

# Phase Transitions in Physics and Computer Science

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# Magnetism

- When cold enough, Iron will stay magnetized, and even magnetize spontaneously
- But above a critical temperature, it suddenly ceases to be magnetic
- Interactions between atoms remain the same, but global behavior changes!
- Like water freezing, outbreaks becoming epidemics, opinions changing...

# The Ising model

- Lattice (e.g. square) with  $n$  sites
- Each has a “spin”  $s_i = \pm 1$ , “up” or “down”
- Energy is a sum over neighboring pairs:

$$E = - \sum_{ij} s_i s_j$$

- Lowest energy: all up or all down
- Highest energy: checkerboard

# Boltzmann Distribution

- At thermodynamic equilibrium, temperature  $T$
- Higher-energy states are less likely:

$$P(s) \sim e^{-E(s)/T}$$

- When  $T \rightarrow 0$ , only lowest energies appear
- When  $T \rightarrow \infty$ , all states are equally likely

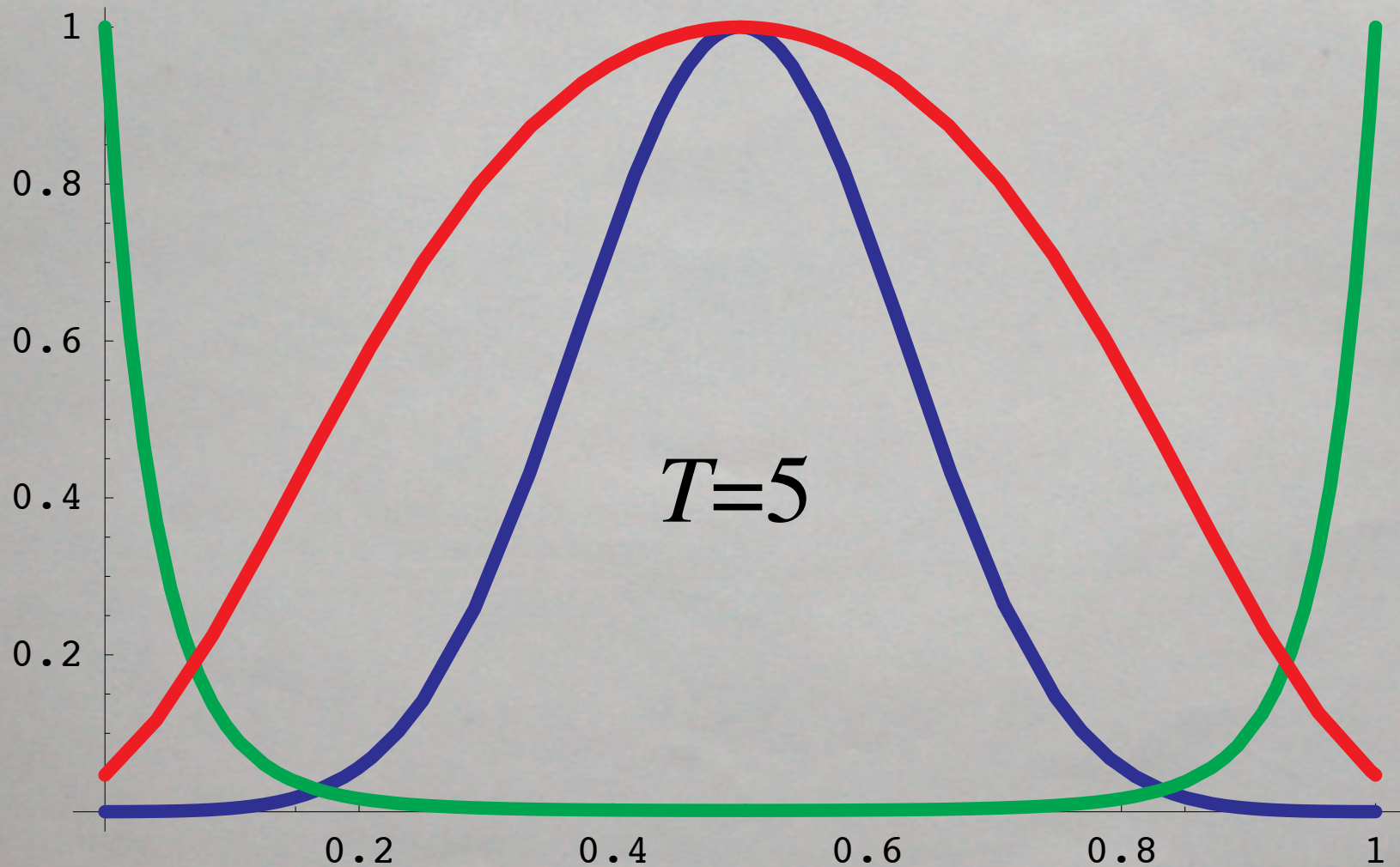
# What Happens

- Below critical temperature, the system “magnetizes”: mostly up or mostly down
- Small islands of the minority state; as  $T$  increases, these islands grow
- Above critical temperature, islands=sea; at large scales, equal numbers of up and down
- When  $T=T_c$ , islands of all scales: system is scale-invariant!

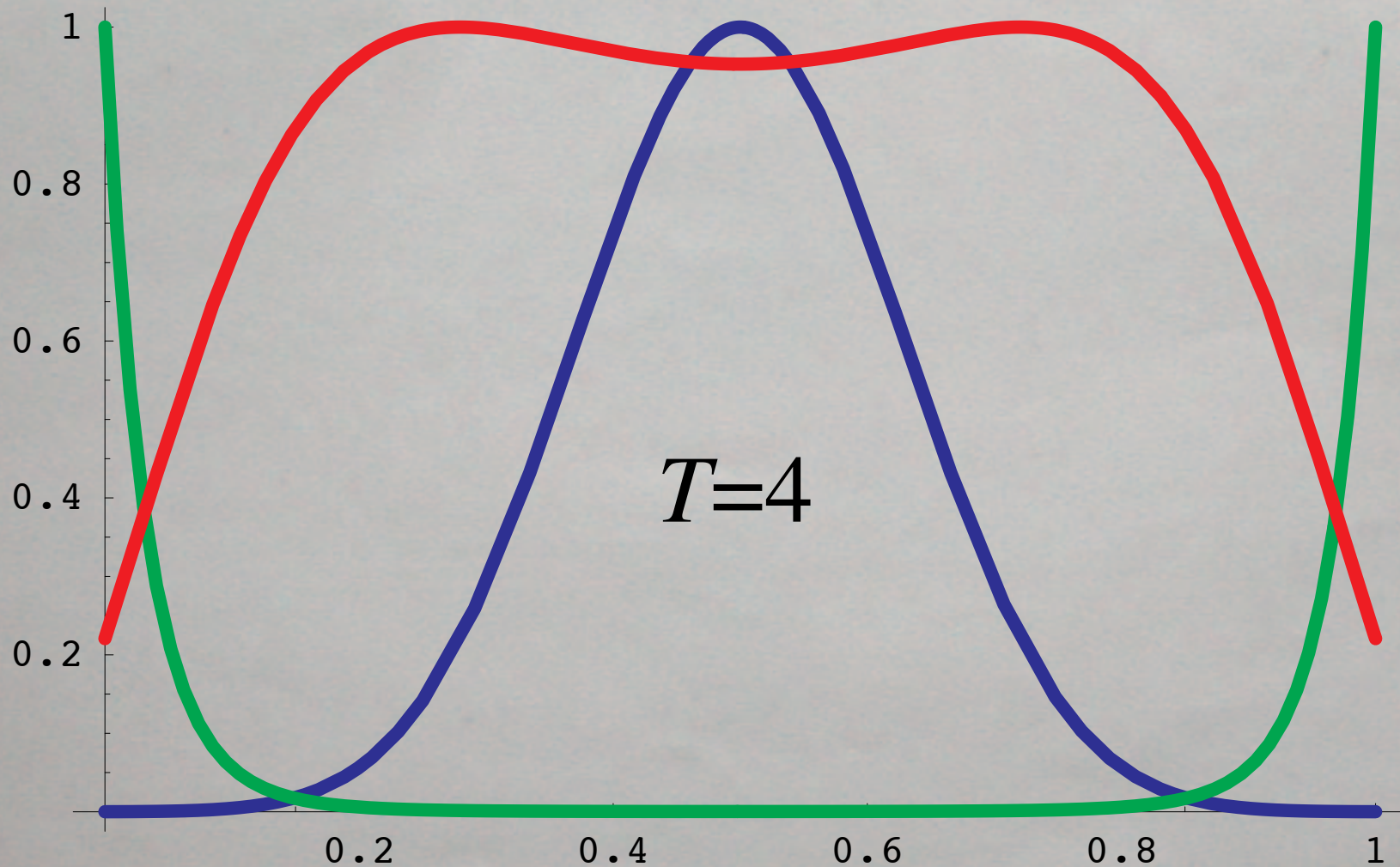
# Mean Field

- Ignore topology: forget lattice structure
- If  $a$  of the sites are up and  $1-a$  are down, energy is  $E = 2n^2 (2a(1-a) - a^2 - (1-a)^2)$
- At any  $T$ , most-likely states have  $a=0$  or  $a=1$
- But the number of such states is  $\binom{n}{an}$ , which is tightly peaked around  $a=1/2$ .
- Total probability( $a$ ) = #states( $a$ ) Boltzmann( $a$ )

# Energy vs. Entropy

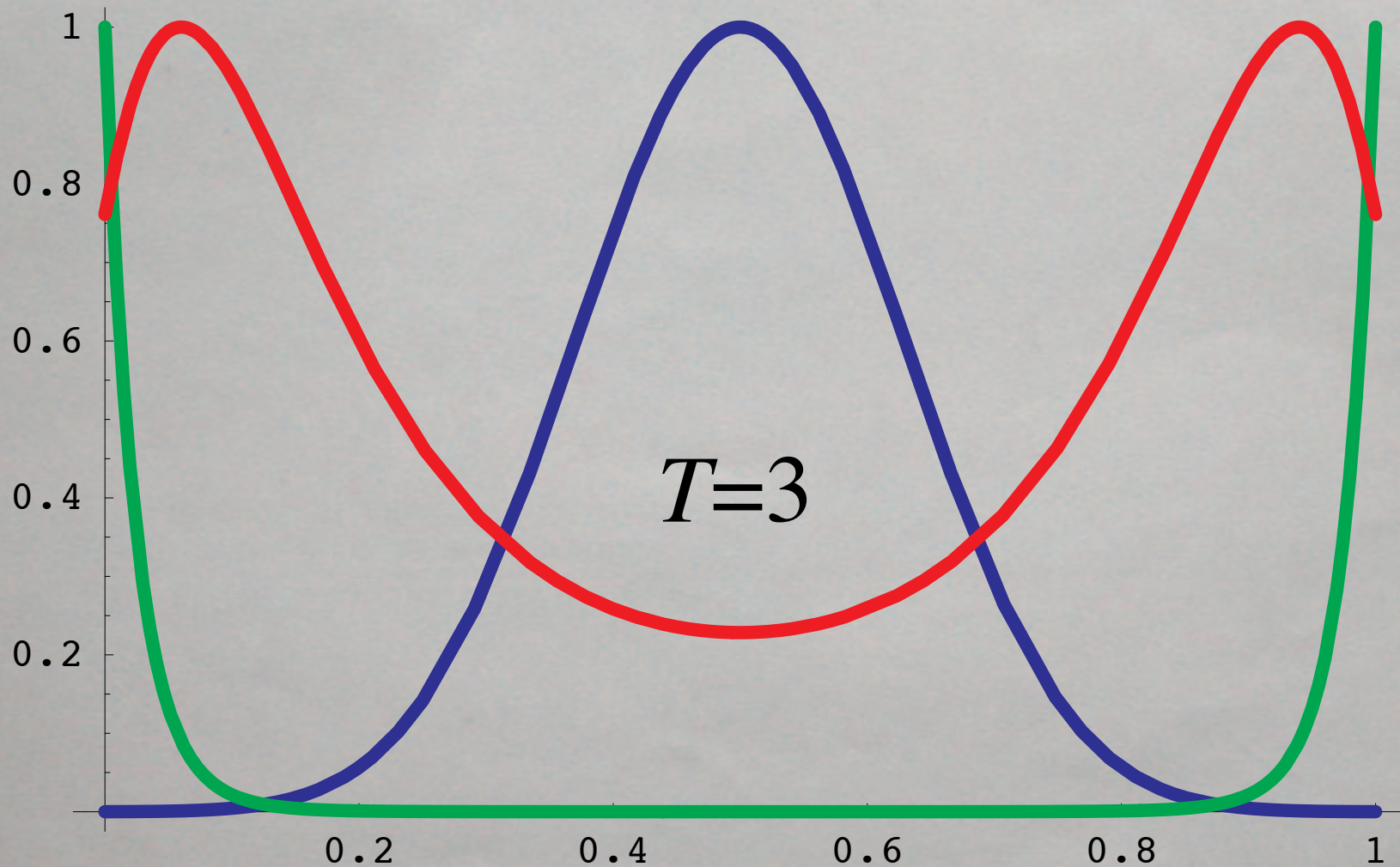


# Energy vs. Entropy





# Energy vs. Entropy



# Correlations

- $C(r)$  = correlation between two sites  $r$  apart
- If  $T > T_c$ , correlations decay exponentially:

$$C(r) \sim e^{-r/\ell}$$

- Correlation length  $\ell$  decreases as  $T$  grows
- As we approach  $T_c$ , correlation length diverges
- At  $T_c$ , power-law correlations (scale-free):

$$C(r) \sim \ell^{-\alpha}$$

# Percolation

- Fill a fraction  $p$  of the sites in a lattice
- When  $p < p_c$ , small islands, whose size is exponentially distributed:

$$P(s) \sim e^{-s/\bar{s}}$$

- When  $p > p_c$ , “giant cluster” appears
- At  $p_c$ , power-law distribution of cluster sizes:

$$P(s) \sim s^{-\alpha}$$

# The Adversary

...designs problems that are as diabolically hard as possible, forcing us to solve them in the worst case. (Hated and feared by computer scientists.)





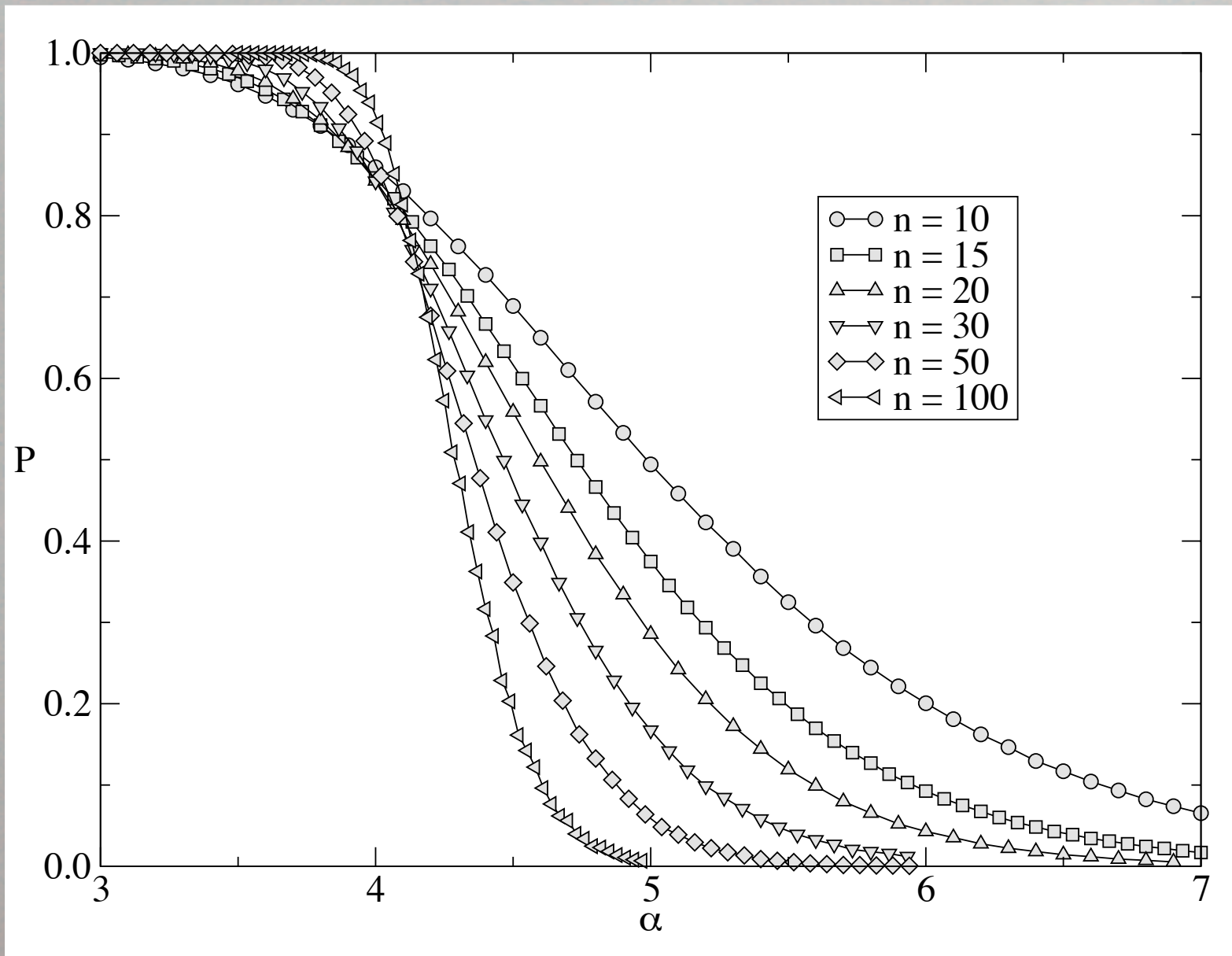
# La Dame Nature

...asks questions whose answers  
are simpler and more beautiful  
than we have any right to  
imagine. (Worshipped by  
physicists.)

# Random NP Problems

- A 3-SAT formula with  $n$  variables,  $m$  clauses
- Choose each clause randomly:  $\binom{n}{3}$  possible triplets, negate each one with probability  $1/2$
- Precedents:
  - Random Graphs (Erdős-Rényi)
  - Statistical Physics: ensembles of disordered systems, e.g. spin glasses
- Sparse Case:  $m = \alpha n$  for some density  $\alpha$

# A Phase Transition



# The Threshold Conjecture

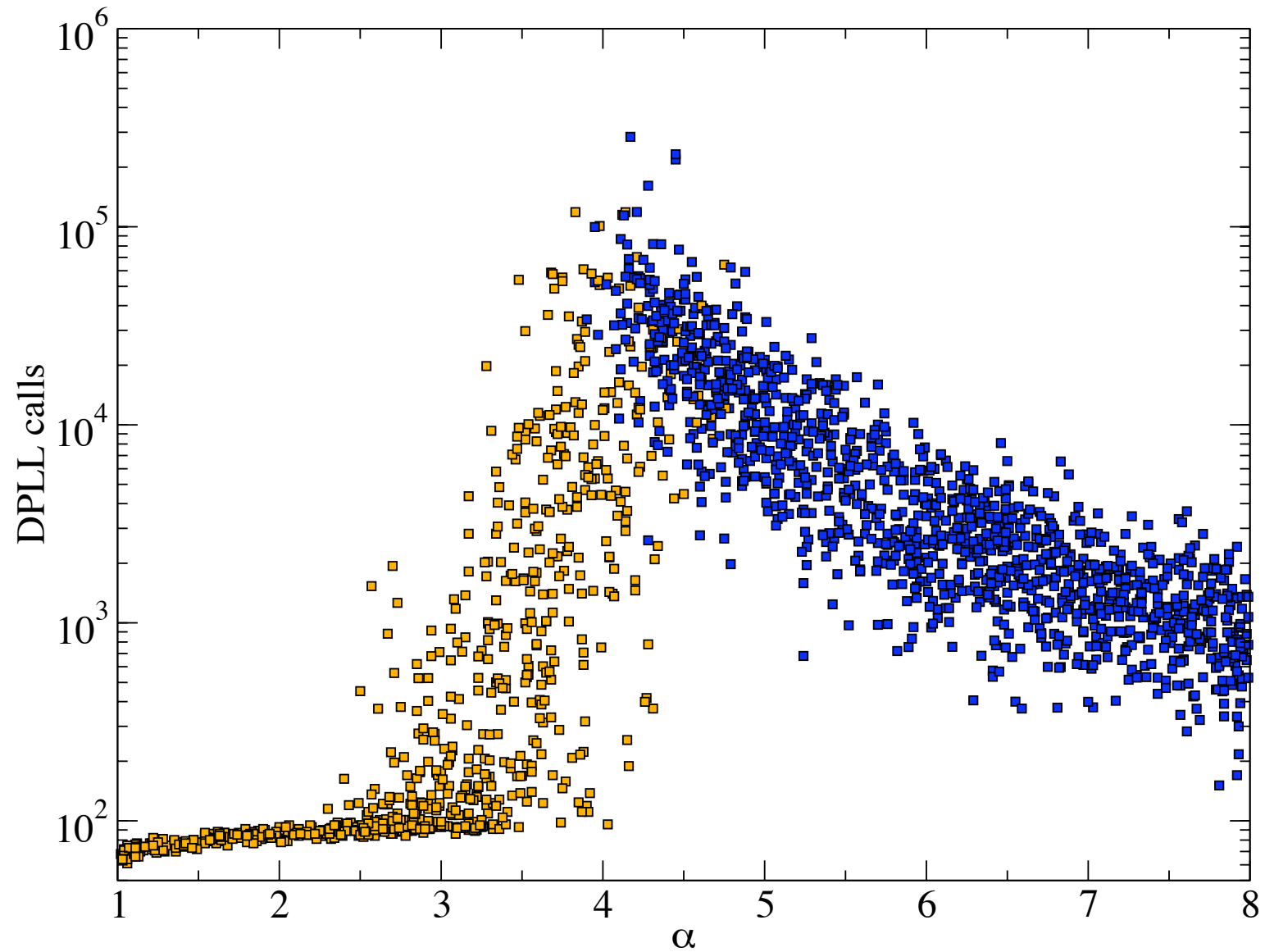
- We believe that for each  $k \geq 3$ , there is a critical clause density  $\alpha_k$  such that

$$\lim_{n \rightarrow \infty} \Pr [F_k(n, m = \alpha n) \text{ is satisfiable}] = \begin{cases} 1 & \text{if } \alpha < \alpha_k \\ 0 & \text{if } \alpha > \alpha_k \end{cases}$$

- So far, only known rigorously for  $k = 2$



# Search Times



# An Upper Bound

- The *average* number of solutions  $E[X]$  is

$$2^n \left(\frac{7}{8}\right)^m = \left(2 \left(\frac{7}{8}\right)^\alpha\right)^n$$

- This is exponentially small whenever

$$\alpha > \log_{8/7} 2 \approx 5.19$$

- But the transition is much lower, at  $\alpha \approx 4.27$ .  
What's going on?

# A Heavy Tail

- In the range  $4.27 < \alpha < 5.19$ , the average number of solutions is exponentially large.
- Occasionally, there are exponentially many...
- ...but most of the time there are none!
- A classic “heavy-tailed” distribution
- Large average doesn't prove satisfiability!

# Lower Bound #1

- Idea: track the progress of a simple algorithm!
- When we set variables, clauses disappear or get shorter:

$$\bar{x} \wedge (x \vee y \vee z) \Rightarrow (y \vee z)$$

- *Unit Clauses* propagate:

$$x \wedge (\bar{x} \vee y) \Rightarrow y$$

# One Path Through the Tree

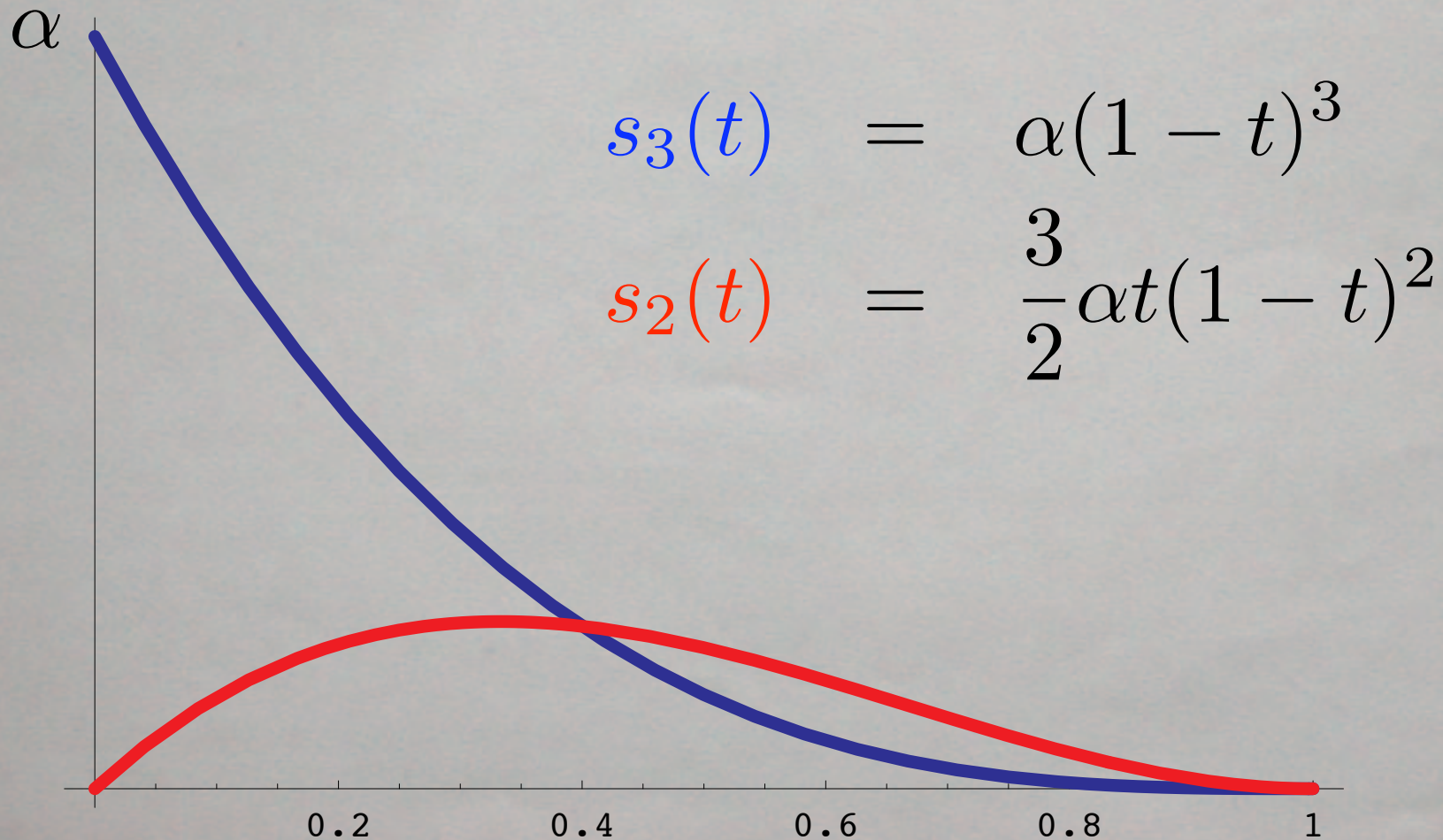
- If there is a unit clause, satisfy it.  
Otherwise, choose a random variable and give it a random value!
- The remaining formula is random for all  $t$ :

$$\frac{ds_3}{dt} = -\frac{3s_3}{1-t}, \quad \frac{ds_2}{dt} = \frac{(3/2)s_3 - 2s_2}{1-t}$$

$$s_3(0) = \alpha, \quad s_2(0) = 0$$

# One Path Through the Tree

- These differential equations give



# Branching Unit Clauses

- Each unit clause has on average  $\lambda$  children, where

$$\lambda = \frac{1}{2} \frac{2s_2}{1-t} = \frac{3}{4} \alpha t (1-t)$$

- When  $\lambda > 1$ , they proliferate and contradictions appear
- Maximized at  $t = 1/2$
- But if  $\alpha < 8/3$ , then  $\lambda < 1$  always, and the unit clauses stay manageable.

# Constructive Methods Fail

- Fancier algorithms, harder math:  $\alpha < 3.52$ .
- But, for larger  $k$ , algorithmic methods are nowhere near the upper bound for  $k$ -SAT:

$$O\left(\frac{2^k}{k}\right) < \alpha < O(2^k)$$

- To close this gap, we need to resort to non-constructive methods.



# Lower Bound #2

- Idea: bound the *variance* of the number of solutions.
- If  $X$  is a nonnegative random variable,

$$\Pr[X > 0] \geq \frac{E[X]^2}{E[X^2]}$$

- $E[X]$  is easy;  $E[X^2]$  requires us to understand *correlations* between solutions.

# Correlations

- The second moment  $E[X^2]$  is the expected number of *pairs* of satisfying assignments.
- If two assignments have *overlap*  $z$ , they satisfy a random  $k$ -SAT clause with probability

$$q(z) = 1 - 2 \cdot 2^{-k} + z^k 2^{-k}$$

- Note that

$$q(1/2) = (1 - 2^{-k})^2$$

as if the pair were independent.

# Correlations

- Now  $E[X^2]$  is the number of pairs with overlap  $z$ , times the probability each pair is satisfying, summed over  $z$ :

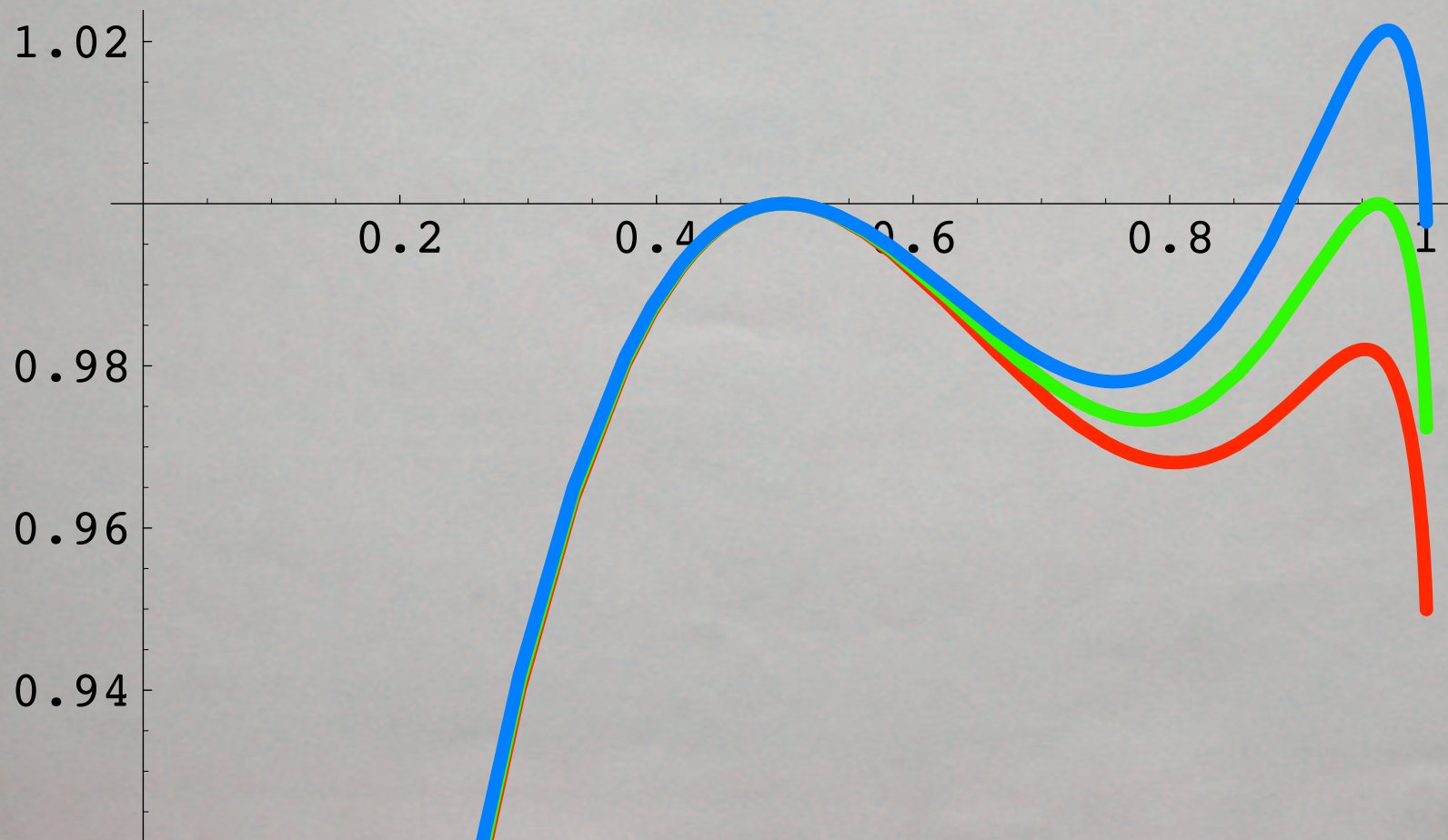
$$E[X^2] \approx \sum_z 2^n \binom{n}{zn} q(z)^{\alpha n} dz$$
$$\approx 2^n \int_0^1 e^{n(h(z) + \alpha \ln q(z))} dz$$

where  $h(z) = -z \ln z - (1 - z) \ln(1 - z)$

- Again, a tradeoff between entropy and “energy.”

# A Function of Distance

- When the expected number of *pairs* of solutions is peaked at  $1/2$ , most pairs are “independent” and the variance is small.



# Determining the Threshold

- A series of results has narrowed the range for the transition in  $k$ -SAT to

$$2^k \ln 2 - O(k) < \alpha < 2^k \ln 2 - O(1)$$

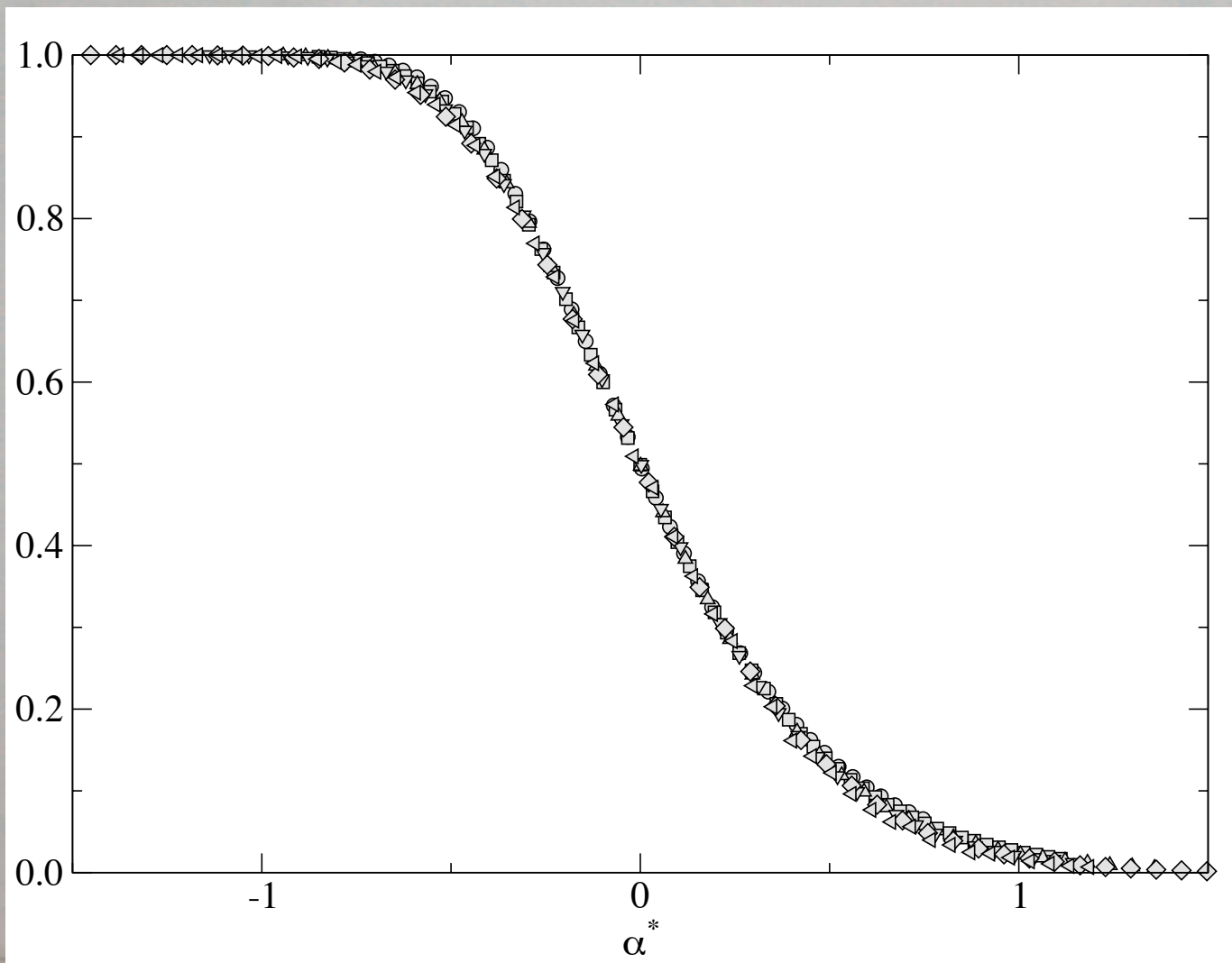
- Prediction from statistical physics:

$$2^k \ln 2 - O(1)$$

- Seems difficult to prove with current methods.

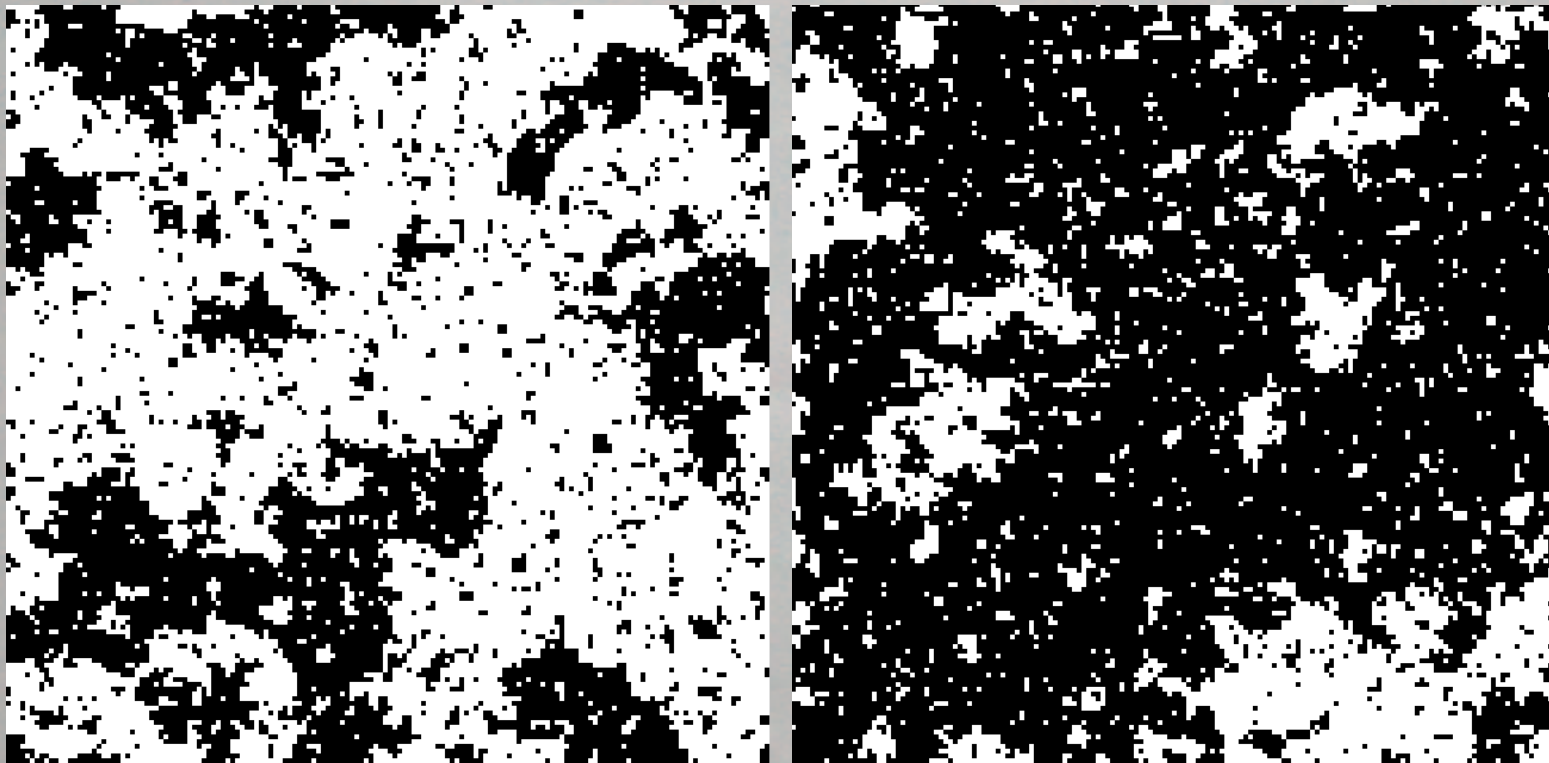
# Scaling and Universality

- Rescaling  $\alpha$  around the critical point causes different  $n$  to coincide. A universal function?



# Clustering

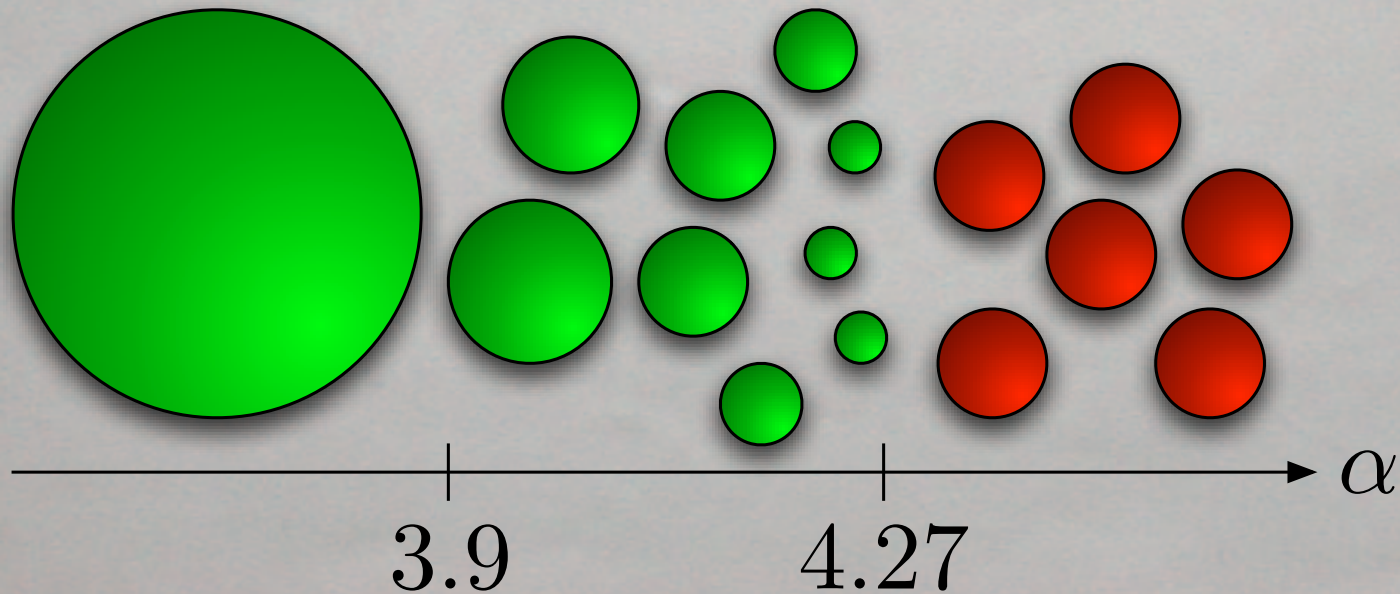
- Below the critical temperature, magnets have two *macrostates* (Gibbs measures)



- Glasses, and 3-SAT, have exponentially many!

# Clustering

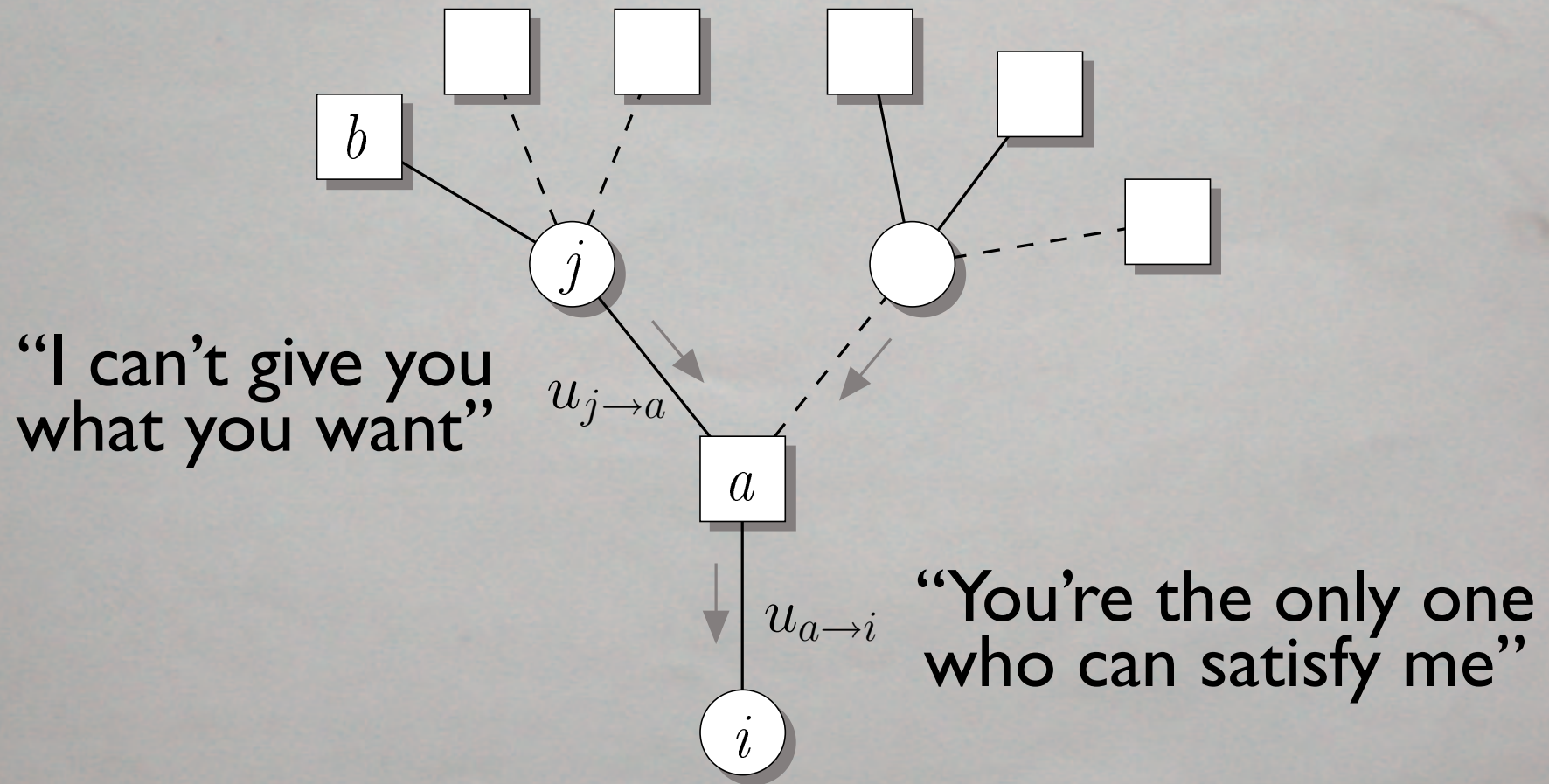
- An idea from statistical physics: there is another transition, from a unified “cloud” of solutions to separate clusters.
- Is this why algorithms fail at  $\alpha \sim 2^k / k$ ?





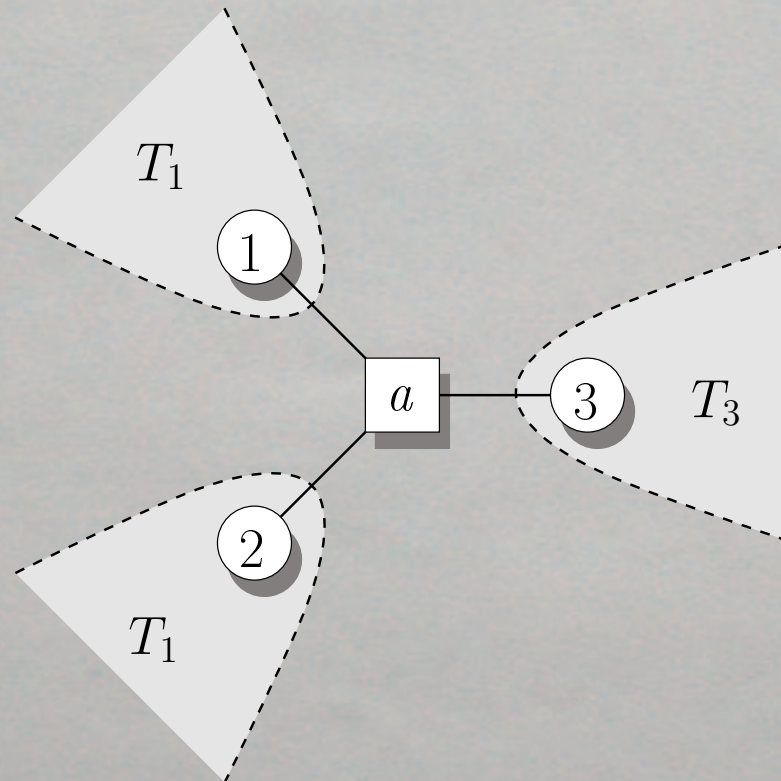
# The Physicists' Algorithm

- A “message-passing” algorithm:



# Why Does It Work?

- Random formulas are locally treelike.
- Assume the neighbors are independent:



- *Proving* this will take some very deep work.

# SHAMELESS PLUGS

*The Nature of Computation*

Computational  
Complexity and  
Statistical Physics

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A VOLUME IN THE  
SANTA FE INSTITUTE STUDIES IN THE SCIENCES OF COMPLEXITY



Mertens and Moore

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