

Random constraint satisfaction problems: a point of view from physics

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Constraint satisfaction problems : definitions

n variables $\underline{x} = (x_1, \dots, x_n) \in \mathcal{X}^n$ discrete alphabet \mathcal{X}

m constraints $\psi_a(\{x_i\}_{i \in \partial a}) = \begin{cases} 1 & \text{satisfied} \\ 0 & \text{unsatisfied} \end{cases}$

solutions $\mathcal{S} = \{\underline{x} : \psi_a(\underline{x}_{\partial a}) = 1 \forall a\}$

Constraint satisfaction problems : examples

- $\mathcal{X} = \{1, \dots, q\}$, $\psi_a(x_i, x_j) = \mathbb{1}(x_i \neq x_j)$
on the edges $a = \langle i, j \rangle$ of a graph

q-coloring problem

- $\mathcal{X} = \{\text{True}, \text{False}\}$, ψ_a depends on k variables $x_{i_a^1}, \dots, x_{i_a^k}$
 - $\psi_a = \mathbb{1}(z_{i_a^1} \vee \dots \vee z_{i_a^k} = \text{True})$, with $z_i \in \{x_i, \bar{x}_i\}$

k-satisfiability problem

- $\psi_a = \mathbb{1}(x_{i_a^1} \oplus \dots \oplus x_{i_a^k} = y_a)$, with $y_a \in \{\text{True}, \text{False}\}$

k-xor-satisfiability problem

Constraint satisfaction problems

For a given instance (formula/graph), several questions :

- decision problem, is $|\mathcal{S}| > 0$? (said satisfiable if yes)
- counting problem, what is $|\mathcal{S}|$?
- optimization problem, what is $\max_{\underline{x}} \left[\sum_a \psi_a(\underline{x}) \right]$?

Worst-case complexity of the decision problem:

- k -xor-satisfiability easy (P) for all k
- k -satisfiability, q -coloring difficult (NP-complete) for $k, q \geq 3$

Random constraint satisfaction problems

What about their “typical case” complexities ?

“typical”= with high probability in some random ensemble of instances

Examples :

- coloring Erdős-Rényi random graphs $G(n, m)$
choose m edges uniformly at random in the $\binom{n}{2}$ possible ones
- random (xor)satisfiability ensembles
choose m hyperedges (k -uplets of variables), among $\binom{n}{k}$

Most interesting regime : $n, m \rightarrow \infty$ with $\alpha = m/n$ fixed

Random constraint satisfaction problems

$P_{n,\alpha} = \mathbb{P}[\text{random problem with } n \text{ variables and } m = \alpha n \text{ constraints is satisfiable}]$

Obvious observations :

- $P_{n,\alpha}$ decreases with α
- $\lim_{n \rightarrow \infty} P_{n,\alpha} = 0$ for α large enough

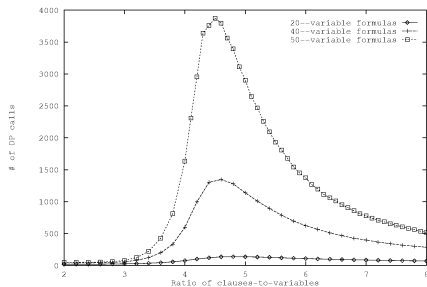
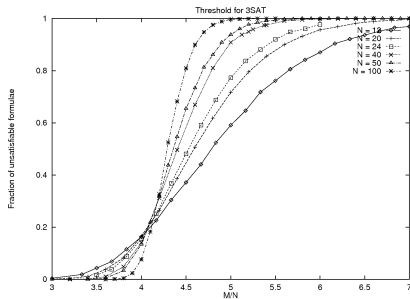
First moment proof (for k -sat) :

$$\begin{aligned} P_{n,\alpha} &= \mathbb{P}[|\mathcal{S}| \geq 1] \leq \mathbb{E}[|\mathcal{S}|] = \sum_{\underline{x}} \mathbb{E} \left[\prod_{a=1}^m \psi_a(\underline{x}) \right] \\ &= 2^n \left(1 - \frac{1}{2^k} \right)^{\alpha n} \rightarrow 0 \end{aligned}$$

for $\alpha > \alpha_{\text{ub}}(k) = 2^k \ln 2$

Random constraint satisfaction problems

“Experimental” observation : phase transition for $1 - P_{n,\alpha}$



associated to a peak in the hardness of solving

Random constraint satisfaction problems

Phase transition conjecture : $\exists \alpha_s(k)$ such that

$$\lim_{n \rightarrow \infty} P_{n,\alpha} = \begin{cases} 1 & \text{if } \alpha < \alpha_s(k) \\ 0 & \text{if } \alpha > \alpha_s(k) \end{cases}$$

Weaker version proven : $\exists \alpha_s(k, n)$ such that

[Friedgut]

$$\lim_{n \rightarrow \infty} P_{n,(1-\epsilon)\alpha_s(k,n)} = 1, \quad \lim_{n \rightarrow \infty} P_{n,(1+\epsilon)\alpha_s(k,n)} = 0$$

Early rigorous results for random k -satisfiability and q -coloring :

- upper and lower bounds on $\alpha_s(k)$
[Chao-Franco, Frieze-Suen, Achlioptas, Dubois, Boufkhad...]
- asymptotics of $\alpha_s(k)$ at large k [Achlioptas, Moore, Naor, Peres]

$$\alpha_s(k) \geq 2^k \ln 2 - k \frac{\ln 2}{2} + O(1) \quad \text{[Achlioptas, Peres (04)]}$$

But :

- no precise value of $\alpha_s(k)$ for small k
- unsatisfactory understanding of algorithmic difficulty at $\alpha < \alpha_s(k)$

Why physics ?

Statistical mechanics :

- configuration space $\underline{x} = (x_1, \dots, x_n)$
- energy function $E(\underline{x})$
- temperature T
- Gibbs-Boltzmann distribution $\mu(\underline{x}) = \exp[-E(\underline{x})/T]/Z$

Low-temperature statistical physics \approx combinatorial optimization

randomness in the distribution of instances \approx disordered systems
(spin glasses)

random graphs \rightarrow mean-field diluted spin glasses

graph coloring corresponds to antiferromagnetic Potts model

Outcomes of the physics approach

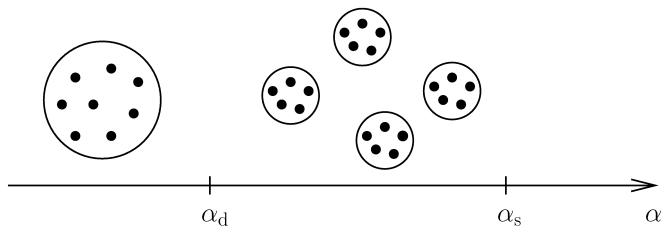
- quantitative estimation of $\alpha_s(k)$
- refined picture of the satisfiable phase
- analysis of known algorithms
- suggestion of new ones

Refined picture of the satisfiable phase

Exponential number of solutions for $\alpha < \alpha_s$, $\sim \exp[ns(\alpha)]$

Clustering transition at another threshold $\alpha_d < \alpha_s$:

apparition of an exponential number of clusters ($\sim \exp[n\Sigma(\alpha)]$),
each containing an exponential number of solutions ($\sim \exp[ns_{\text{int}}(\alpha)]$)



at α_s cancellation of $\Sigma(\alpha)$, not of $s(\alpha)$

Refined picture of the satisfiable phase

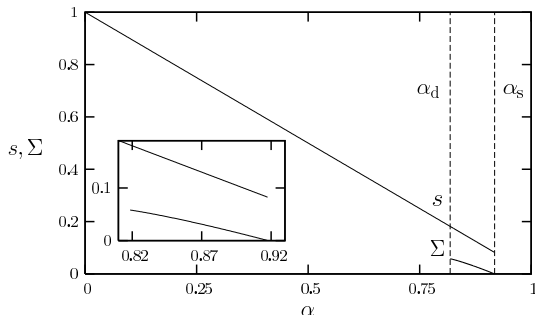
Can be proven rigorously for random xorsat

[Creignou, Daudé]

[Cocco, Dubois, Mandler, Monasson]

[Mézard, Ricci-Tersenghi, Zecchina]

At α_d , apparition of a 2-core in the hypergraph



$$s(\alpha) = \Sigma(\alpha) + s_{\text{int}}(\alpha)$$

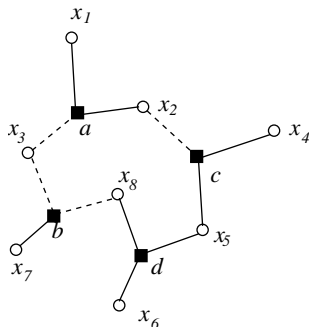
$$\Sigma(\alpha_s) = 0$$

[more complicated picture for sat and col]

If the formula F has solutions,

define $\mu(\underline{x}) = \frac{1}{Z} \prod \psi_a(\underline{x}_{\partial a})$ uniform measure on \mathcal{S}

Factor graph representation
of a formula :

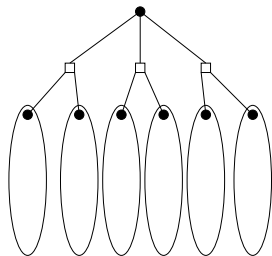


Crucial property : in the $n, m \rightarrow \infty$ limit with $\alpha = m/n$ fixed
local convergence of the factor graph to a random Galton-Watson tree

Methods

For a tree factor graph $\mu(\underline{x})$ is a rather simple object
(Belief Propagation is exact)

All marginal probabilities can be easily computed recursively :



Sparse random graphs converge locally to trees

Is it enough for their $\mu(\underline{x})$ to converge locally to the measure on the associated tree ?

It depends... (on the correlations decay)

- Yes in “replica symmetric” cases
 - Ferromagnetic Ising models
 - Matchings
 - Random CSP for $\alpha < \alpha_d$

[Dembo, Montanari]
[Bordenave, Lelarge, Salez]

Removing a finite subtree, the variables at the boundary are asymptotically independent

Allows to compute in particular $\lim \frac{1}{n} \log Z$ entropy of solutions

- No in presence of “replica symmetry breaking” $(\alpha_d < \alpha < \alpha_s)$

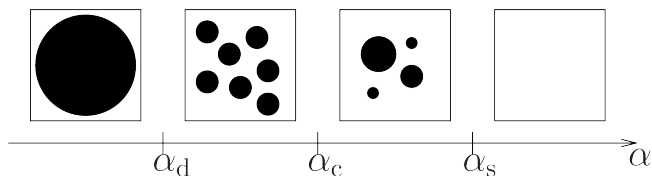
Configuration space partitioned in clusters \Rightarrow long-range correlations

Removing a finite subtree, the variables at the boundary remains correlated, self-consistent ansatz on these boundary conditions

With $\mu(\underline{x}) = \sum_C w_C \mu_C(\underline{x})$, each μ_C has short-range correlations, can be treated as above

Properties of w_C encode the value of $\Sigma(\alpha)$, hence allows the computation of α_s

Clustering for k -sat and q -col



Further “condensation” transition :

exponential number of clusters only for $\alpha \in [\alpha_d, \alpha_c]$

In general $\alpha_c < \alpha_s$, for XORSAT they coincide

Recent rigorous results

Physical intuition and heuristic results turned into rigorous proofs

- Existence of clusters and frozen variables [Achlioptas, Molloy]
- Improved bounds on $\alpha_s(k)$ at large k

$$\alpha_s(k) \geq 2^k \ln 2 - \frac{3 \ln 2}{2} + o(1) \quad [\text{Coja-Oghlan, Panagiotou (12)}]$$

$$\alpha_s(k) = 2^k \ln 2 - \frac{1 + \ln 2}{2} + o(1) \quad [\text{Coja-Oghlan (14)}]$$

- independence number of d -regular graphs for large (but finite) d
[Ding, Sly, Sun (13)]

Schemes of rigorous proofs:

- Large k, q results, of combinatorics nature
[Achlioptas, Coja-Oghlan]
- Interpolation method, originally devised for the Sherrington-Kirkpatrick model
[Guerra, Toninelli, Aizenman, Talagrand, Panchenko]
- Local convergence approach [Aldous]
 - done for sparse random graphs in the “replica symmetric” regime
 - open problem in the “replica symmetry breaking” regime

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