

Random 3-XORSAT

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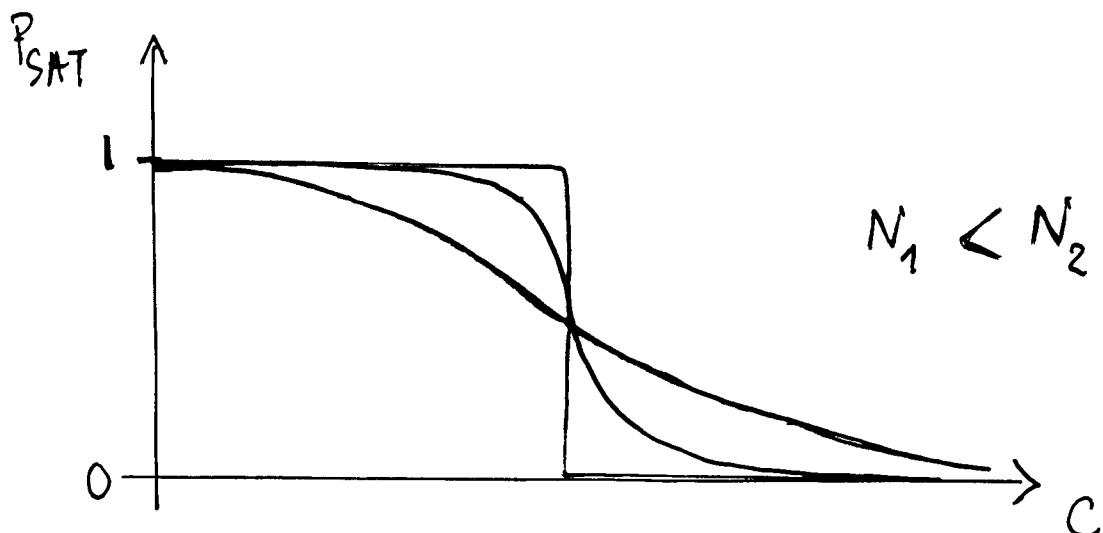
N Boolean variables $x_i = 0, 1$

M clauses: $\begin{cases} i_m < j_m < k_m & (m=1, \dots, M) \\ v_m = 0, 1 & \text{prob } \frac{1}{2}, \frac{1}{2} \end{cases}$

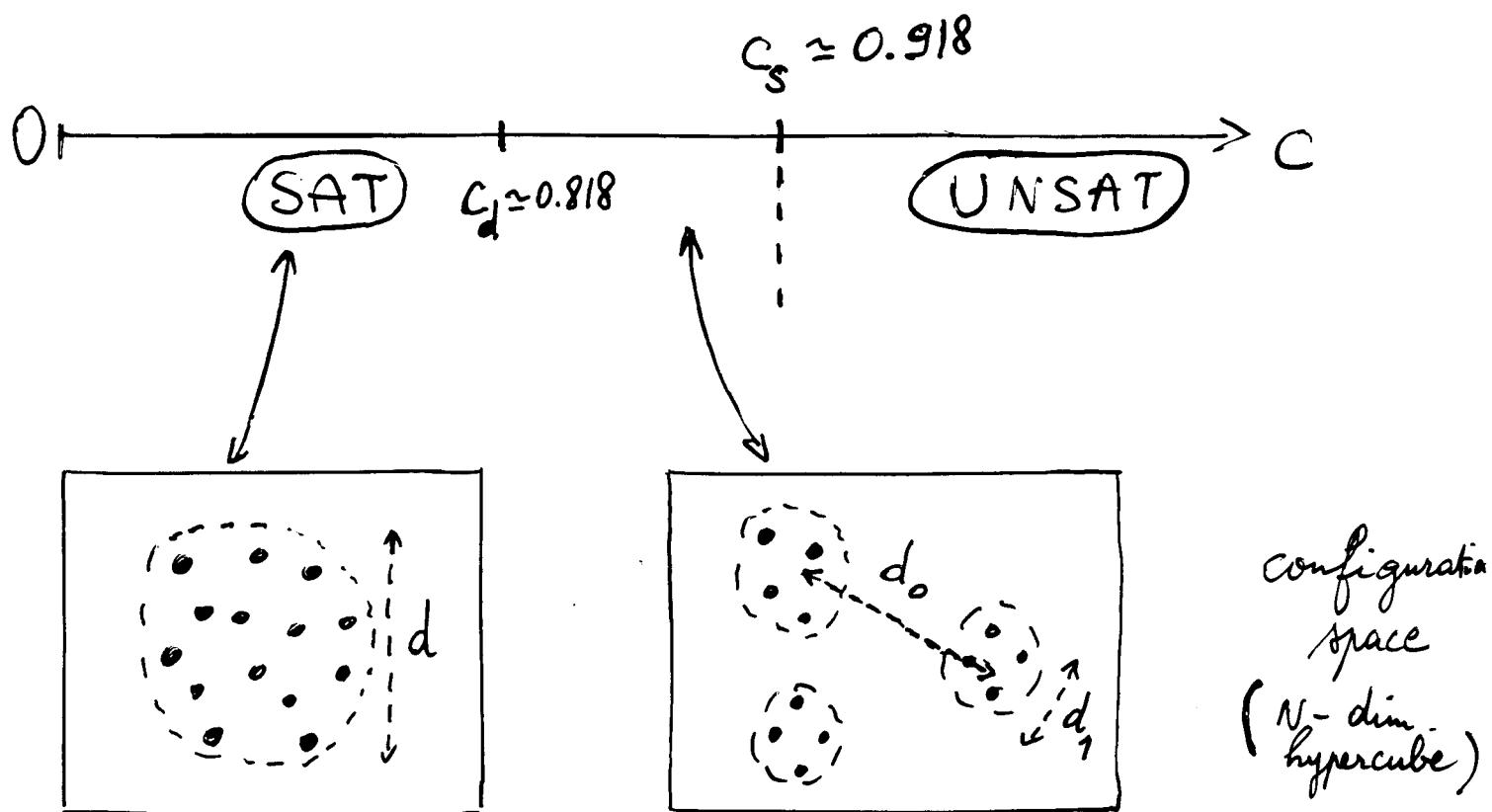
$$x_{i_m} + x_{j_m} + x_{k_m} = v_m \quad (\forall m)$$

- Computationally easy (P)
- SAT vs. UNSAT ?
 - same phenomenology as Random 3-SAT

$$c = \frac{M}{N}$$



Random 3-XORSAT: Results



2^{Ns} solutions

$$s = 1 - c$$

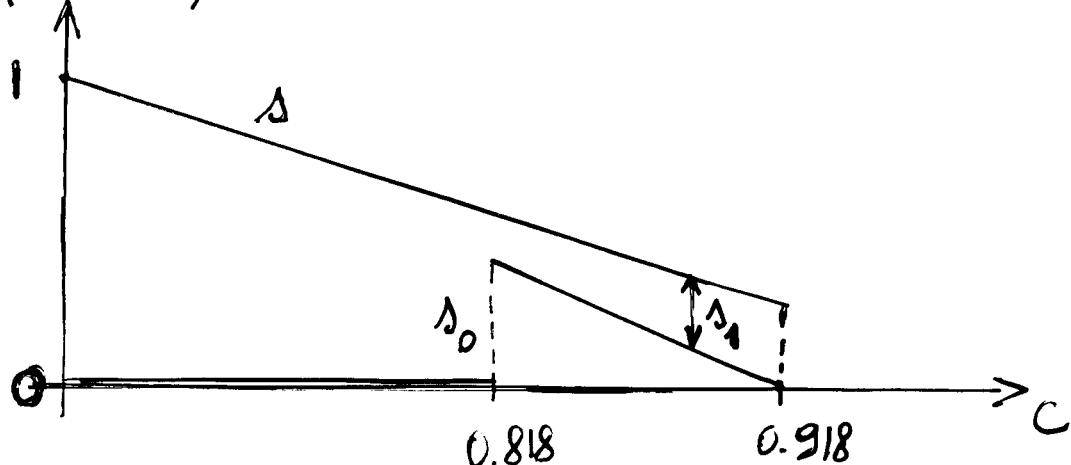
$$d = \frac{1}{2}$$

2^{Ns_0} clusters with
 2^{Ns_1} solutions each

$$s_0 + s_1 = 1 - c$$

$$d_0 = \frac{1}{2}, \quad d_1 \ll \frac{1}{2}$$

$\log(\# \text{ solutions})$



Replica approach (I)

$$N = \sum_{\{x_i=0,1\}} \prod_{l=1}^M \underbrace{\mathbb{1}_{x_{il} + x_{jl} + x_{kl} = v_l}}_{\text{instance}}$$

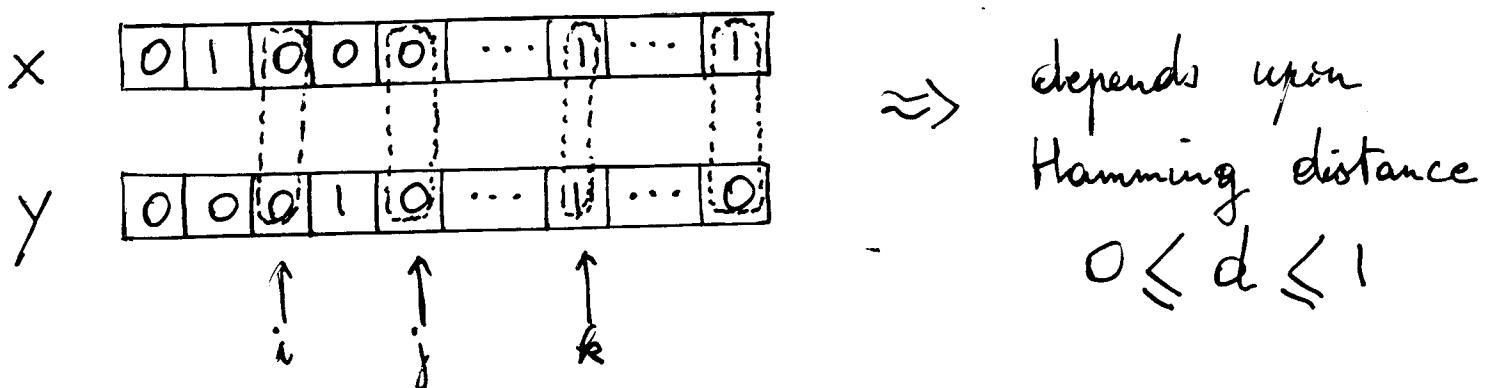
↑
solutions

Compute moments of N ...

$$\begin{aligned}
 E(N) &= \sum_{\{x_i=0,1\}} E \left(\prod_{l=1}^M \mathbb{1}_{x_{il} + x_{jl} + x_{kl} = v_l} \right) \\
 &\quad \underbrace{\qquad\qquad\qquad}_{\prod_{l=1}^M E \left(\mathbb{1}_{x_{il} + x_{jl} + x_{kl} = v_l} \right)} \\
 &= \sum_{\{x_i=0,1\}} \left[\underbrace{\frac{1}{(N)} \sum_{i < j < k} \frac{1}{2} \sum_{v=0,1} \mathbb{1}_{x_i + x_j + x_k = v}}_1 \right]^M \\
 &= 2^N \cdot \left(\frac{1}{2} \right)^M = 2^{N(1-c)} \\
 c_1 &= 1 \quad (\text{upper bound})
 \end{aligned}$$

Replica approach (II)

$$\begin{aligned}
 E(N^2) &= E\left(\left[\sum_{\{x_i=0,1\}} \prod_{e=1}^M \mathbb{1}_{x_{ie} + x_{je} + x_{ke} = v_e}\right]^2\right) \\
 &= E\left(\sum_{\{x_i=0,1\}} \sum_{\{y_i=0,1\}} \prod_{e=1}^M \mathbb{1}_{x_{ie} + x_{je} + x_{ke} = v_e} \mathbb{1}_{y_{ie} + y_{je} + y_{ke} = v_e}\right) \\
 &= \sum_{\{x_i=0,1\}} \sum_{\{y_i=0,1\}} \underbrace{\left\{E\left(\mathbb{1}_{x_i + x_j + x_k = v} \mathbb{1}_{y_i + y_j + y_k = v}\right)\right\}}_M \\
 &\quad \frac{1}{2} E\left(\mathbb{1}_{x_i + x_j + x_k = y_i + y_j + y_k}\right)
 \end{aligned}$$



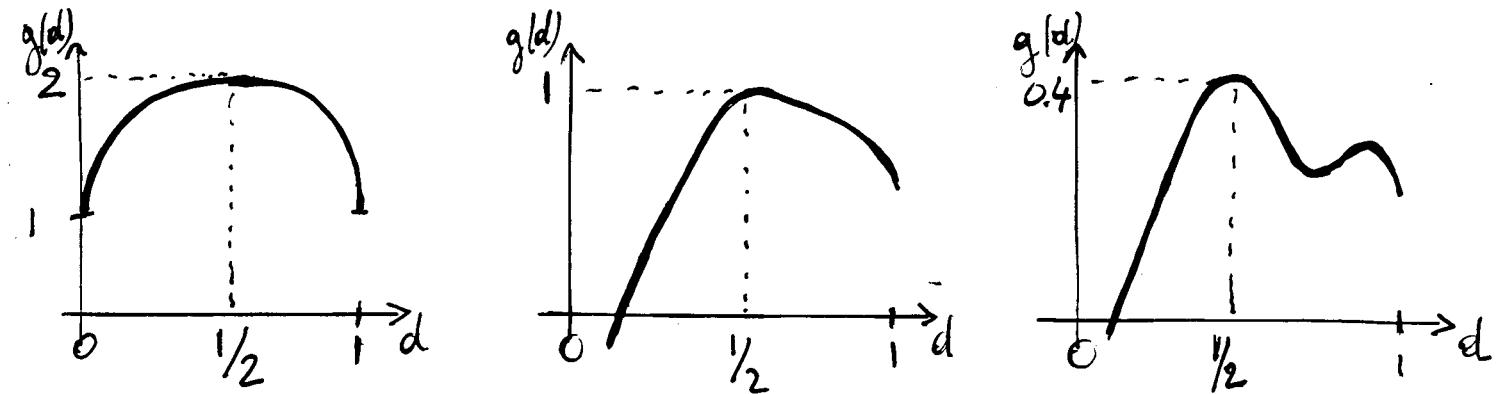
$$\begin{aligned}
 &= \sum_{\{x_i=0,1\}} \sum_{\{y_i=0,1\}} \left\{ \frac{1}{4} \left(1 - (2d(x,y) - 1)^3 \right) \right\}^M \\
 &= 2^N \sum_{\{y_i=0,1\}} \left[\frac{1}{4} \left(1 - (2d(0,y) - 1)^3 \right) \right]^M
 \end{aligned}$$

$$= 2^N \sum_{d=0, \frac{1}{N}, \dots, 1} \left(\frac{1 - (2d-1)^3}{4} \right)^M \mathcal{E}(d)$$

$$\mathcal{E}(d) = \# \text{ configurations at distance } d \text{ from } (0,0,\dots,0) = \binom{N}{Nd}$$

$$E(N^2) = \sum_{d=0, \frac{1}{N}, \dots, 1} 2^N g(d)$$

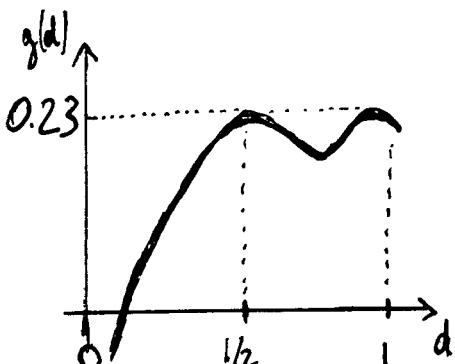
$$g(d) = 1 + c \cdot \log_2 \left(\frac{1 - (2d-1)^3}{4} \right) - (1-d) \log_2 (1-d) - d \log_2 d + o(1)$$



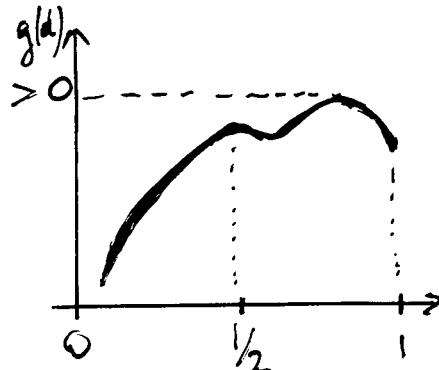
$$c = 0$$

$$c = \frac{1}{2}$$

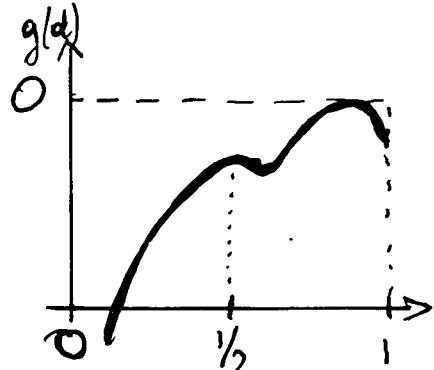
$$c = 0.8$$



$$c = 0.889$$

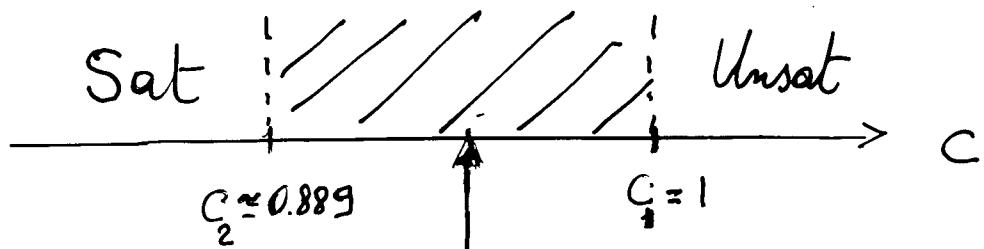


$$c = 0.95$$



$$c = 1.06$$

1st and 2nd moments methods



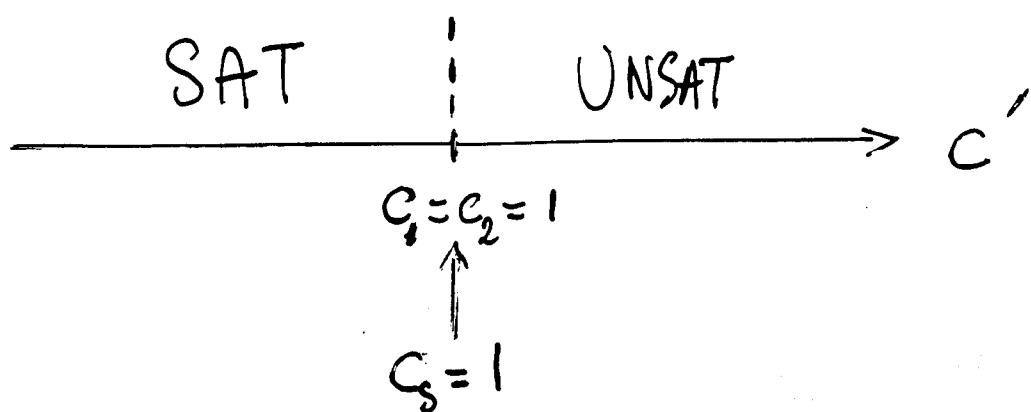
"Random
3-XORSAT"
ensemble

Dubois, Mandler
Cocco, R.M.



pure variable
algorithm

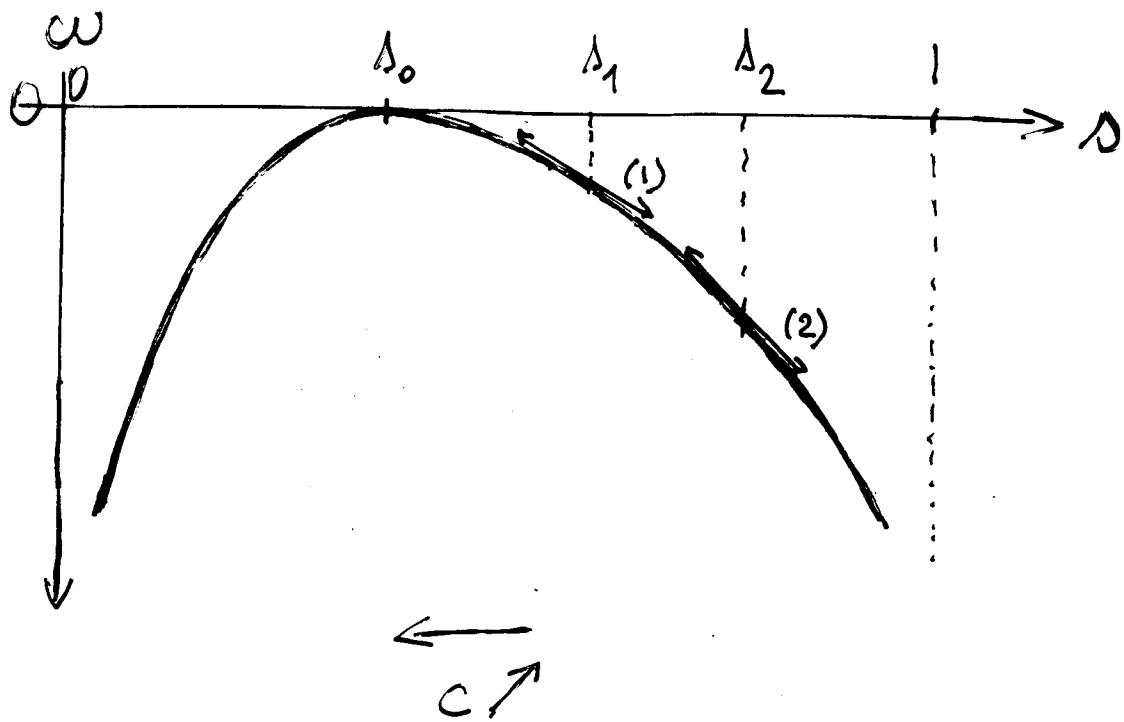
"Constrained
Random 3-XORSAT"
ensemble



Replicas and Large deviations

$$\mathcal{N} = 2^{Ns}$$

$$P(N) = 2^{N\omega(s)}$$



$$\begin{aligned} E(N^q) &= \sum_{N=1}^{2^N} P(N) N^q \\ &\sim \int_0^\infty ds 2^N (\omega(s) + q s) \end{aligned}$$

Laplace method : maximize: $\omega(s) + qs (= g(q))$

$$\frac{\partial \omega}{\partial s} \Big|_{s^*} = -q$$

Typical case: $q \rightarrow 0$

Calculation of $E(N^q)$

• $E(N^1)$: immediate

• $E(N^2)$: $\{x^1, x^2\} \rightarrow d(x^1, x^2) = d^{12}$

• $E(N^q)$: $\{x^1, x^2, \dots, x^q\} \rightarrow \begin{cases} d(x^a, x^b) = d^{ab} \\ d(x^a, x^b, x^c) = d^{abc} \\ d(x^a, x^b, x^c, x^d) = d^{abcd} \end{cases}$

① maximize $g[d^{ab}, d^{abc}, d^{abcd}, \dots]$ $2^q - q - 1$ order parameters
 ↴
 $\frac{\partial g}{\partial d^{a_1 \dots a_l}} = 0, \quad 2 \leq l \leq q$

② get $g(q)$, then $\omega(s) = \min_q [g(q) - qs]$
 and $s_0 = \left. \frac{dg}{dq} \right|_{q=0}$

⚠ * q integer-valued, then $q \rightarrow 0$.

* $E(N^q) \leq (2^N)^q \rightarrow$ moment theorem applies

but here $N \rightarrow \infty$ first, $q \rightarrow 0$ next

\Rightarrow No unicity of $q \rightarrow 0$ continuation.

Replica Symmetric Theory:

Kacz (1967) (em Dyson, 1953)

Edwards - Anderson (1974)

Sherrington - Kirkpatrick (1975)

$$\left\{ \begin{array}{l} d^{a_1, a_2} = d_2 \\ d^{a_1, a_2, a_3} = d_3 \\ d^{a_1, a_2, a_3, a_4} = d_4 \\ \dots \end{array} \right.$$

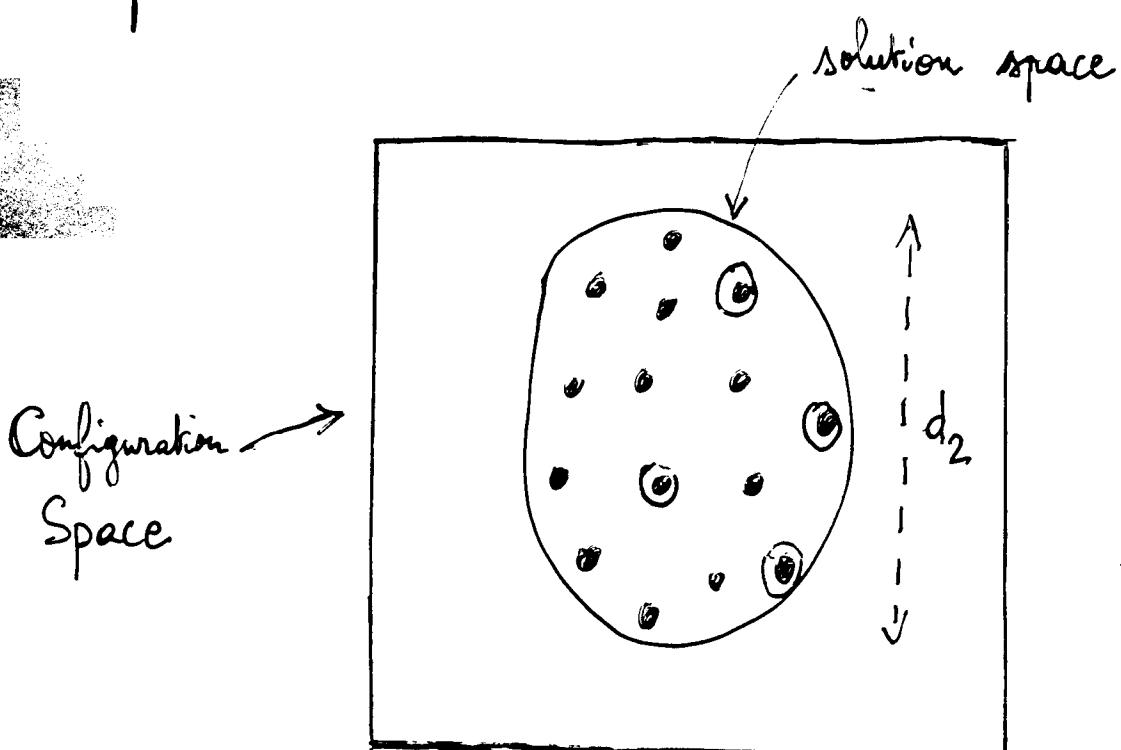
$\forall a_1, a_2, a_3, a_4, \dots$

"replicas cannot be distinguished from each other"

Optimize over d_2, d_3, d_4, \dots

Then, send $q \rightarrow 0$.

Interpretation:



example of
 $q=4$
replicas

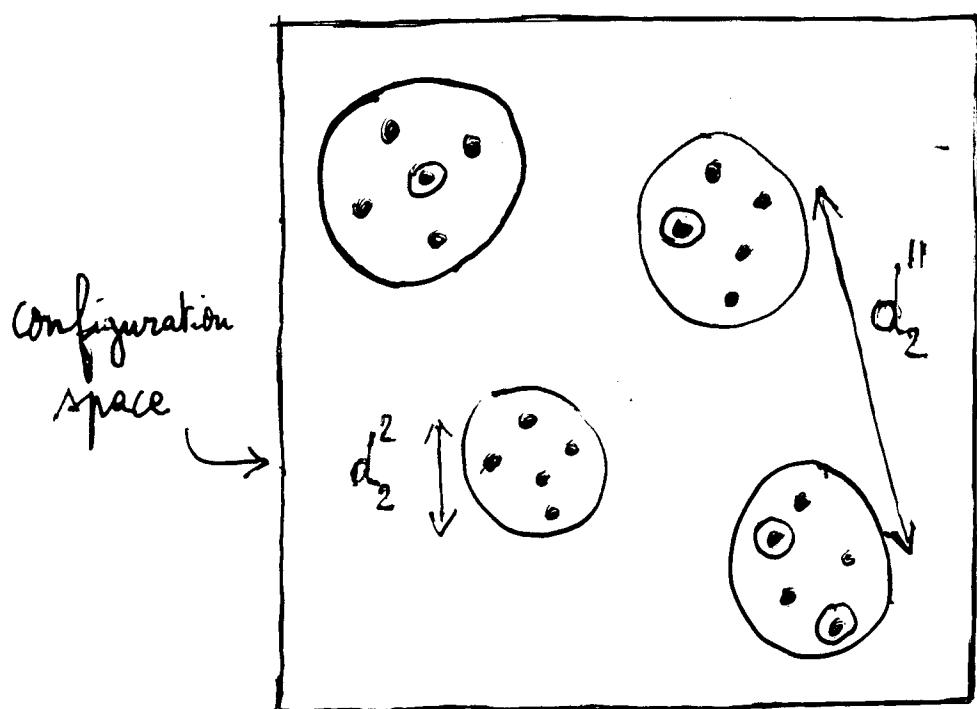
Replica symmetry broken theory

Thouless, Anderson, Palmer 1977
 => Parisi 1979, 1980

$$\left\{ \begin{array}{l} d^{a_1, a_2} = d_2^{11}, d_2^2 \\ d^{a_1, a_2, a_3} = d_3^{111}, d_3^{21}, d_3^3 \\ d^{a_1, a_2, a_3, a_4} = d_4^{1111}, d_4^{211}, d_4^{22}, d_4^{31}, d_4^4 \\ \dots \end{array} \right.$$

$\Rightarrow d^{a_1, a_2, \dots, a_l}$ has as many values as the number of partitions of l .

Interpretation:



example of
 $q=4$
 replicas
 (d_4^{1211})

higher levels
 of symmetry
 breaking ...

Replica predictions for 3-XORSAT

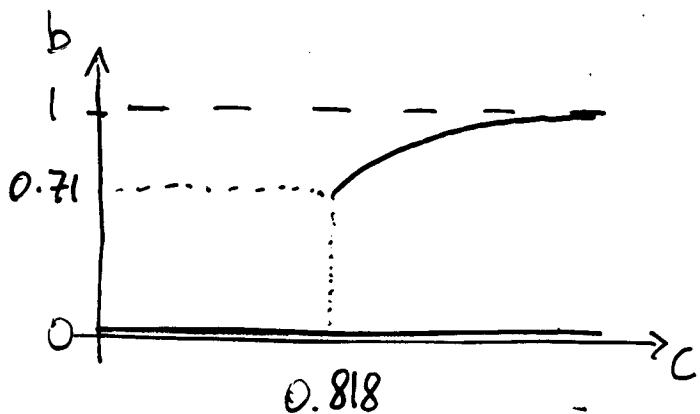
Franz, Ricci, Leone,
Weigt, Zecchina (2001)

$$\begin{cases} S_0 = b - 3cb^2 + 2cb^3 & (\log \# \text{ clusters}) \\ S_1 = 1 - c - S_0 & (\log \# \text{ solutions in each cluster}) \end{cases}$$

with:

$$b = 1 - e^{-3c b^2}$$

↑
local backbone (in a cluster)



$$\rightarrow \begin{cases} d_2^{(2)} = \frac{1-b}{2} \\ d_2^{(1)} = \frac{1}{2} \end{cases}$$

Hand-waving argument

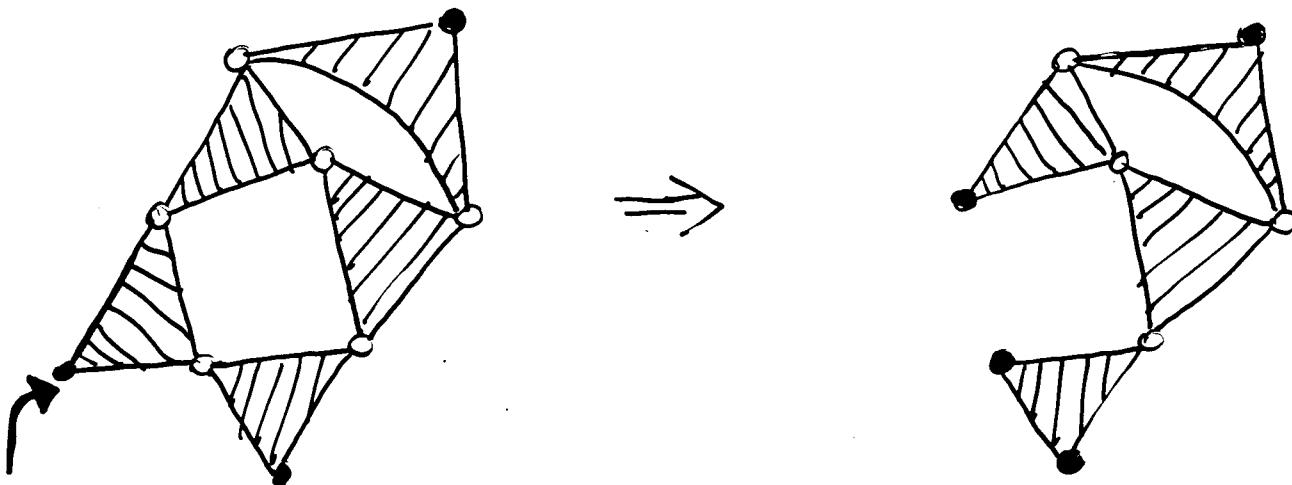
$$\text{variable } x \quad \left\{ \begin{array}{l} x + \dots + \dots = \dots \\ x + \dots + \dots = \dots \\ x + \dots + \dots = \dots \end{array} \right.$$

$$1 - b = \sum_{l=0}^{\infty} e^{-3c} \frac{(3c)^l}{l!} (1 - b^2)^l$$

↑
proba(x free)

"Pure variable" algorithm
 Bieler et al. 1983
 Franco 1980's
 Pittel, Spencer, Wormald 1996

= 2-core percolation on hypergraphs



after T steps (removals) :

- $N_\ell(T) = \# \text{ variables with } \ell \text{ occurrences}$
- $M(T) = M - T = \# \text{ clauses}$

$$E(N_\ell(T+1) - N_\ell(T) | \mathcal{F}) = -\mathbb{1}_{\ell,1} + \mathbb{1}_{\ell,0} + 2P_{\ell+1} - 2P_\ell$$

$$P_j = \frac{j^{E(N_j(T))}}{3M(T)}$$

Analysis of the algorithm

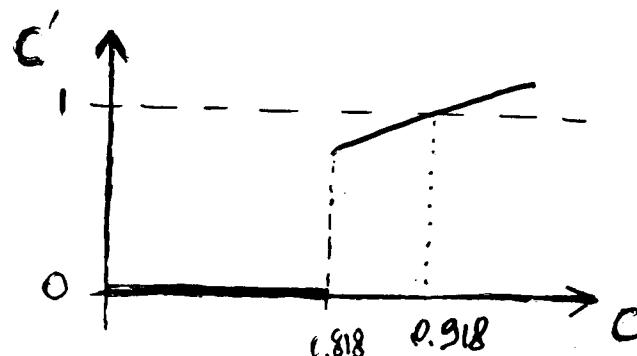
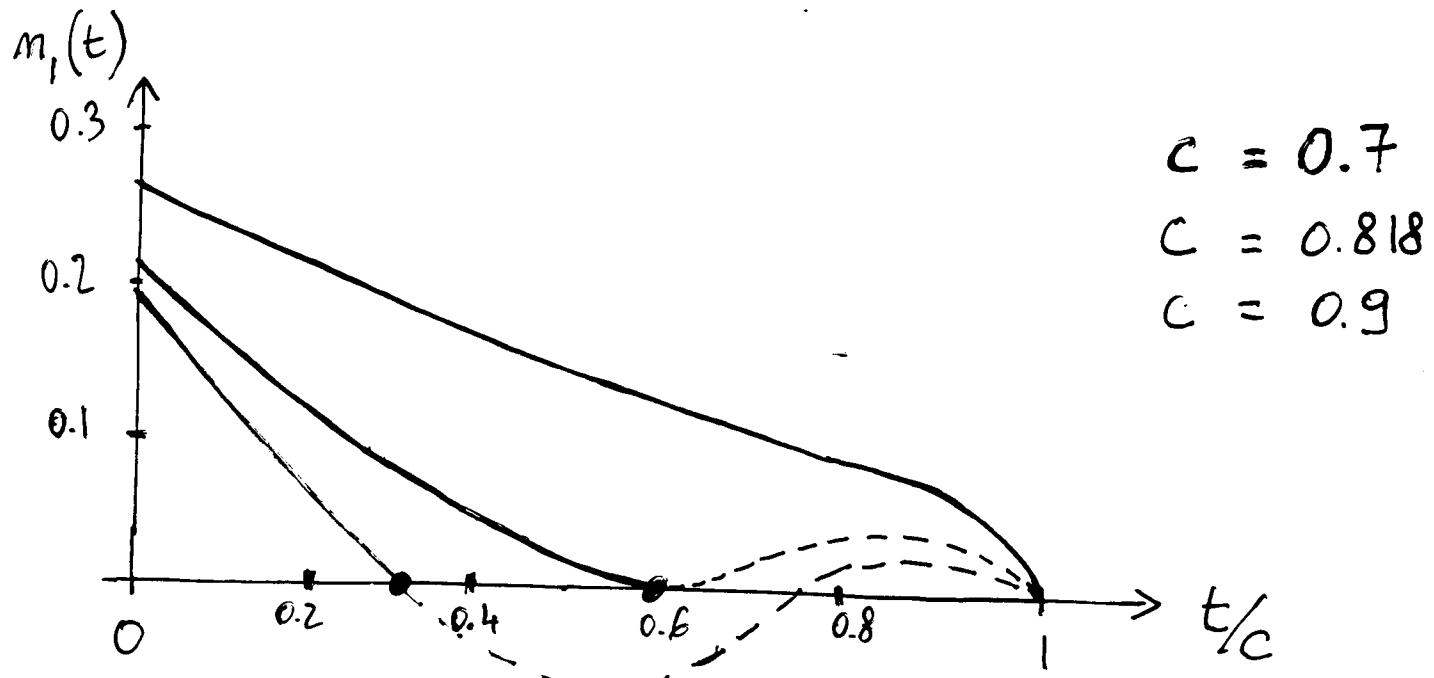
$$N_e(T) \rightarrow N. n_e(t = \frac{T}{N})$$

$$\begin{cases} \frac{dn_e}{dt} = \frac{2}{3(c-t)} ((l+1)n_{e+1} - ln_e) - \mathbb{1}_{e,1} + \mathbb{1}_{e,0} \\ n_e(0) = e^{-3c} \frac{(3c)^e}{e!} \end{cases}$$

Solution:

$$\begin{cases} n_1(t) = 3cb(t)^2 \left(e^{-3cb(t)^2} + b(t) - 1 \right) \\ n_e(t) = e^{-3cb(t)^2} \frac{(3cb^2(t))^e}{e!} \quad (e \geq 2) \end{cases}$$

$$\text{where } b(t) = \left(1 - \frac{t}{c}\right)^{1/3}$$



Constrained random 3-XORSAT

$$\mathcal{F} \longrightarrow \mathcal{F}' \subset \mathcal{F}$$

↑
each variable appears at least twice.

$$c' = \frac{M'}{N'}$$

- $\overline{\mathcal{N}^P} = 2^{N' (1 - c')}$ $\Rightarrow c'_s \leq 1$

- $\overline{\mathcal{N}^{P^2}}$ more complicated but can be done
(Dubois, Mandler 2002)

$$\frac{\overline{\mathcal{N}^P}^2}{\overline{\mathcal{N}^{P^2}}} > 0 \text{ if } c < 1 \Rightarrow c'_s \geq 1$$

$$c'_s = 1$$

- solutions of \mathcal{F}' :

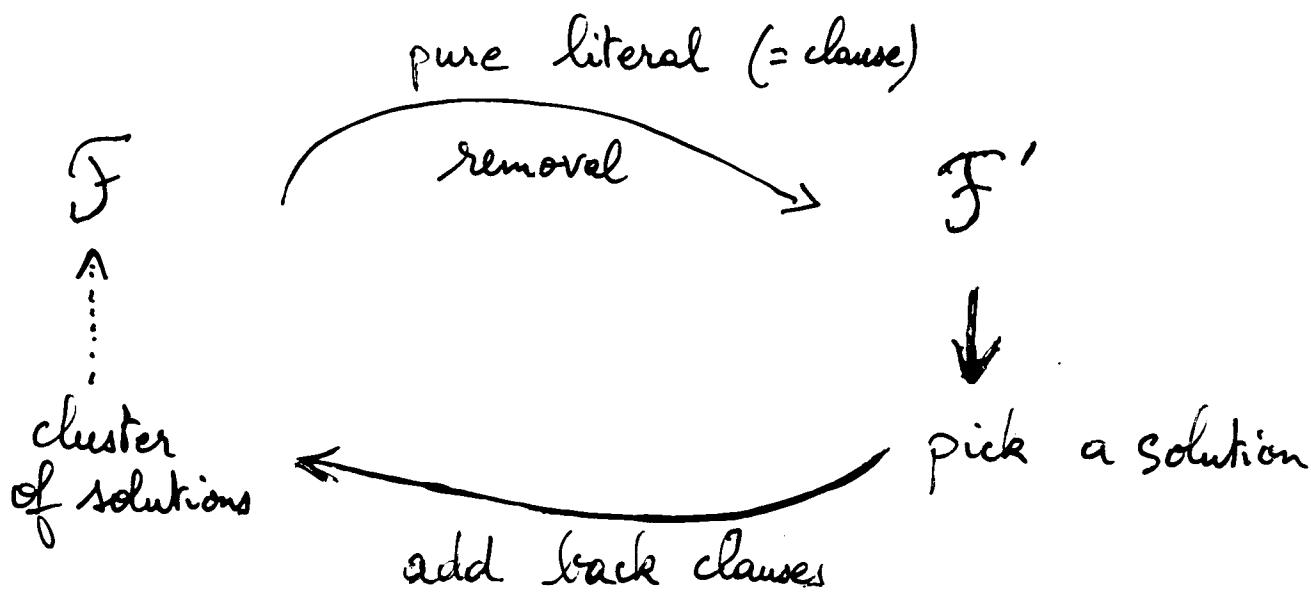
$$d_0 = \frac{1}{2}$$

$$d_0 = \log_2 \mathcal{N}^P = 1 - c'$$

$$= b - 2cb^2 + 3cb^3$$

↓
clusters !

Reconstructing solutions



solution of \tilde{F}' + clause $(x + y + z)$

↑
pure literal

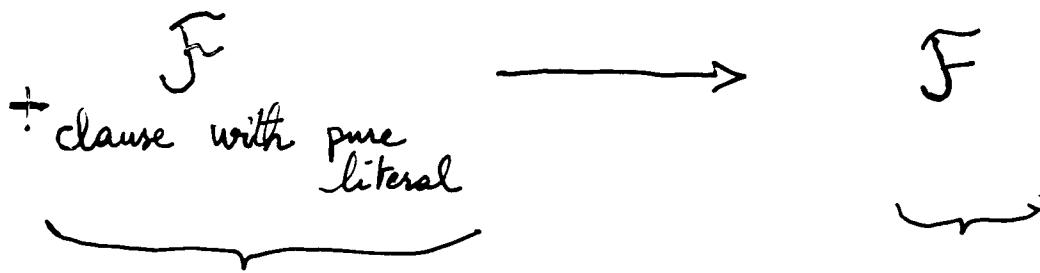
$$\Delta_1 = 2 \int_0^{t_{\text{stop}}} d\tau p_1(\tau) + e^{-3c}$$

↑
variables that never appear

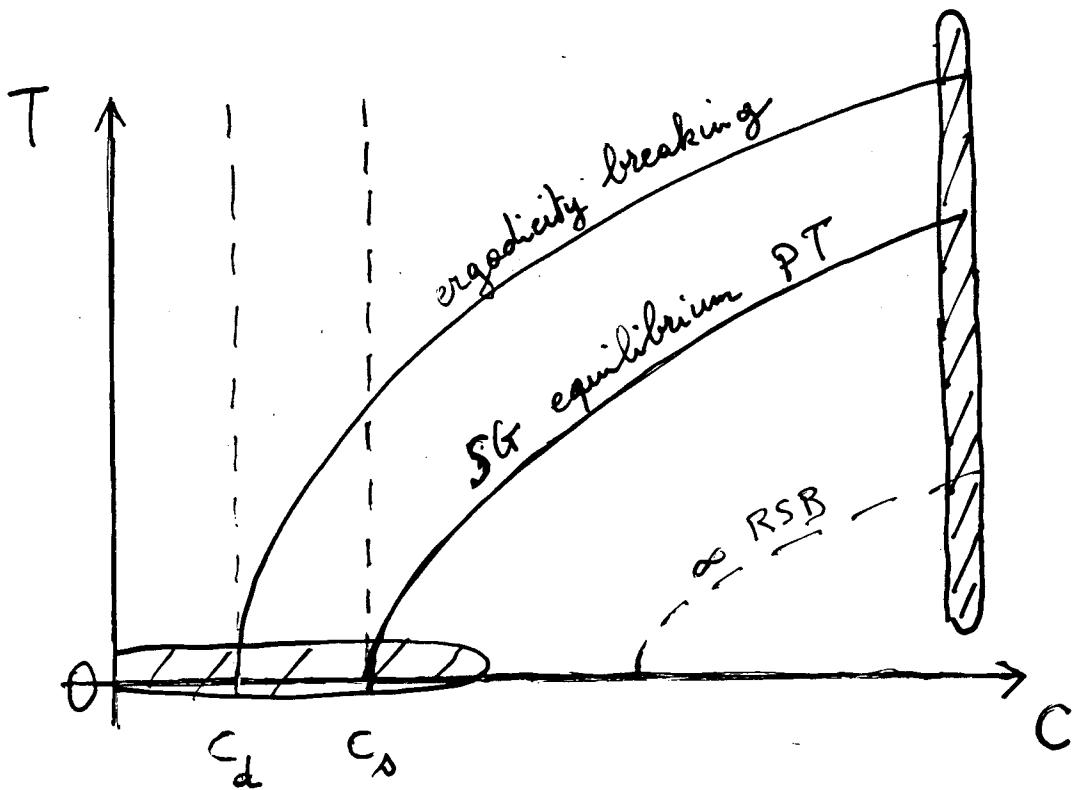
1 solution of \tilde{F}' + frozen var. + free var.
seed of the cluster
backbone b

$$d_1 = \frac{1-b}{2}$$

Finite temperature



$$Z = \Omega^{K-1} (e^{-\beta} + e^{\beta}) \cdot Z'$$



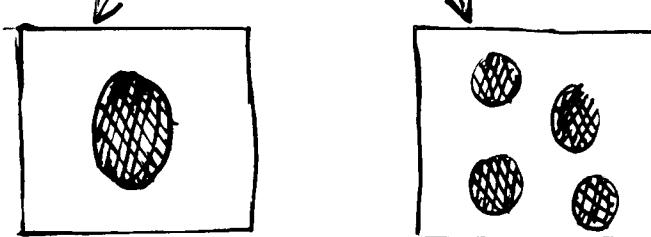
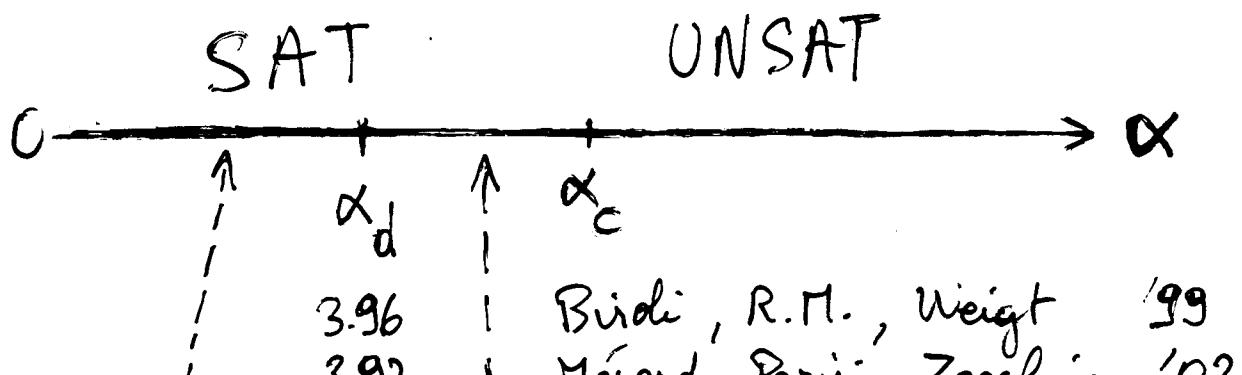
- easy to generalize to larger $K \geq 4$
- case $K=2$.

Back to random K-SAT

$$x_i = 0, 1 \leftrightarrow S_i = -1, +1$$

3-XORSAT : Energy = $-\frac{1}{2} \sum_{e=1}^M J_e S_{ie} S_{je} S_{ke} + \frac{M}{2}$

3-SAT : Energy = $+\sum_{e=1}^M \left(\frac{1 + \sigma_{1e} S_{ie}}{2} \right) \left(\frac{1 + \sigma_{2e} S_{je}}{2} \right) \left(\frac{1 + \sigma_{3e} S_{ke}}{2} \right)$
 $= -\sum_e J_e S_{ie} S_{je} S_{ke} + SS + S + est$
 $\qquad\qquad\qquad \underbrace{-\frac{1}{8} \sigma_{1e} \sigma_{je} \sigma_{ke}}$



$b=0$	$ $	$b_e > 0$	$ $	$b_g > 0$
		$b_g = 0$		

α_d α_c