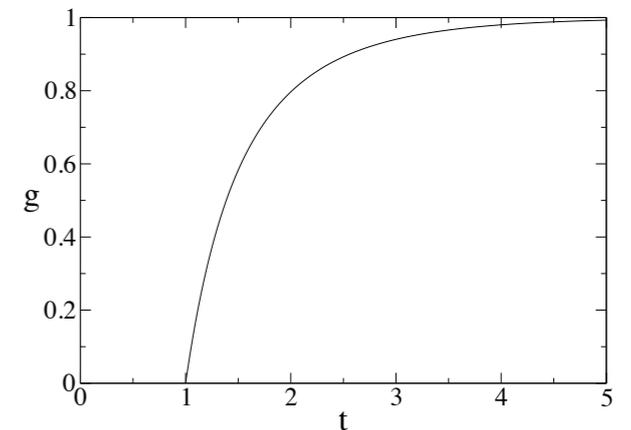
$$x = (1 - g)e^{gt} \quad p = e^{-gt}$$

- Size of giant component found immediately

$$g = 1 - \sum_k k c_k = 1 - p(0)$$

- Satisfies a closes equation

$$1 - g = e^{-gt}$$



- Nontrivial solution beyond the percolation threshold

$$t_g = 1$$

The giant component emerges when  
the average degree equals one

# Bounded Random Graphs

- Total size of components provides insufficient description
- Describe components by a  $d+1$  dimensional vector whose components specify number of nodes with given degree

$$(k_0, k_1, \dots, k_d) \qquad k = k_0 + k_1 + \dots + k_d$$

$(0, 2, 1, 2)$

$(0, 3, 1, 1)$

- Multivariate aggregation process
- Aggregation rate is product of the number of active nodes

$$K(\mathbf{l}, \mathbf{m}) = (l - l_d)(m - m_d)$$

- Why can't we get away with two variables only?
- Node degrees are coupled!

$$0 \rightarrow 1 \rightarrow 2 \rightarrow 3$$

# Hamilton-Jacobi Theory II

- Master equation is a first-order PDE

$$\frac{\partial C}{\partial \tau} = \frac{1}{2\nu} \left( \sum_{j=0}^{d-1} x_{j+1} \frac{\partial C}{\partial x_j} \right)^2 - \sum_{j=0}^{d-1} x_j \frac{\partial C}{\partial x_j} \quad C(\mathbf{x}, 0) = x_0$$

- Recognize as a Hamilton-Jacobi equation

$$\frac{\partial C(\mathbf{x}, \tau)}{\partial \tau} + H(\mathbf{x}, \nabla C, \tau) = 0$$

- By identifying “momentum” and “Hamiltonian”

$$H(\mathbf{x}, \mathbf{p}, \tau) = \sum_{j=0}^{d-1} x_j p_j - \frac{\Pi_1^2}{2\nu(\tau)} \quad \Pi_j = \sum_{i=j}^d x_i p_{i-j}$$

- Hamilton-Jacobi equation give  $2(d+1)$  coupled ODEs

$$\frac{dx_j}{dt} = \frac{\partial H}{\partial p_j}, \quad \frac{dp_j}{dt} = -\frac{\partial H}{\partial x_j} \quad \Longrightarrow \quad \frac{dx_j}{dt} = x_j - \frac{\Pi_1}{\nu} x_{j+1}, \quad \frac{dp_j}{dt} = \frac{\Pi_1}{\nu} p_{j-1} - p_j$$

Initial coordinates unknown, final coordinates known!

Equations are now in  $d+1$  dimensions!

Hamiltonian no longer conserved!

# Hamilton-Jacobi Equations

- Coupled differential equations for coordinate and momenta

$$\frac{dx_j}{d\tau} = \begin{cases} x_j - \frac{\Pi_1}{\nu} x_{j+1} & j < d \\ 0 & j = d \end{cases} \quad \text{and} \quad \frac{dp_j}{d\tau} = \frac{dx_j}{d\tau} = \begin{cases} x_j - \frac{\Pi_1}{\nu} x_{j+1} & j < d \\ 0 & j = d \end{cases}$$

- Initial conditions: (i) known for momenta (ii) unknown coordinates!

$$p_j(0) = \delta_{j,0} \quad \text{and} \quad x_j(0) = y_j \quad C(\mathbf{x}, 0) = x_0$$

- Identify conservation laws!

$$\frac{d\Pi_0}{d\tau} = 0 \quad \text{and} \quad \frac{dx_d}{d\tau} = 0 \quad \Pi_j = \sum_{i=j}^d x_i p_{i-j}$$

- Backward evolution equations for the initial coordinates!

$$\frac{dy_j}{d\tau} = \sum_{i=0}^{d-j-1} \left[ \frac{du}{d\tau} x_{i+j+1} - x_{i+j} \right] p_i \quad u = \int_0^\tau d\tau' \frac{\Pi_1(\tau')}{\nu(\tau')}$$

# Solution II

- Find hidden conservation laws and explicit backward equations
- reduce  $2(d+1)$  first order ODE to 1 second order ODE

$$\frac{d^2 u}{d\tau^2} + \frac{n_{d-1}}{\nu} \frac{du}{d\tau} - x_d \frac{p_{d-1}}{\nu} = 0 \quad u = \int_0^\tau d\tau' \frac{\Pi_1(\tau')}{\nu(\tau')}$$

- Nontrivial solution when  $d > 2$
- Numerical solution gives percolation threshold ( $d=3$ )

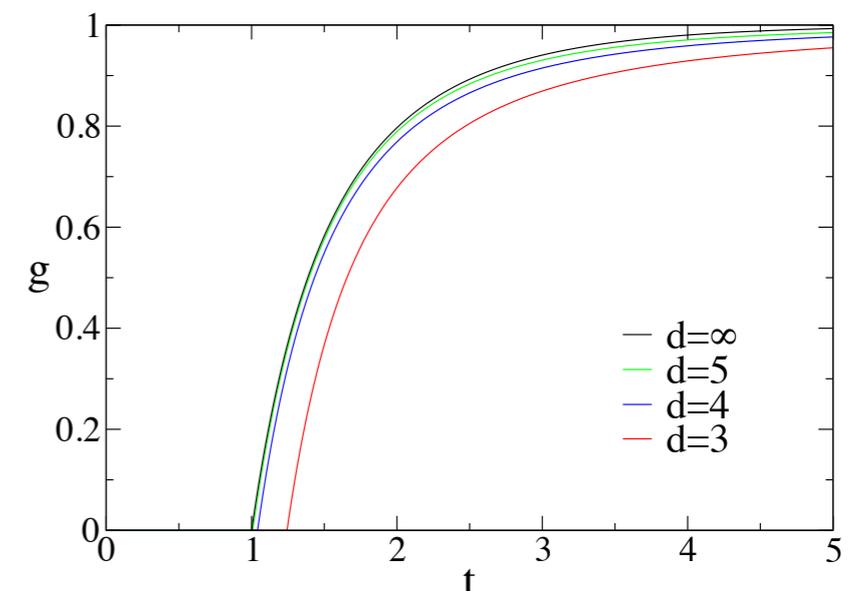
$$t_g = 1.243785, \quad L_g = 0.577200$$

- The size distribution of components at the critical point

$$c_k \simeq A k^{-5/2}$$

- Mean-field percolation

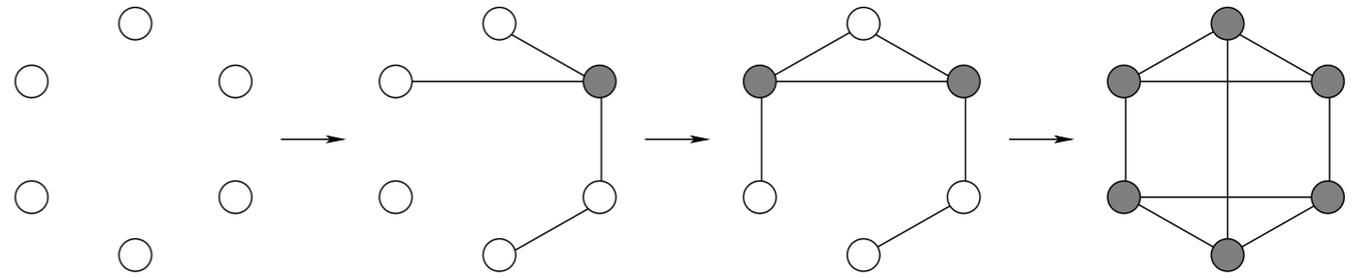
Hamilton-Jacobi theory gives  
all percolation parameters



# Finite-size scaling

## Degree distribution

$$n_j \simeq \frac{(d-1)!}{j!} t^{-1} (\ln t)^{-(d-1-j)}$$



## Regular random graph emerges in several steps

### 1. Giant component emerges at finite time

$$t_1 = 1.243785$$

deterministic

### 2. Graph becomes fully connected emerges at time

$$N n_0 \sim 1 \implies t_2 \sim N (\ln N)^{-(d-1)}$$

stochastic

### 3. Regular random graph emerges at time

$$N n_{d-1} \sim 1 \implies t_3 \sim N$$

stochastic

Giant fluctuations in completion time

# General Random Graphs

- Theory straightforward to generalize
- Degree controls linking process

$(i, j) \rightarrow (i + 1, j + 1)$  with rate  $C_{i,j}$

- Connection rate is **arbitrary**
- Equation for generating function

$$\frac{\partial C}{\partial t} = \frac{1}{2} \sum_{i,j} C_{i,j} \left[ \left( x_{i+1} \frac{\partial C}{\partial x_i} \right) \left( x_{j+1} \frac{\partial C}{\partial x_j} \right) - 2n_i \left( x_j \frac{\partial C}{\partial x_j} \right) \right]$$

- **Hamiltonian**

$$H(\mathbf{x}, \mathbf{p}, t) = \sum_j \nu_j(t) x_j p_j - \frac{1}{2} \sum_{i,j} C_{i,j} (x_{i+1} p_i) (x_{j+1} p_j)$$

**Conservation laws neither obvious nor guaranteed**  
**Multi-dimensional Newton solver**

# Summary

- Dynamic formation of regular random graphs
- Degree distribution is truncated Poissonian
- Hamilton-Jacobi formalism powerful
- Percolation parameters with essentially arbitrary precision
- Mean-field percolation universality class
- A multitude of finite-size scaling properties
- Giant fluctuations in completion time

Theory applicable to broader set of evolving graphs

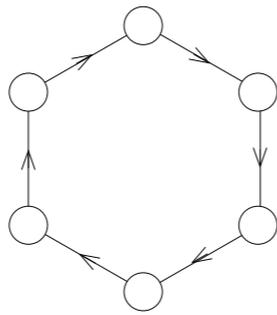
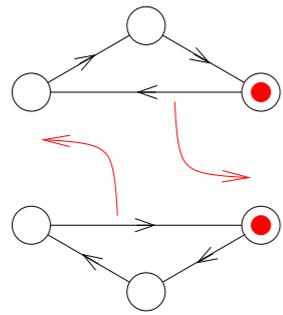
# Shuffling Algorithm

1 2 3 4 5 6  $\rightarrow$  1 5 3 4 2 6  $\rightarrow$  1 5 4 3 2 6  $\rightarrow \dots$

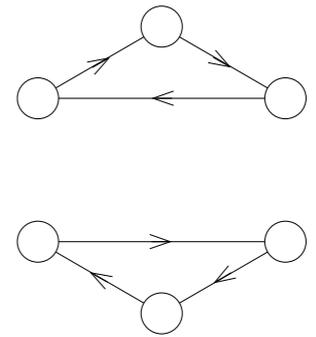
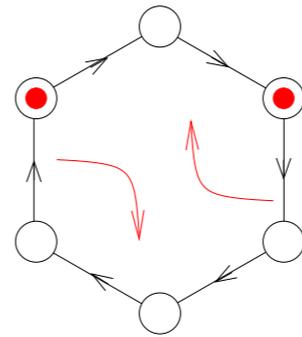
- Initial configuration:  $N$  ordered integers
- Pairwise shuffling:
  1. Pick 2 numbers at random
  2. Exchange positions
  3. Augment time  $t \rightarrow t + \frac{1}{2N}$
- Each integer is shuffled once per unit time
- Efficient algorithm, computational cost is  $\mathcal{O}(N)$

Isomorphic to dynamical regular random graph with  $d=2$ !

# Cycles and Permutations



$$(1\underline{23})(4\underline{56}) \rightarrow (156423)$$



$$(1\underline{56}4\underline{23}) \rightarrow (123)(456)$$

- Cycle structure of a permutation

$$134265 \implies (1)(234)(56)$$

- **Aggregation:** inter-cycle shuffling

$$i, j \xrightarrow{K_{ij}} i + j \quad \text{with} \quad K_{ij} = ij$$

- **Fragmentation:** intra-cycle shuffling

$$i + j \xrightarrow{F_{ij}} i, j \quad \text{with} \quad F_{ij} = \frac{i + j}{N}$$

Identical aggregation and fragmentation rates

# Steady-State Distribution

- Steady-state size distribution satisfies

$$0 = \frac{1}{2} \sum_{i+j=k} K_{ij} c_i c_j - c_k \sum_{j \geq 1} K_{kj} c_j + \sum_{j \geq 1} F_{kj} c_{j+k} - \frac{1}{2} c_k \sum_{i+j=k} F_{ij}$$

- Detailed balance condition

$$K_{ij} c_i c_j = F_{ij} c_{i+j}$$

Lowe 95

- Substitute aggregation and fragmentation rates

$$K_{ij} = ij \quad F_{ij} = \frac{i+j}{N}$$

- Steady-state solution

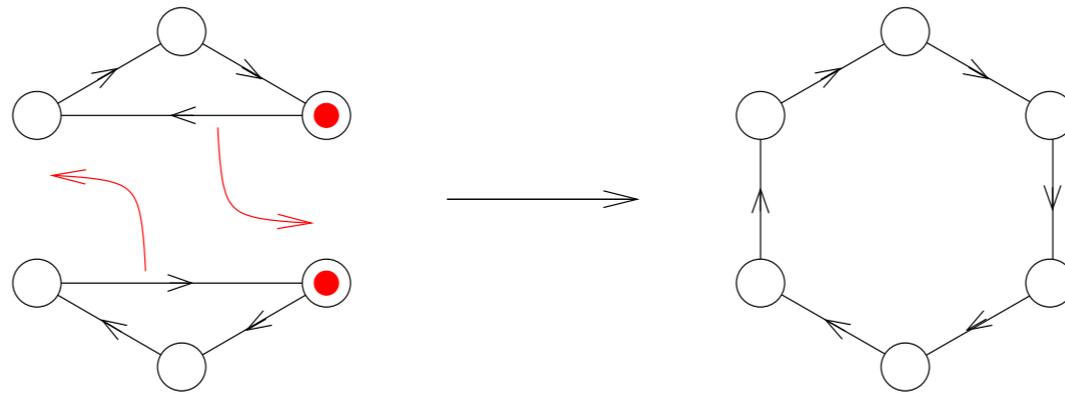
$$(ic_i)(jc_j) = \frac{1}{N} (i+j)c_{i+j} \implies Nc_k = \frac{1}{k}$$

- Average number of cycles

$$N_k = \frac{1}{k}$$

Kunth, art of comp. prog. vol 3

# Redirection Process



- **Dynamical redirection**

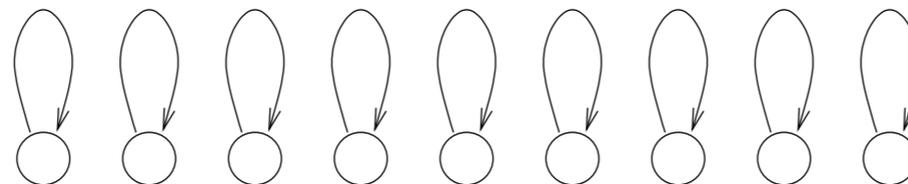
1. Pick 2 nodes at random

2. Connect 2 nodes by redirecting 2 associated links

3. Augment time  $t \rightarrow t + \frac{1}{2N}$

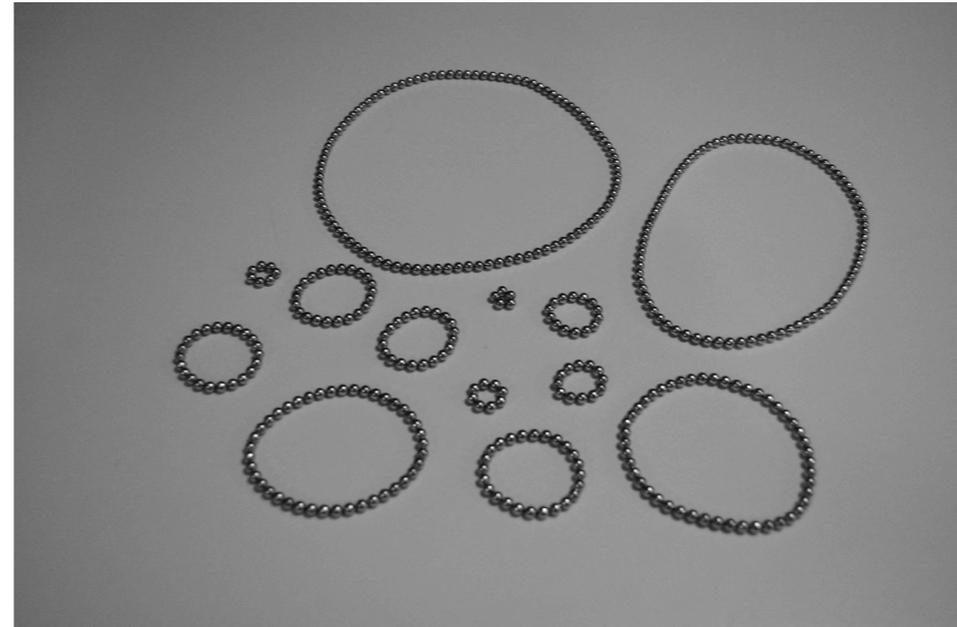
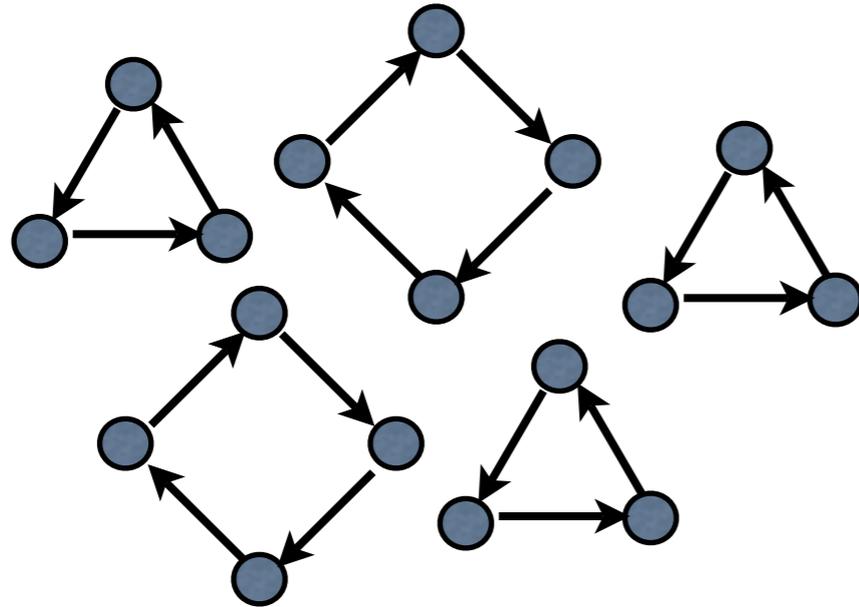
- A node experiences one redirection event per unit time

- Initial condition: isolated nodes, each has a self-link



**Redirection process maintains ring topology**

# Rings

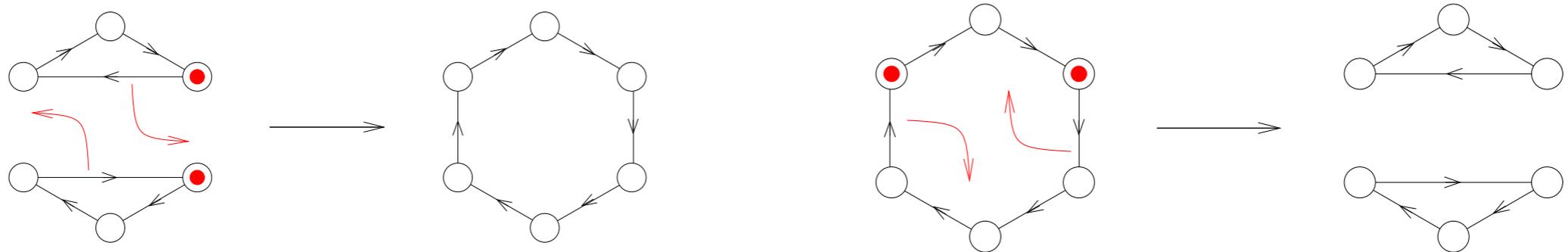


Talia Ben-Naim, age 9

- All nodes have identical degree
- Motivation: rings of magnetic particles
- Consider simplest case: rings; all nodes have degree 2
- Consider directed links (without loss of generality)
- In a system of  $N$  nodes, there are exactly  $N$  links

**Number of links is conserved!**

# Aggregation-Fragmentation Process



- **Aggregation:** inter-ring redirection

Identical to random graph process

$$i, j \xrightarrow{K_{ij}} i + j \quad \text{with} \quad K_{ij} = ij$$

- **Fragmentation:** intra-ring redirection

Fragmentation rate depends on system size!

$$i + j \xrightarrow{F_{ij}} i, j \quad \text{with} \quad F_{ij} = \frac{i + j}{N}$$

- Total fragmentation rate is quadratic

$$F_k = \sum_{i+j=k} F_{ij} = \frac{k(k-1)}{2N}$$

**Reversible process**

# Rate Equations

- Size distribution satisfies

$$\frac{dr_k}{dt} = \frac{1}{2} \sum_{i+j=k} ij r_i r_j - k r_k + \frac{1}{N} \left[ \sum_{j>k} j r_j - \frac{k(k-1)}{2} r_k \right]$$

- Rate equation includes explicit dependence on  $N$

- Perturbation theory

$$r_k = f_k + \frac{1}{N} g_k$$

finite rings
giant rings  
↓
↓

- Fragmentation irrelevant for finite rings  $F_k \sim \frac{k^2}{N}$

$$\frac{df_k}{dt} = \frac{1}{2} \sum_{i+j=k} ij f_i f_j - k f_k$$

Recover random graph equation

# Finite Rings Phase ( $t < 1$ )

- All rings are finite in size

$$M(t) = \sum_{k=1}^{\infty} f_k = 1$$

- Size distribution

$$f_k(t) = \frac{1}{k \cdot k!} (kt)^{k-1} e^{-kt}$$

- Second moment diverges in finite time  $M_2 = \sum_k k^2 f_k$

$$\frac{dM_2}{dt} = M_2^2 \quad \Longrightarrow \quad M_2 = (1 - t)^{-1}$$

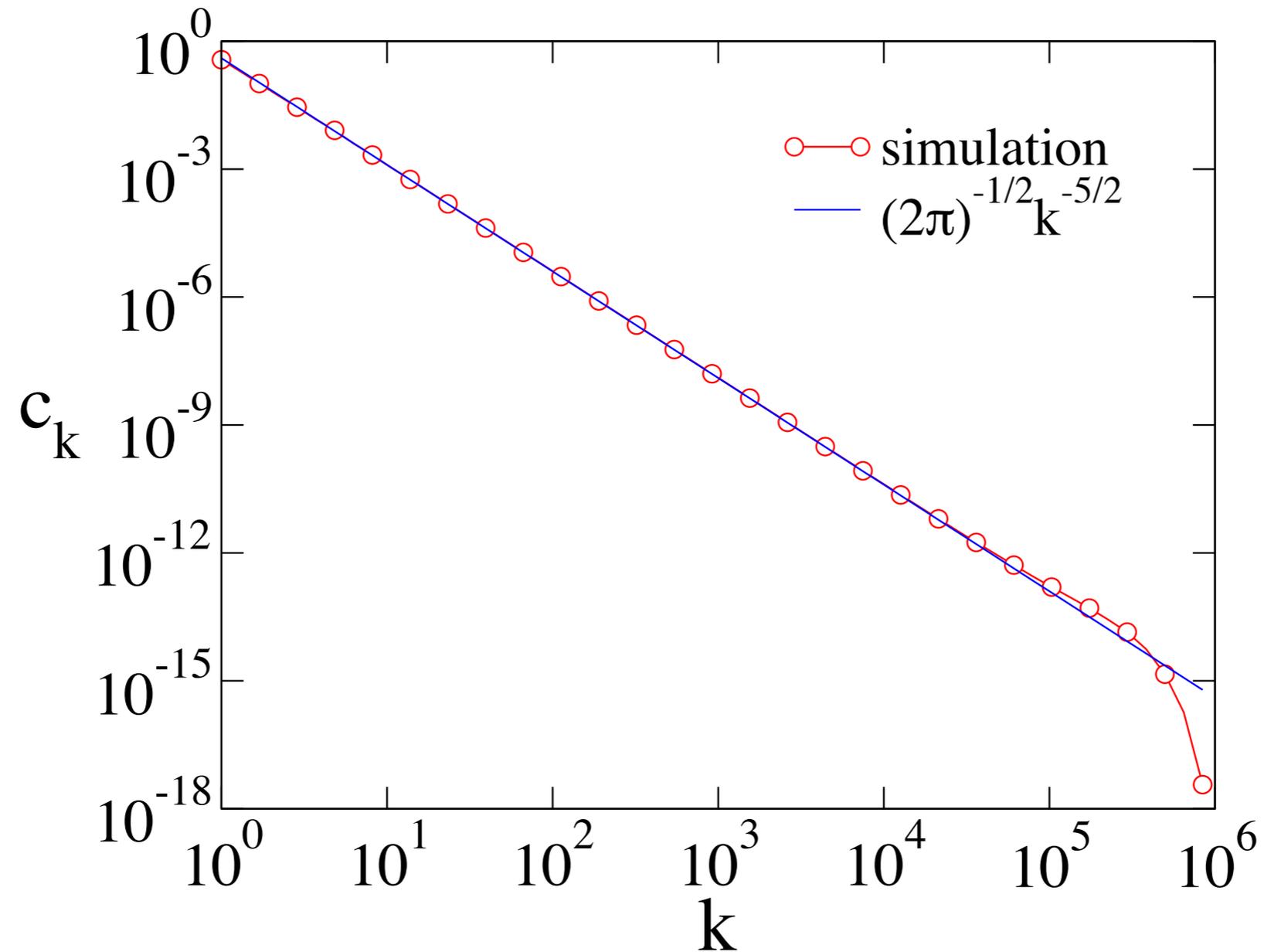
- Critical size distribution

$$f_k(1) \simeq \frac{1}{\sqrt{2\pi}} k^{-5/2}$$

Identical behavior to good-old random graph

# Critical Size Distribution

## Simulation results



Excellent agreement between theory and simulation

# Giant Rings Phase ( $t > 1$ )

- Finite rings contain only a fraction of  $g$  all mass

$$M(t) = \sum_{k=1}^{\infty} k f_k = 1 - g$$

- “Missing Mass”  $1-g$  must be found in giant rings

$$g = 1 - e^{-gt}$$

- Expect giant, macroscopic rings
- Very fast aggregation and fragmentation processes

$$F_k \sim \frac{k^2}{N} \sim N \quad \text{when} \quad k \sim N$$

Fragmentation comparable to aggregation  
No longer negligible

# Distribution of giant rings

- Quantify giant rings by normalized size  $\ell = \frac{k}{N}$
- Average number of giant rings of normalized size  $\ell$

$$g(t) = \int_0^{g(t)} d\ell \ell G(\ell, t)$$

- Rate equation

~~$$\frac{1}{N} \frac{\partial G(\ell, t)}{\partial t} = \frac{1}{2} \int_0^\ell ds s(\ell - s) G(s, t) G(\ell - s, t) - \ell(g - \ell) G(\ell, t)$$

$$+ \int_\ell^g ds s G(s, t) - \frac{1}{2} \ell^2 G(\ell, t)$$~~

agg gain =  $\ell/2$                       agg loss =  $g - \ell$   
frag gain =  $g - \ell$                       frag loss =  $\ell/2$

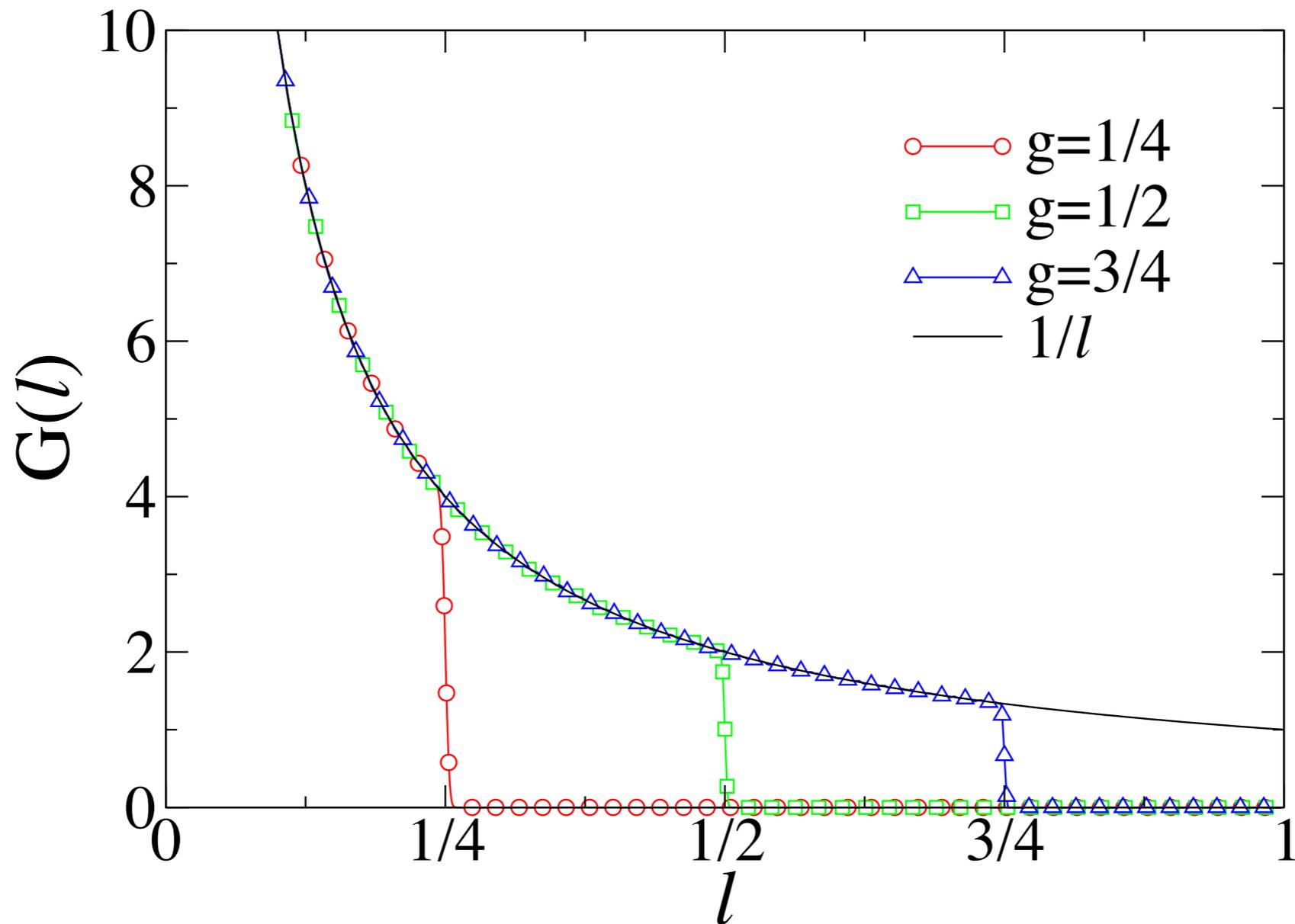
- Quasi steady-state

$$G(\ell, t) = \begin{cases} \ell^{-1} & \ell < g(t), \\ 0 & \ell > g(t). \end{cases}$$

Universal distribution, span grows with time

# Average Number of Giant Rings

Simulation results



$$G(l, t) = \begin{cases} l^{-1} & l < g(t), \\ 0 & l > g(t). \end{cases}$$

# Comments

- Rate equation for average number of giant rings

$$\frac{1}{N} \frac{\partial G(\ell, t)}{\partial t} = \frac{1}{2} \int_0^\ell ds s(\ell - s) G(s, t) G(\ell - s, t) - \ell(g - \ell) G(\ell, t) + \int_\ell^g ds s G(s, t) - \frac{1}{2} \ell^2 G(\ell, t)$$

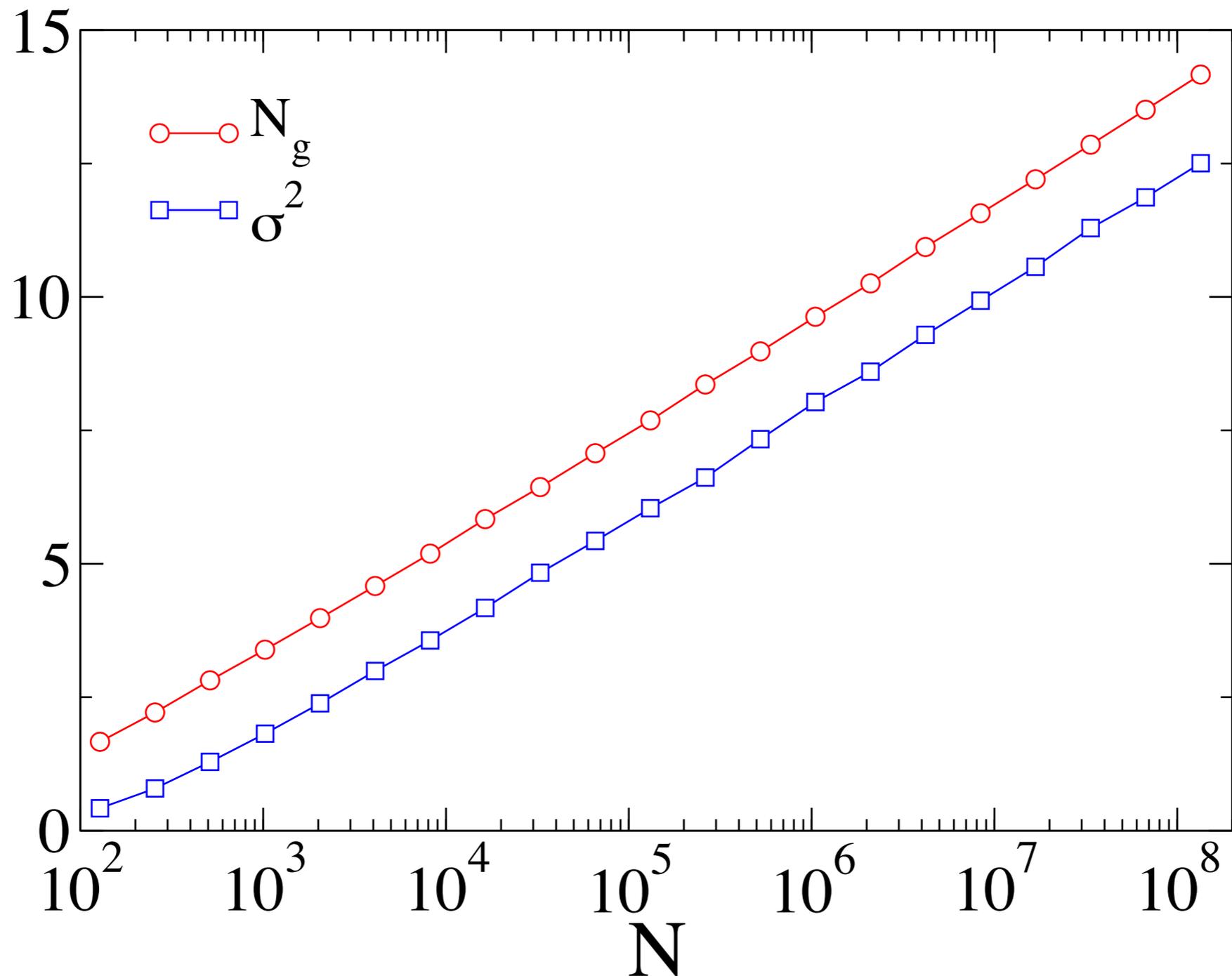
- Practically closed equation; coupling to finite rings only through total mass  $g(t)$
- Steady flux  $N dg/dt$  from finite rings to giant rings
- Number of giant rings is not proportional to  $N$ !

$$N_g \simeq \ln N$$

Number of microscopic rings proportional to  $N$   
Number of macroscopic rings logarithmic in  $N$

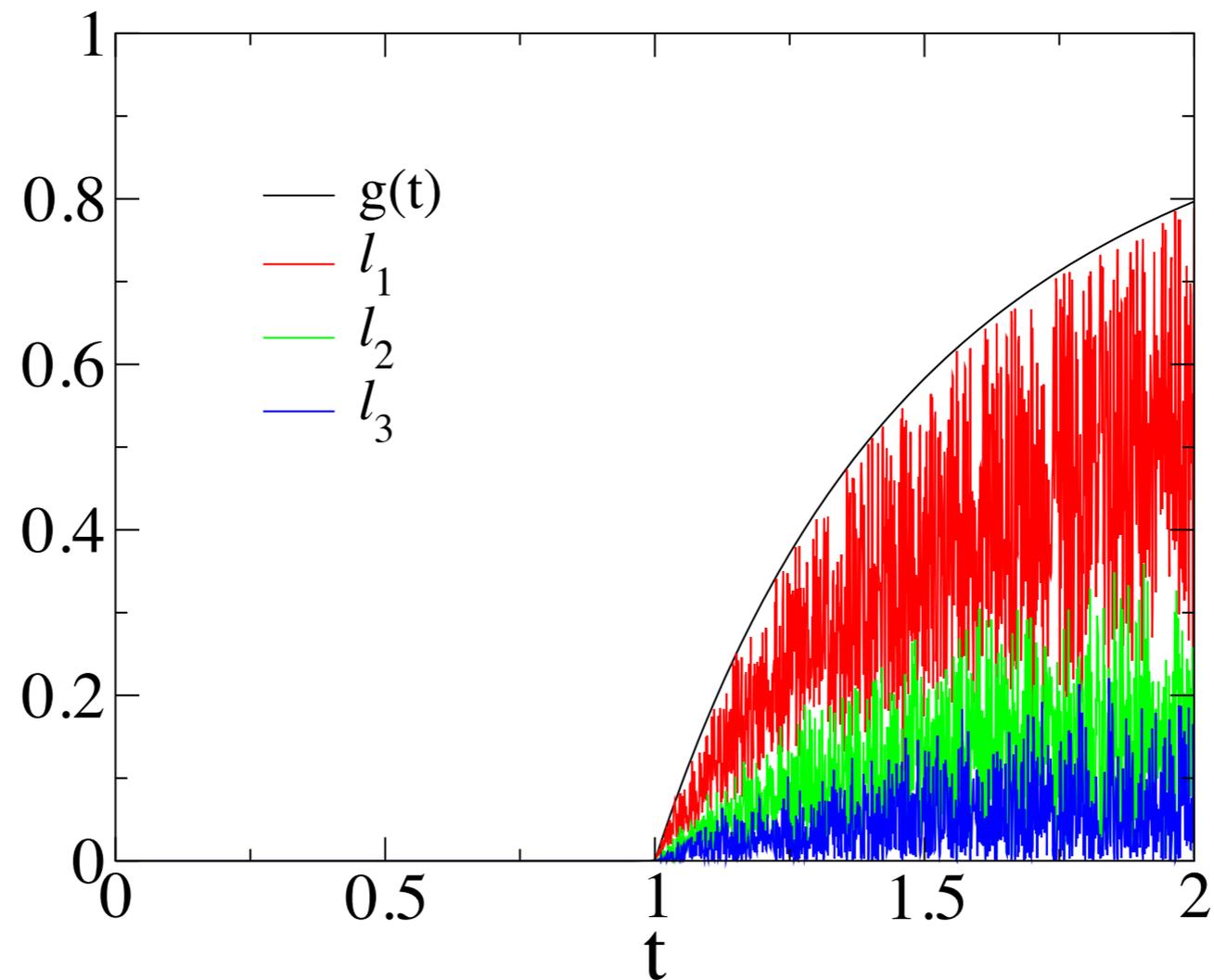
# Total Number of Giant Rings

Simulation results



Law of large numbers

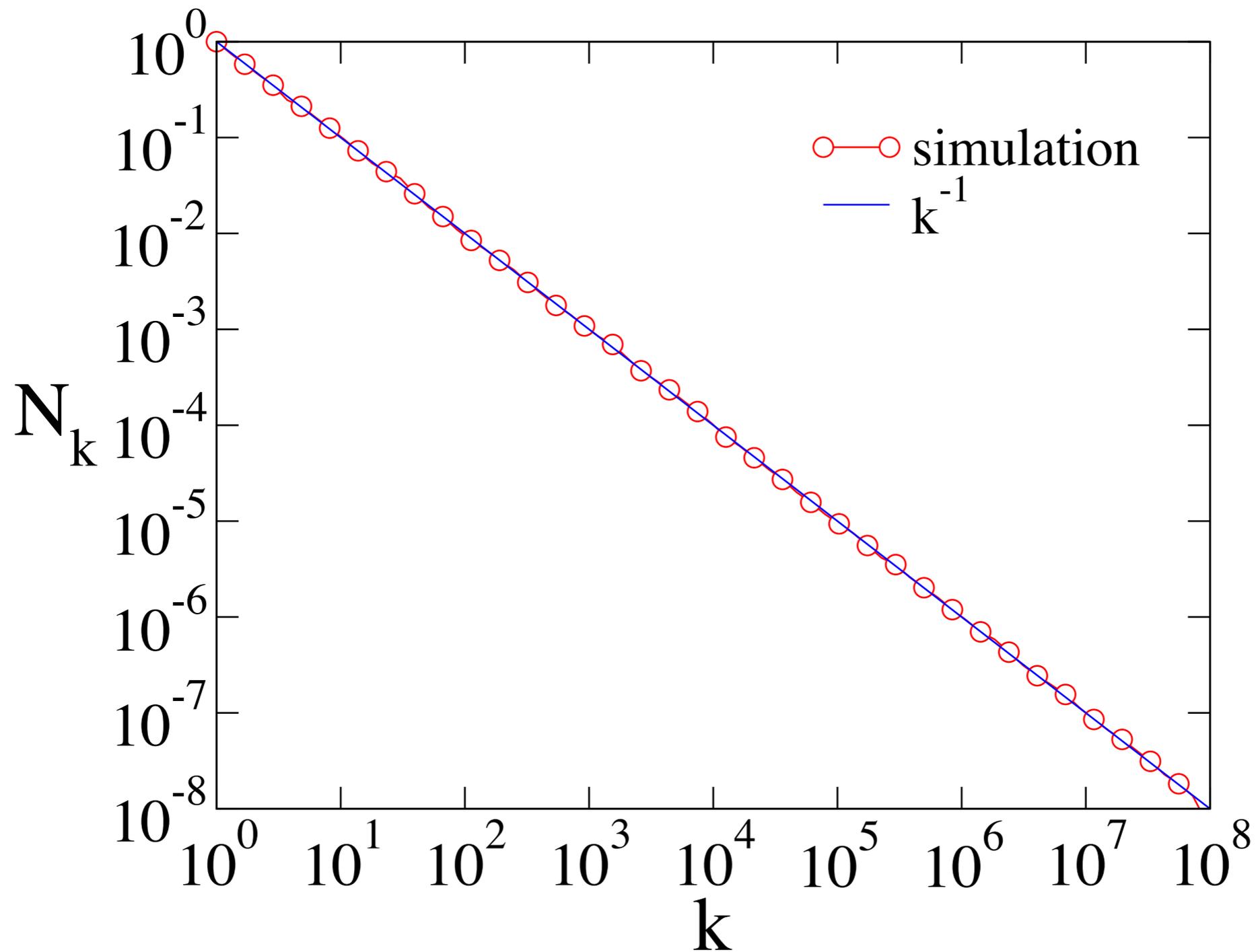
# Multiple Coexisting Giant Rings



Total mass of giant rings is a deterministic quantity  
Mass of an individual giant ring is a stochastic quantity!  
Giant rings break and recombine very rapidly

# Final Distribution

Simulation results



# Implications to Shuffling

- $N$  pairwise shuffles generate a giant cycle
- Size of emergent giant cycle is  $N^{2/3}$
- $N \ln N$  pairwise shuffles generate random order

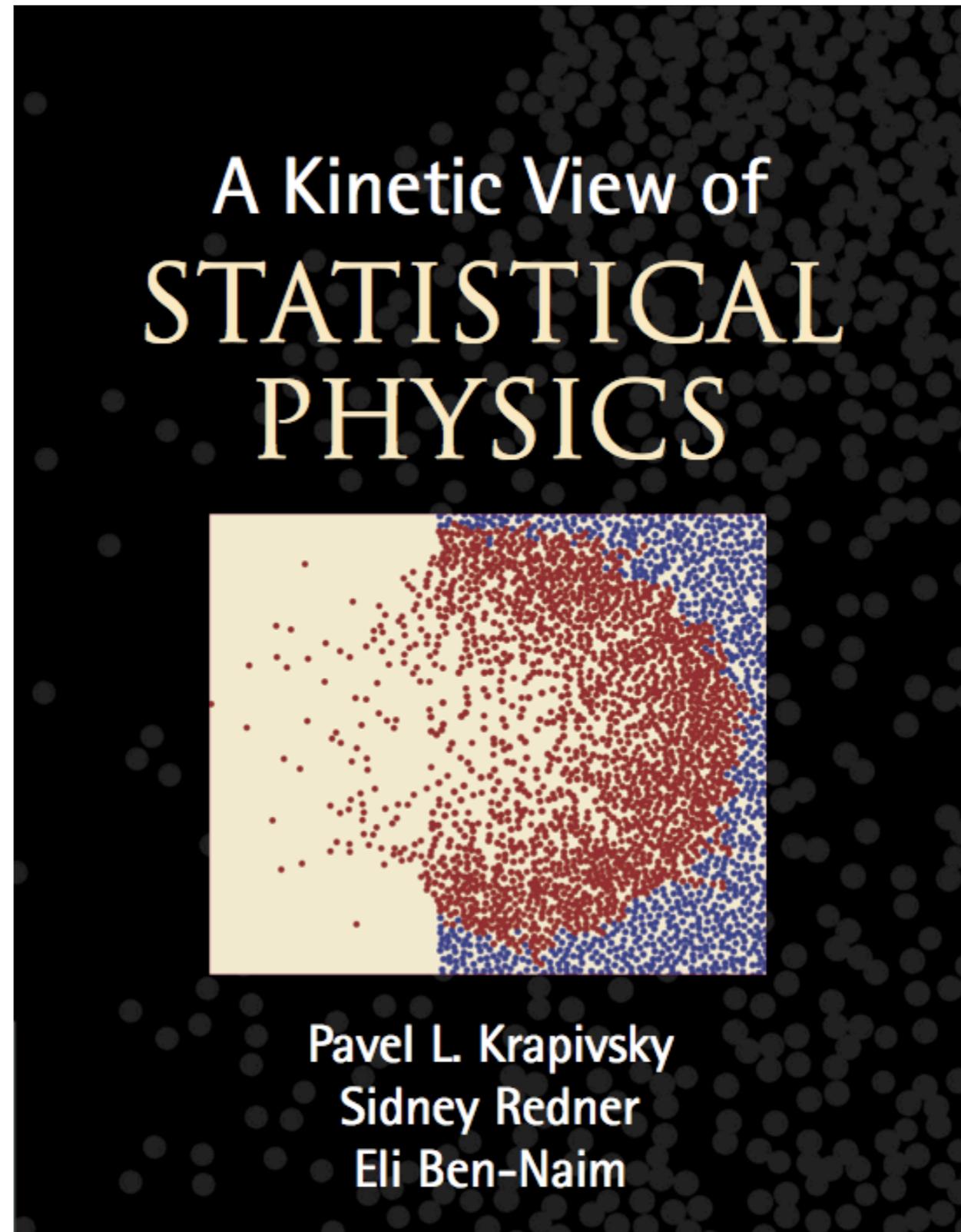
# Summary

- Kinetic formulation of a regular random graph
- Equivalent to: (i) aggregation-fragmentation (ii) shuffling
- Finite rings phase: fragmentation is irrelevant
- Giant rings phase
  - Multiple giant rings coexist
  - Number of giant rings fluctuates
  - Total mass is a deterministic quantity
  - Very rapid evolution

chapter 5  
aggregation

chapter 12  
population  
dynamics

chapter 13  
complex  
networks



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