

# Computational Complexity and Growth Models in Statistical Physics

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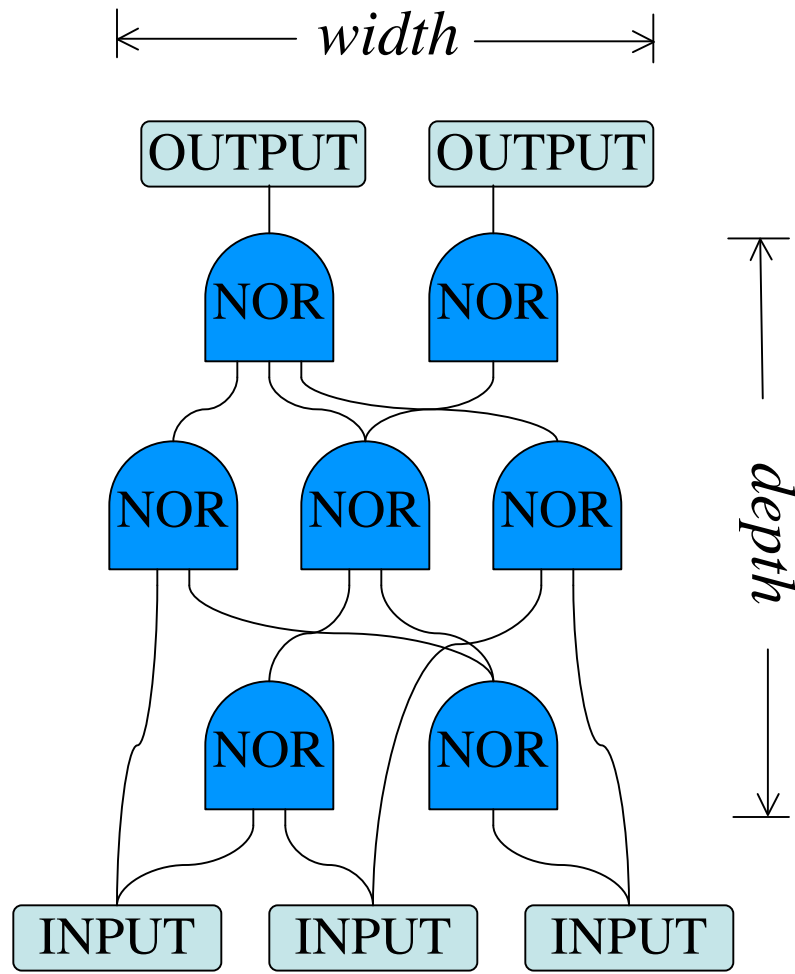
# Outline

- Parallel computing and computational complexity
- Diffusion limited aggregation
- Growing networks
- Physical complexity and computational complexity

# Computational Complexity

- How do computational resources scale with the size of the problem?
  - Time
  - Hardware
- Equivalent results independent of the model of computation.
  - Turing machine
  - ✓ **Parallel random access machine**
  - ✓ **Boolean circuit family**
  - Formal logic

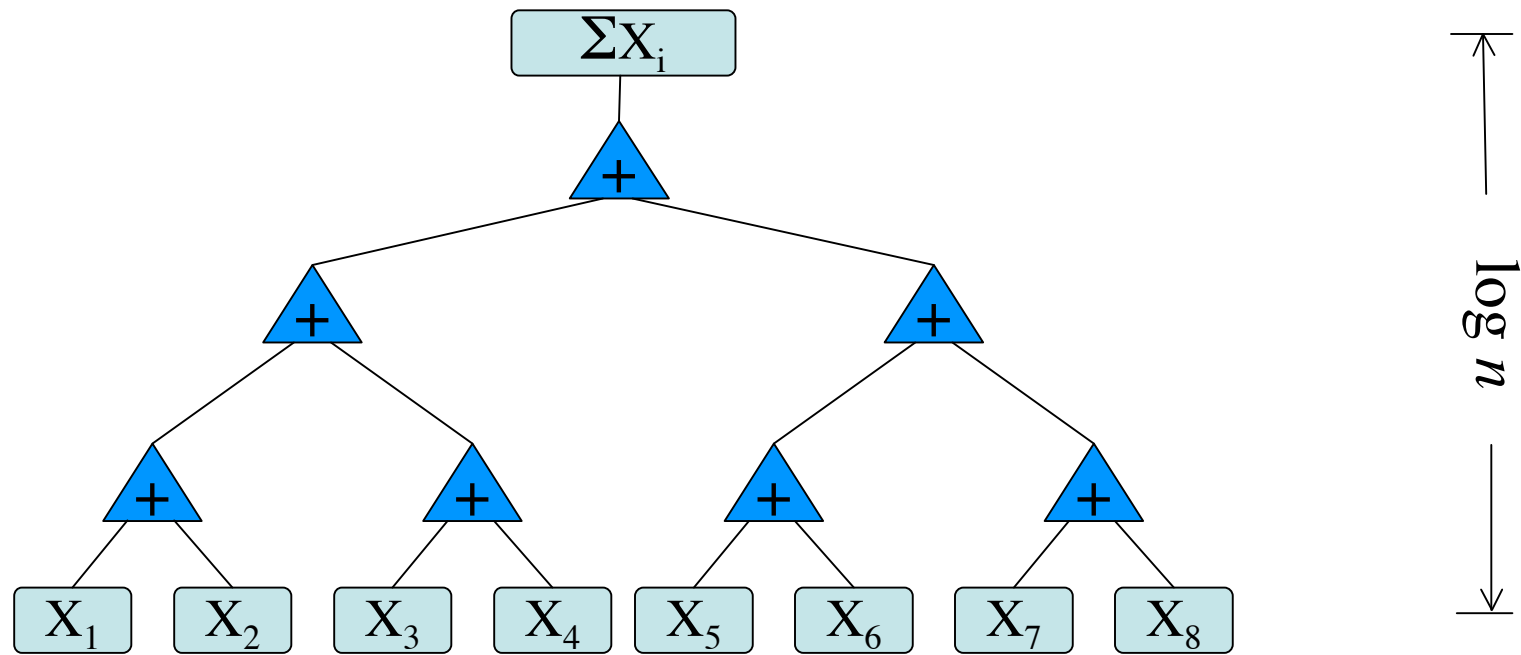
# Boolean Circuit Family



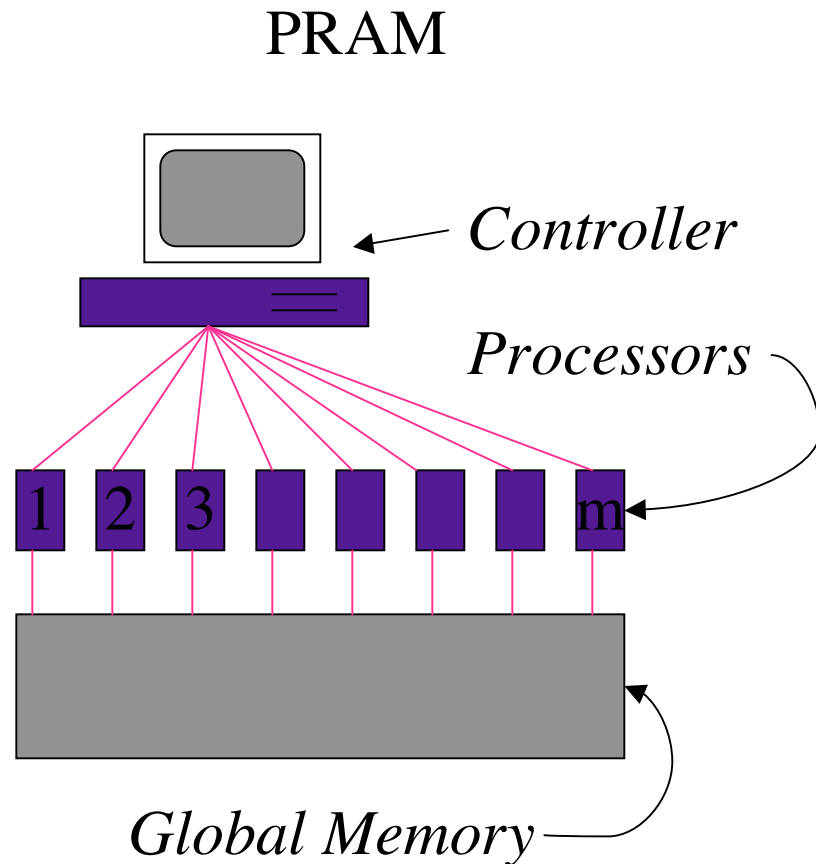
- Gates evaluated one level at a time from input to output with no feedback.
- One hardwired circuit for each problem size.
- Primary resources
  - Depth**=number of levels,  $D_c$
  - Width**=maximum number of gates in a level
  - Work**=total number of gates

# Parallel Computing

Adding  $n$  numbers can be carried out in  $O(\log n)$  steps using  $O(n)$  processors.



# Parallel Random Access Machine



- Each processor runs the same program but has a distinct label

- Each processor communicates with any memory cell in a single time step.

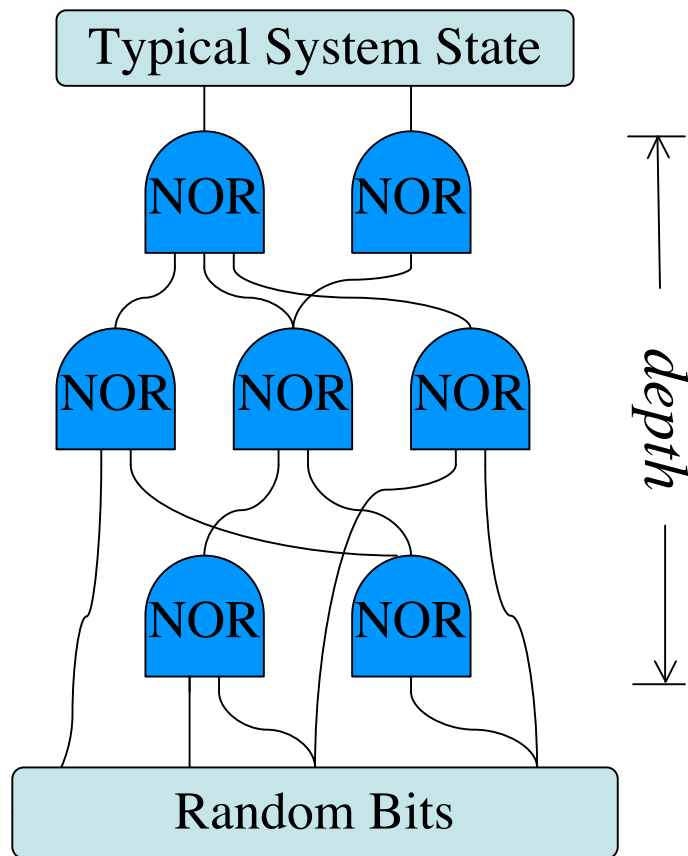
- Primary resources:

- *Parallel time ~ depth*

- *Number of processors ~ width*



# Sampling Complexity

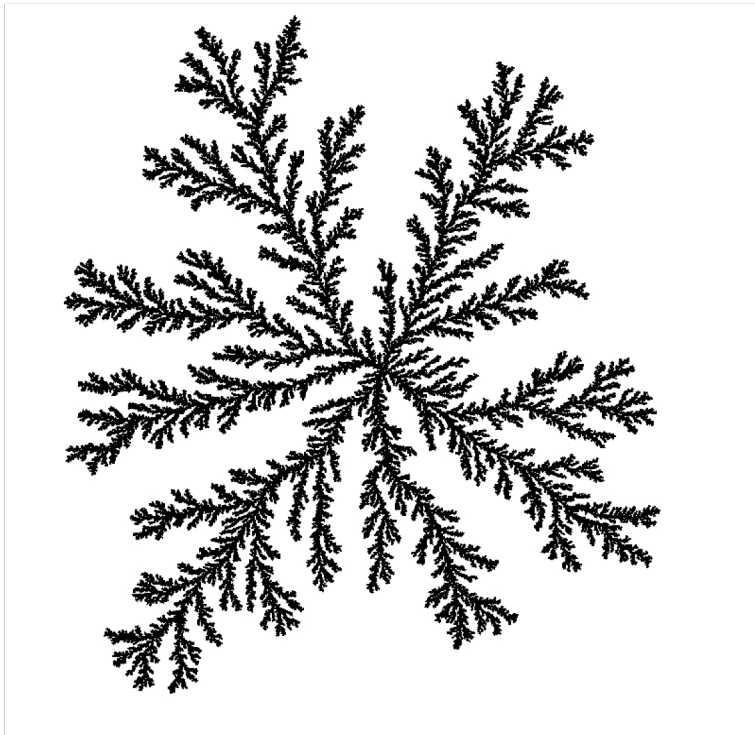


- Monte Carlo simulations convert random bits into descriptions of a typical system states.
- **What is the depth of the shallowest circuit (running time of the fastest PRAM program) that generates typical states?**



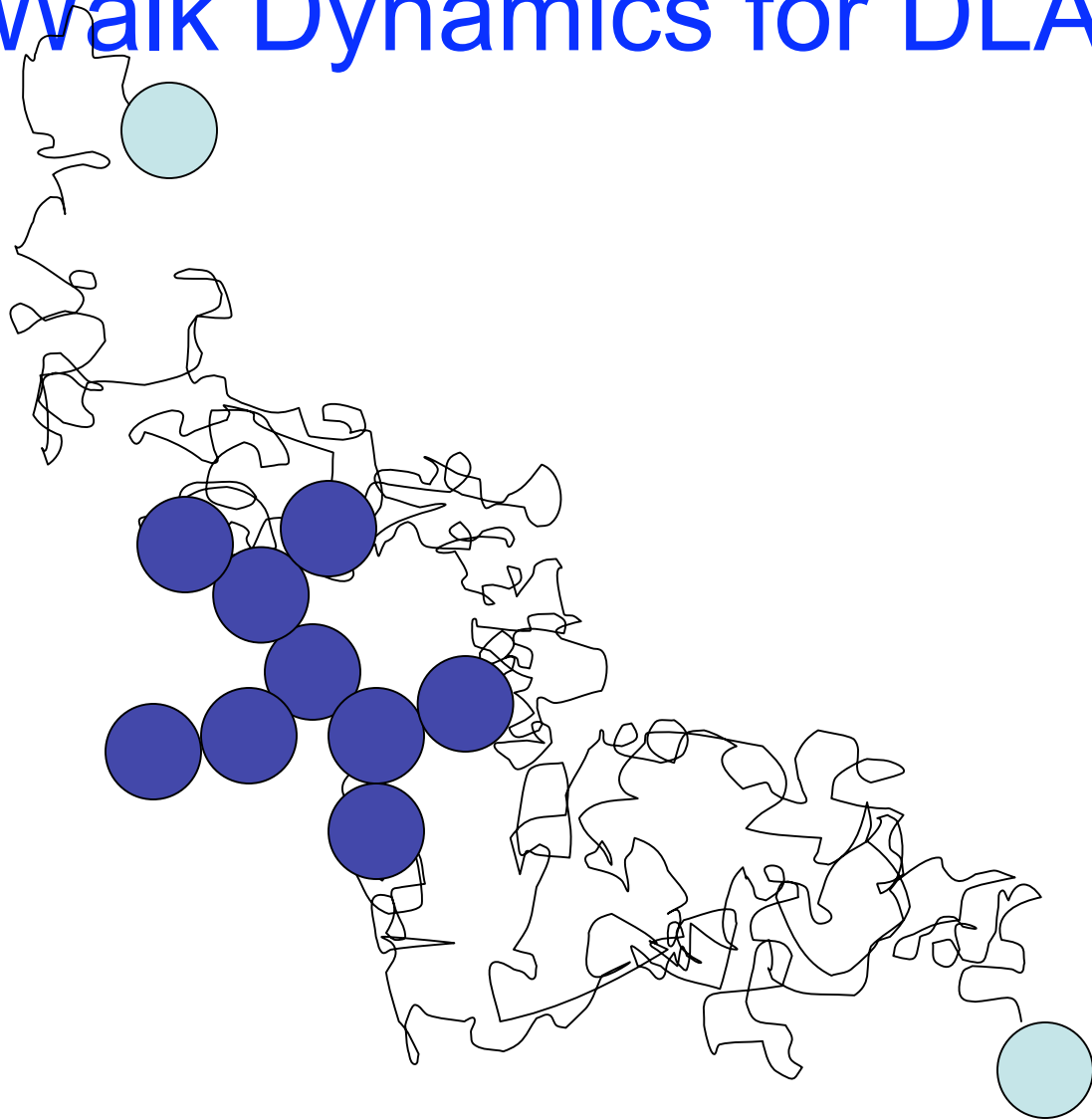
# Diffusion Limited Aggregation

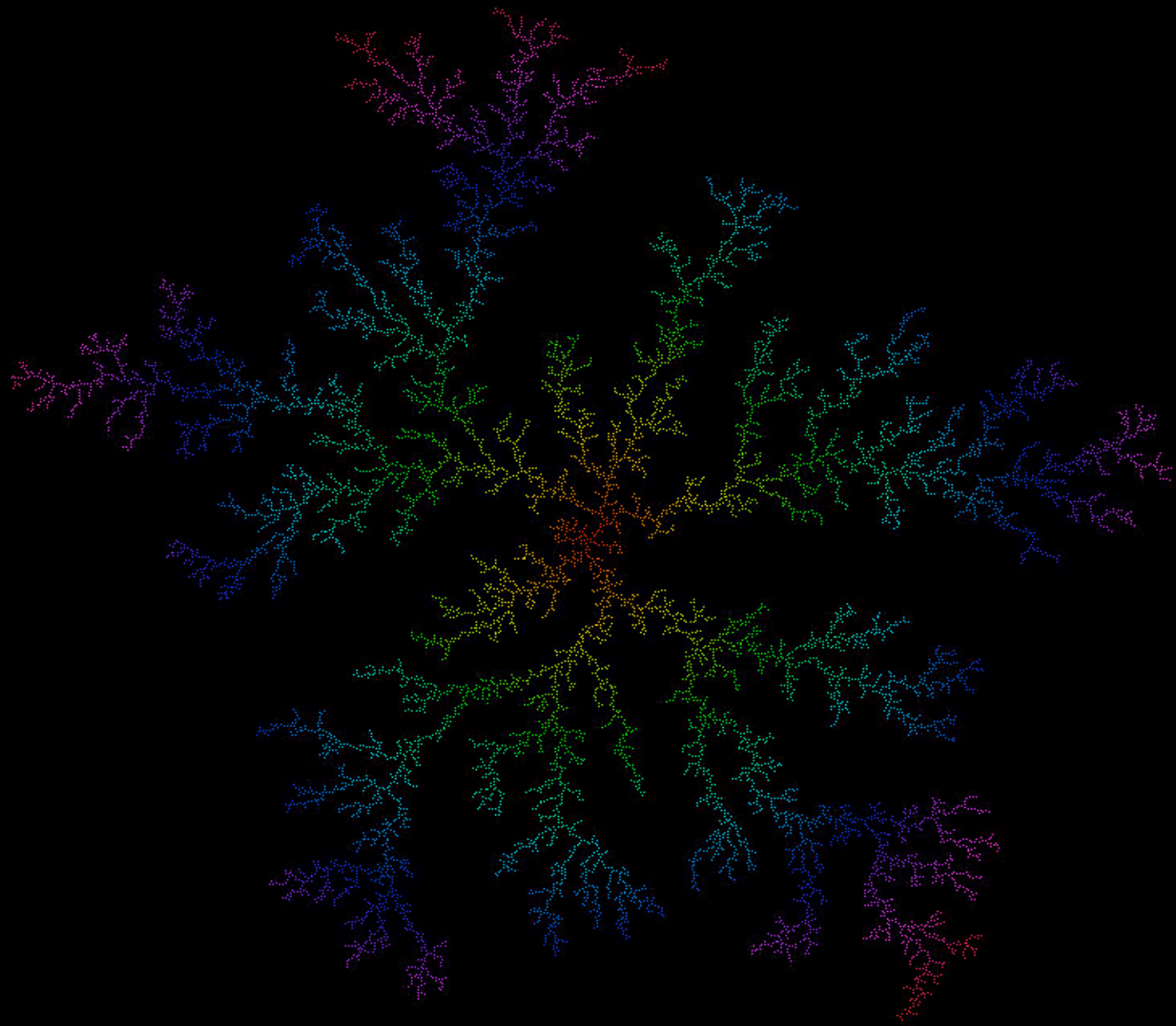
Witten and Sander, PRL 47, 1400 (1981)



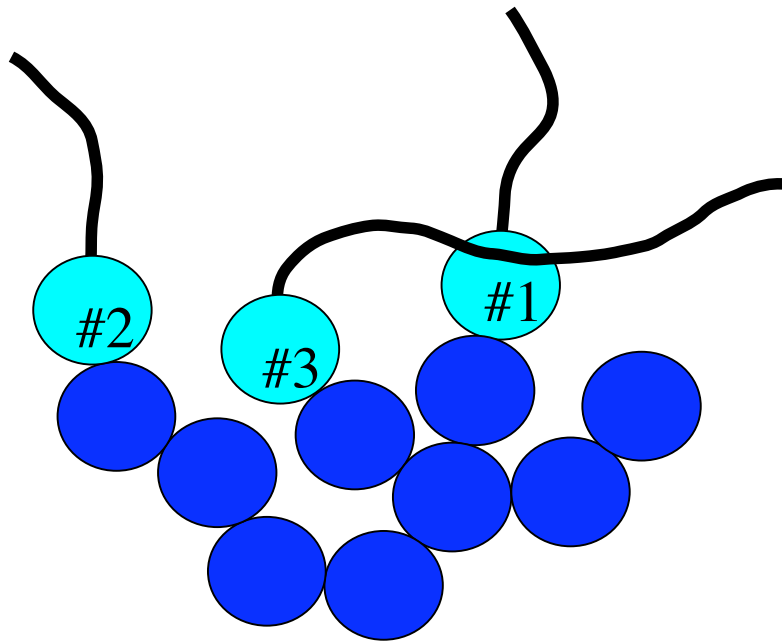
- Particles added *one at a time* with sticking probabilities given by the solution of Laplace's equation.
- Self-organized critical object  
 $d_f = 1.715\dots$  (2D)
- Physical systems:
  - Fluid flow in porous media
  - Electrodeposition
  - Bacterial colonies

# Random Walk Dynamics for DLA

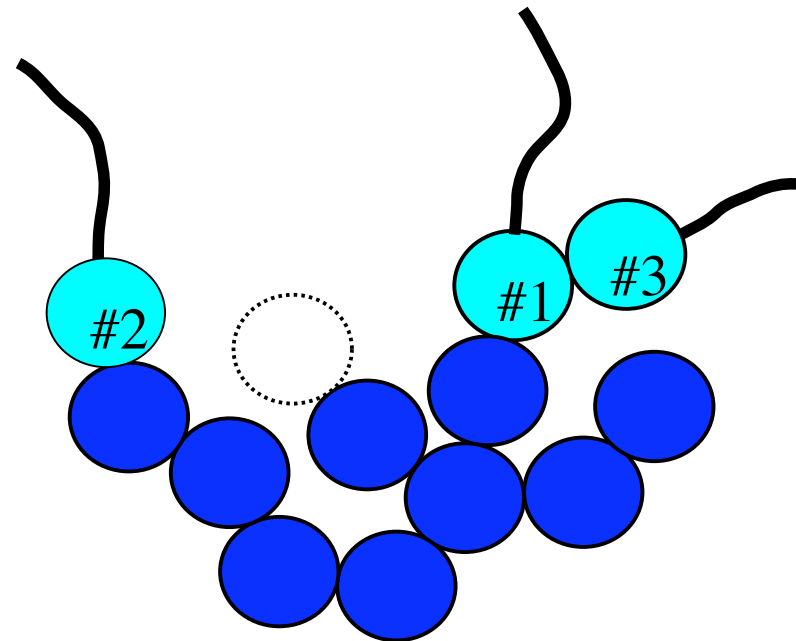




# The Problem with Parallelizing DLA



Parallel dynamics ignores  
*interference* between 1 and 3

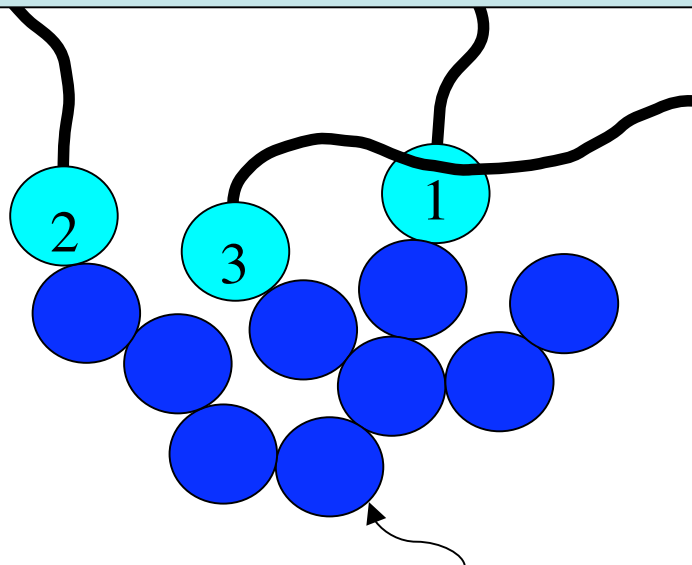


Sequential dynamics

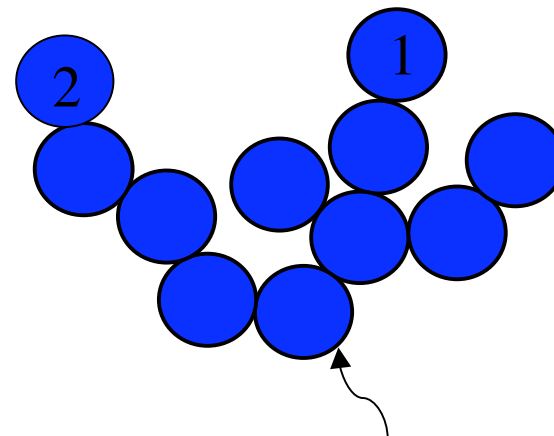
# Parallel Algorithm for DLA

D. Tillberg and JM, PRE **69**, 051403 (2004)

1. Start with seed particle at the origin and  $N$  walk trajectories
2. In parallel move all particles along their trajectories to tentative sticking points on the existing *semi-secure* cluster, which is initially the seed particle at the origin.
3. New semi-secure cluster obtained by removing all particles that interfere with earlier particles.
4. Continue until all particles are correctly placed.



semi-secure cluster, step  $N$

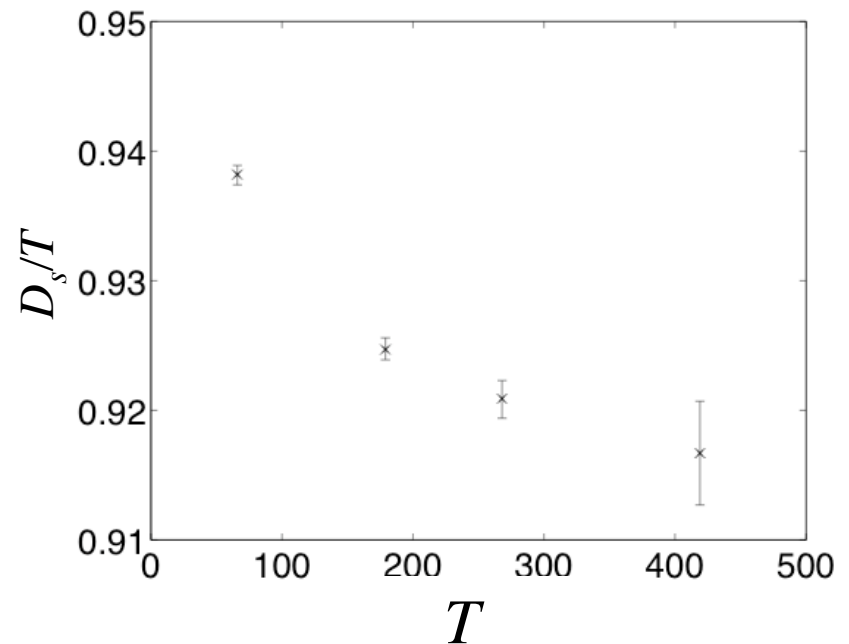
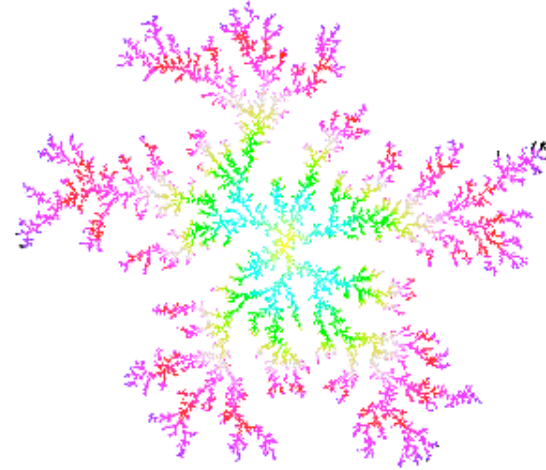


semi-secure cluster, step  $N+1$

# Efficiency of the Algorithm

- DLA is a tree whose structural depth,  $D_s$  scales as the radius of the cluster.
- The running time,  $T$  of the algorithm is asymptotically proportional to the structural depth.

$$T \sim D_s \sim N^{1/d_f}$$



# Depth of DLA

*Theorem:* Determining the shape of an aggregate from the random walks of the constituent particles is a **P**-complete problem.

Proof sketch: Reduce the Circuit Value Problem to DLA dynamics.

*Caveats:*

1. **P**≠**NC** not proven
2. Average case may be easier than worst case
3. Alternative dynamics may be easier than random walk dynamics

*Conjecture:*  $D_c \sim D_s$

# Growing Networks

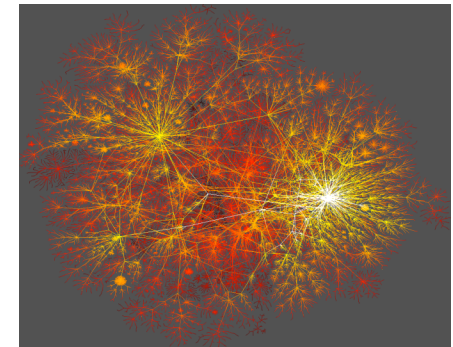
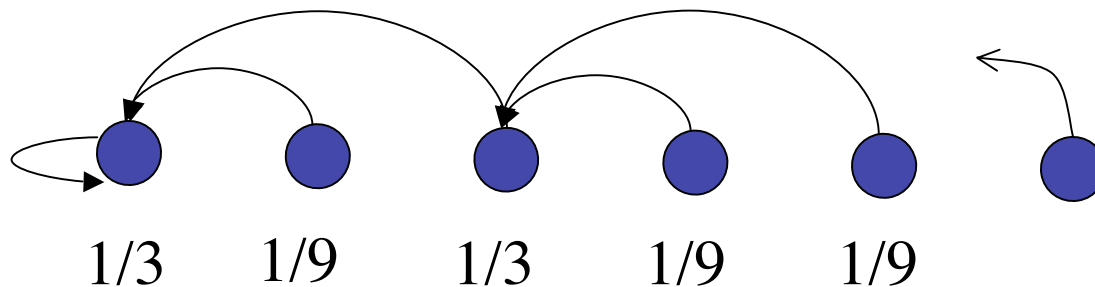
Barabasi and Albert, Science **286**, 509 (1999)

Krapivsky, Redner, Leyvraz, PRL **85**, 4629 (2000)

Add nodes one at a time, connecting new nodes to old nodes according to a “rich get richer” preferential attachment rule:

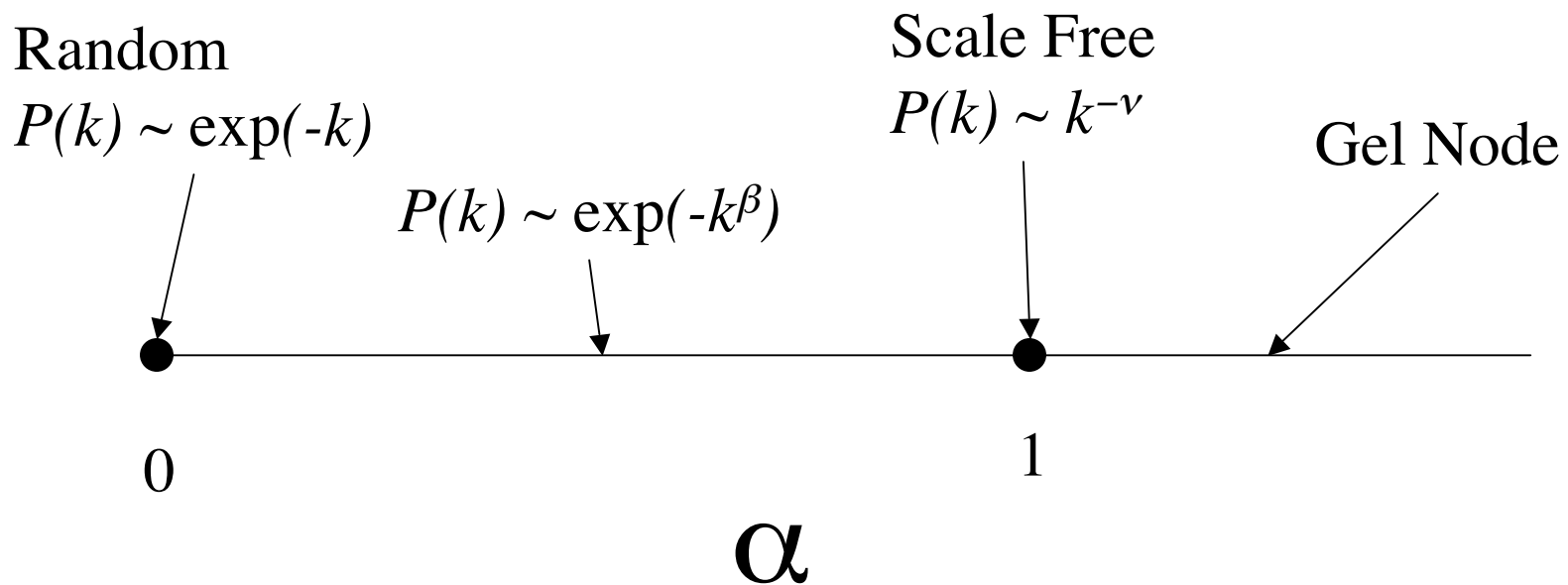
$$\pi_n(t) = \text{Prob}[t \text{ connects to } n] \propto k_n(t)^\alpha$$

where  $k_n(t)$  is the number of connection to node  $n$  at time  $t$ .





# Behavior of Growing Networks



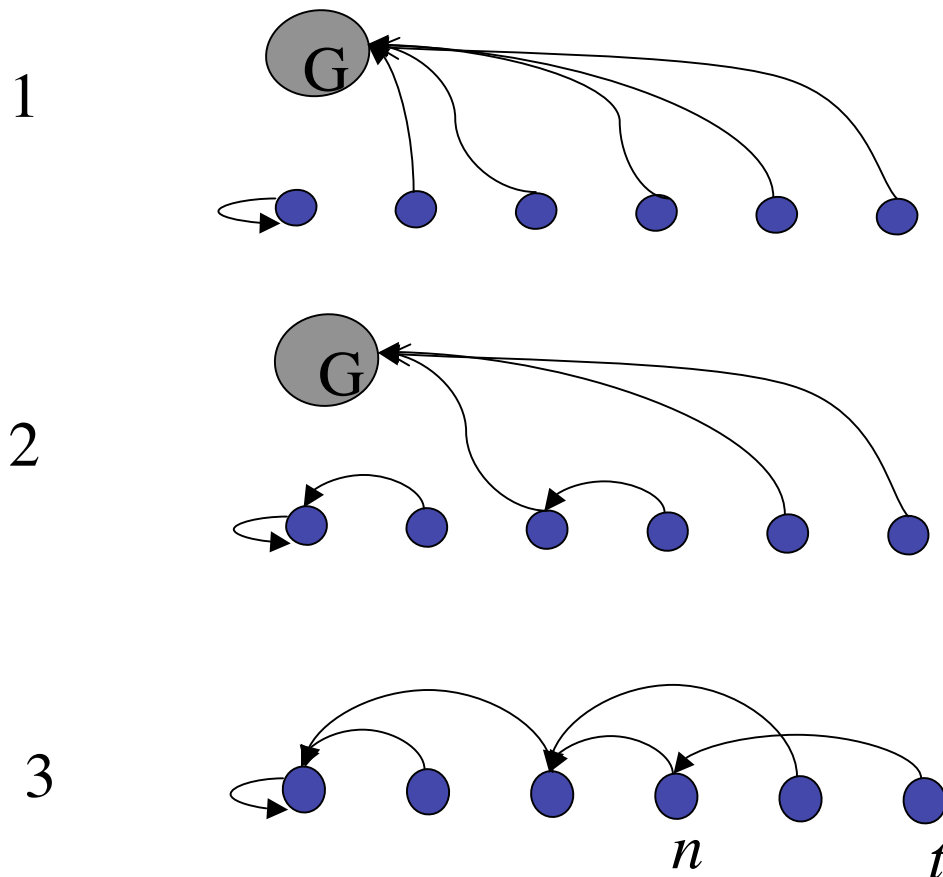
$P(k)$  is the degree distribution

Discontinuous transition at  $\alpha=1$

# Parallel Algorithm for Networks

B. Machta and JM, cond-mat/0408372

For the “high temperature” phase  $0 \leq \alpha \leq 1$



- Initially all nodes are connected to a “ghost node.”
- New connections are made according to bounds on connection probabilities computed from existing node degrees.
- Until all connections are determined.

Bound on step  $S$  of probability of node  $t$  connecting to node  $n$ :

$$p_n^S(t) = k_n^S(t)^\alpha / \tilde{Z}^S(t)$$

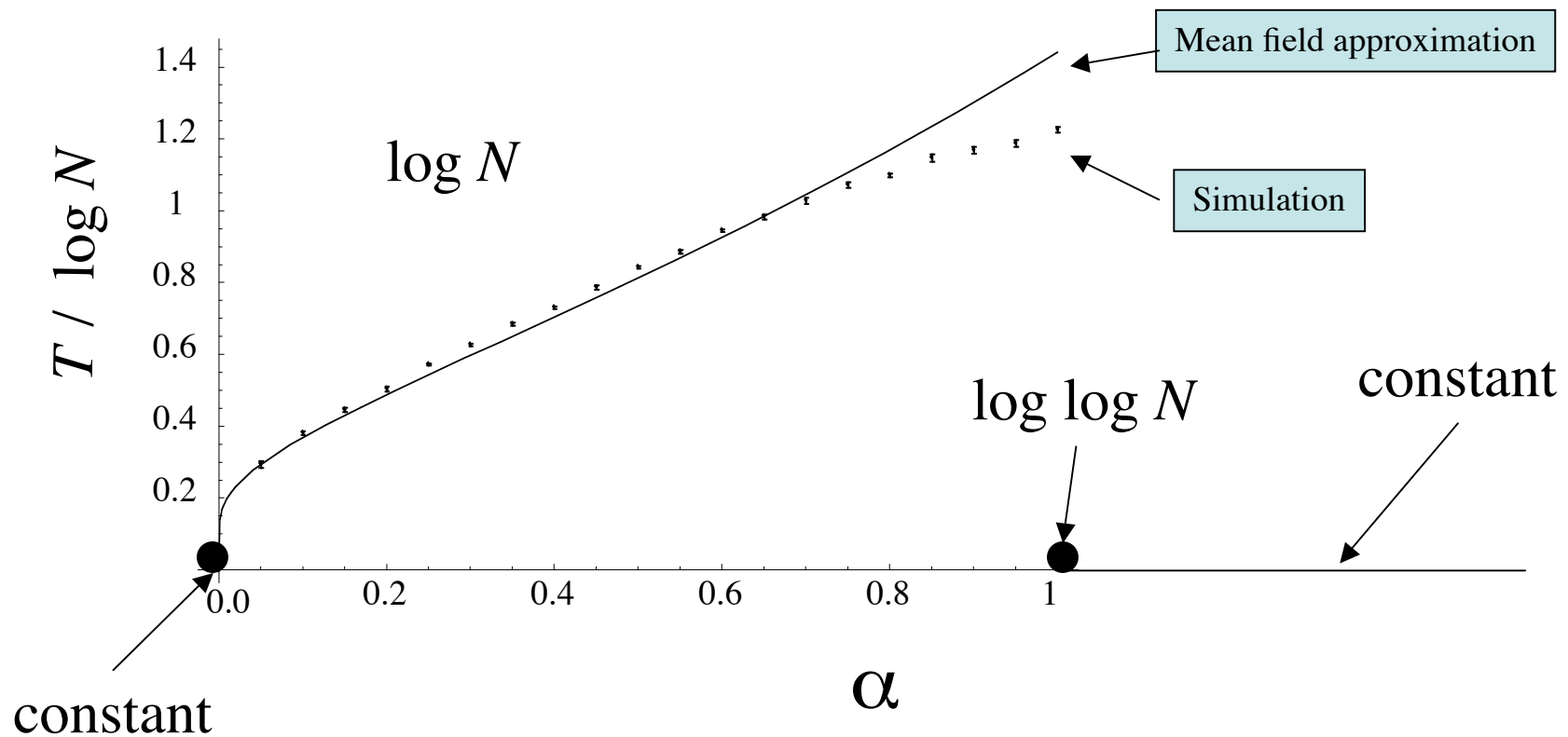
$$\tilde{Z}^S(t) = ck_g^S(t) + \sum_{m=0}^{t-1} k_m^S(t)^\alpha \quad c = 2^\alpha - 1$$

$$p_n^1(t) \leq p_n^2(t) \leq \dots \leq p_n^T(t) = \pi_n(t)$$

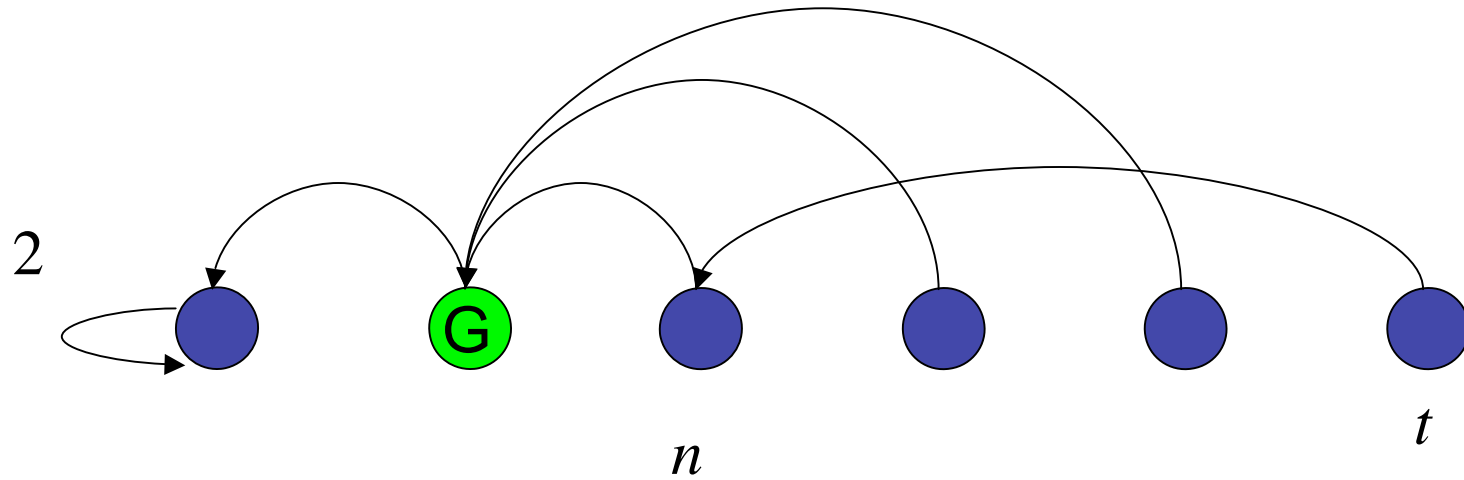
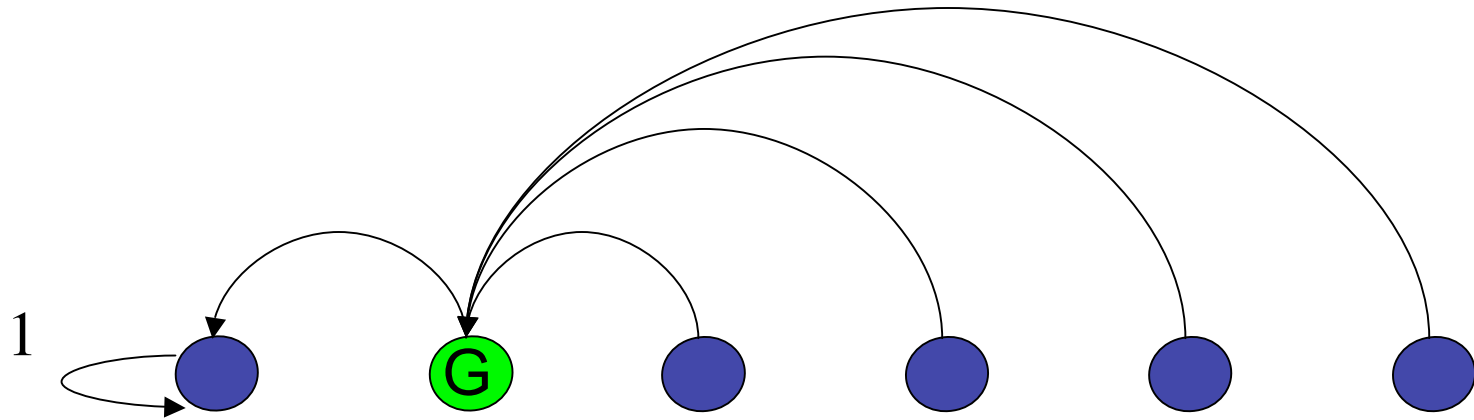
Conditional probability of node  $t$  connecting to node  $n$  on step  $S$  given it was previously connected to  $g$ :

$$\rho_n^S(t) = \frac{p_n^S(t) - p_n^{S-1}(t)}{1 - \sum_{m=0}^{t-1} p_m^{S-1}(t)}$$

# Depth of Growing Networks

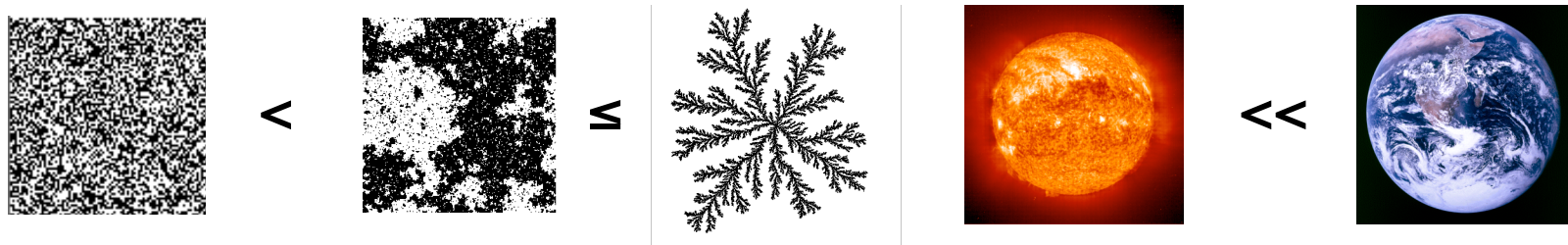


For the “low temperature” phase  $\alpha > 1$



# What is Physical Complexity?

- “I shall not today attempt further to define the kinds of material I understand to be embraced with that shorthand description. ... But I know it when I see it.”
  - Justice Potter Stewart on pornography

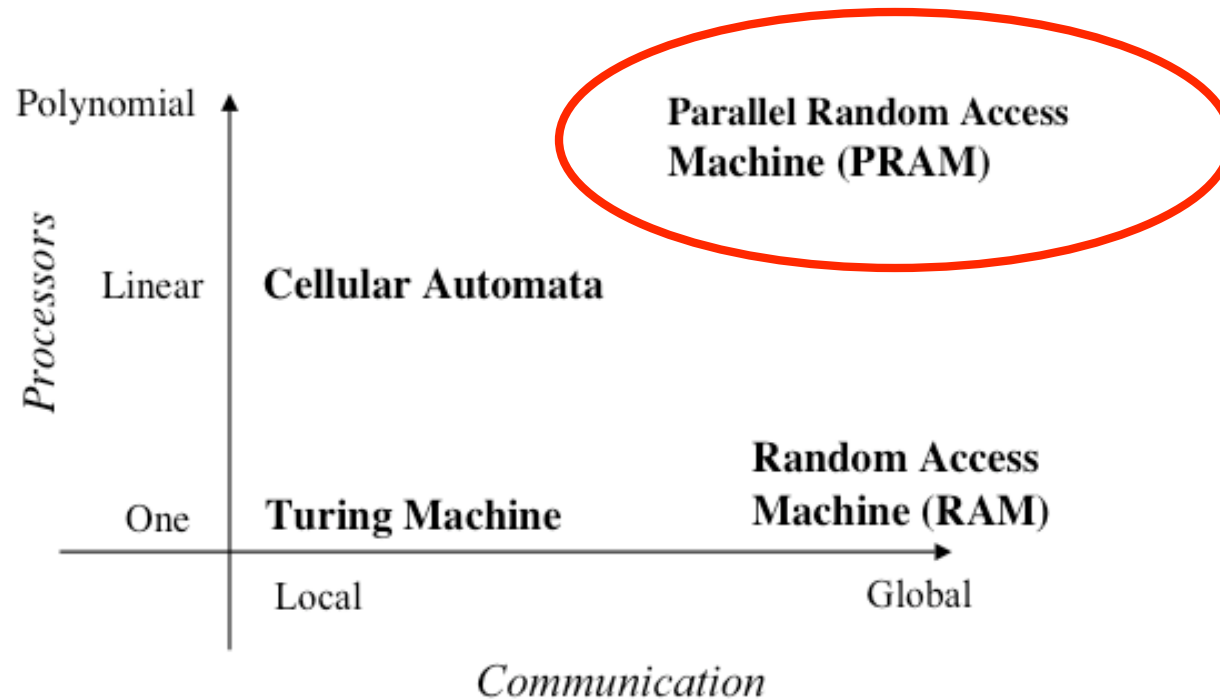


# History and Complexity

–Charles Bennett

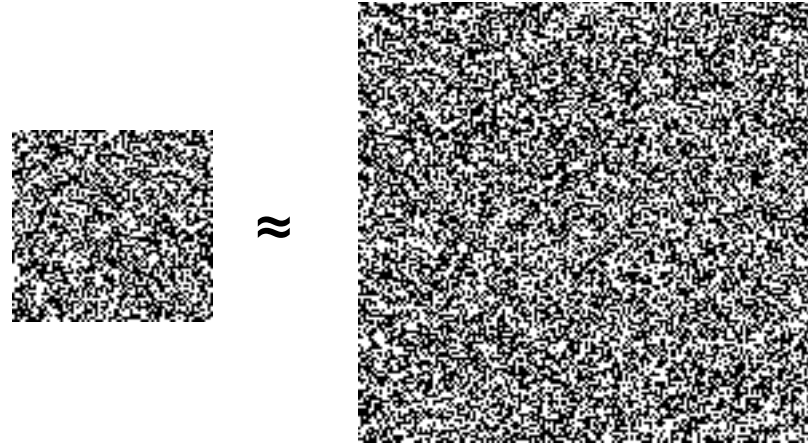
- The emergence of a complex system from simple initial conditions requires a long history.
- History can be quantified in terms of the computational complexity of simulating states of the system.

# What Model of Computation?





# Discount Hardware



Complexity of a system composed of nearly independent subsystem is given by the most complex subsystem.

# Discount Communication



Complexity emerges from interactions, not from signal propagation.

# Choose the Fastest Algorithm

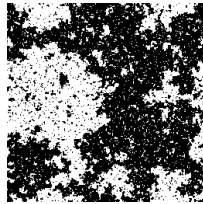


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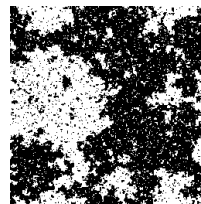


1 sec

1 yr



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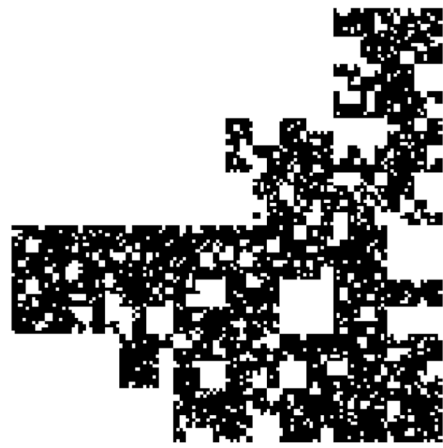
Swendsen-Wang    Metropolis

# Depth of Physical Systems

**The depth of a physical system is the depth of a Boolean circuit (or parallel time on a PRAM) to simulate a typical system state using the most efficient algorithm.**

# Hierarchy of Depth

constant

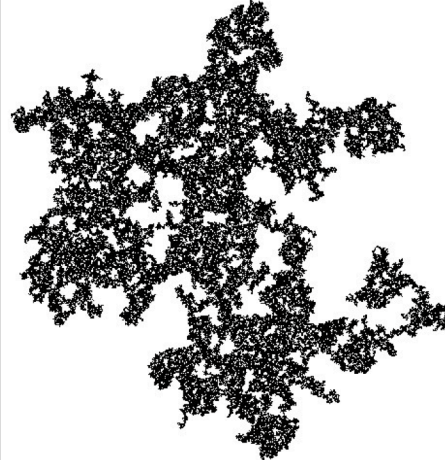


**Mandelbrot percolation**

Growing network  $\alpha > 1$

$T > T_c$  Ising

polylog



**Invasion percolation**

Growing network  $\alpha \leq 1$

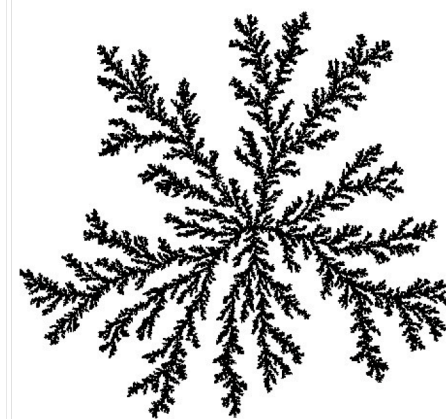
Eden growth

Ballistic deposition

Bak-Sneppen model

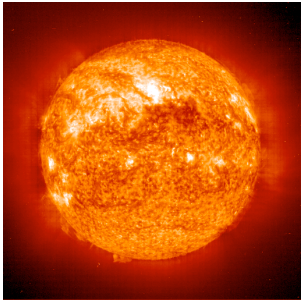
Internal DLA

polynomial



**DLA**

$T = T_c$  Ising



# Conclusions



- Computational complexity theory provides interesting perspectives on physical systems.
- DLA has power law depth.
- Growing networks display a complexity transition from logarithmic to constant depth at  $\alpha=1$ .
- Depth, defined as the minimum number of parallel steps needed to simulate a system, is correlated with intuitive notions of physical complexity.

