

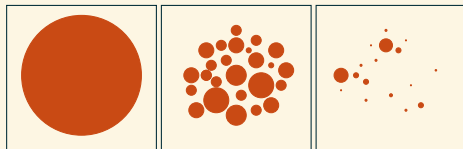
Diluted spin glass models

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Overview

- ▶ replica symmetry and Belief Propagation
- ▶ construction of Bethe states
- ▶ application: random matrices and the k -XORSAT threshold

Reminder: predictions from the cavity method

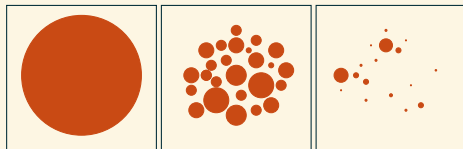


Replica symmetry breaking

[MP00,KMRTSZ07]

- ▶ replica symmetry
- ▶ dynamic replica symmetry breaking
- ▶ (static) replica symmetry breaking

Reminder: predictions from the cavity method

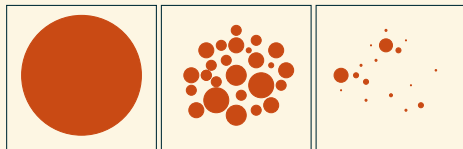


Replica symmetry

[MP00,KMRTSZ07]

- ▶ $\mu_{\mathbb{G},\beta}(\{\sigma_{x_1} = s, \sigma_{x_2} = t\}) \sim \mu_{\mathbb{G},\beta}(\{\sigma_{x_1} = s\})\mu_{\mathbb{G},\beta}(\{\sigma_{x_2} = t\})$
- ▶ in other words, $\mu_{\mathbb{G},\beta}$ is $o(1)$ -extremal

Reminder: predictions from the cavity method



Bethe states

[MPRTRLZ99,MP00,KMRTSZ07]

- ▶ the phase space decomposes into **pure states**
- ▶ each of them induces a BP fixed point (but not vice versa)
- ▶ replica symmetry iff there is just one Bethe state

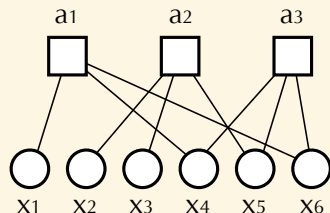
The replica symmetric case

An Erdős-Rényi factor graph model

- ▶ a random factor graph model $\mathbb{G} = \mathbb{G}(n, \mathbf{m})$ with variables x_1, \dots, x_n
- ▶ the variable range over Ω
- ▶ k -ary factor nodes a_1, \dots, a_m with $\mathbf{m} \sim \text{Po}(dn/k)$
- ▶ the factor nodes are *independent*
- ▶ suppose they all carry the same weight function $\psi(\cdot) > 0$
- ▶ the model induces a Boltzmann distribution

$$\mu_{\mathbb{G}}(\sigma) = \frac{1}{Z(\mathbb{G})} \prod_{i=1}^m \psi(\sigma_{\partial a_i})$$

The replica symmetric case



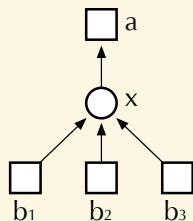
Replica symmetry assumption

- ▶ we assume that

$$\sum_{s, t \in \Omega} \mathbb{E} \left| \mu_{\mathbb{G}}(\{\sigma_{x_1} = s, \sigma_{x_2} = t\}) - \mu_{\mathbb{G}}(\{\sigma_{x_1} = s\}) \mu_{\mathbb{G}}(\{\sigma_{x_2} = t\}) \right| = o(1)$$

- ▶ in other words, $\mu_{\mathbb{G}}$ is $o(1)$ -extremal with high probability

The replica symmetric case



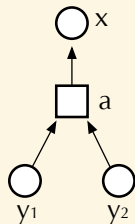
The standard messages

- ▶ obtain $\mathbb{G} - a$ by removing a from \mathbb{G} and let

$$\mu_{\mathbb{G}, x \rightarrow a}(s) = \mu_{\mathbb{G} - a, x}(s) \quad (s \in \Omega)$$

- ▶ *the marginal of x in $\mathbb{G} - a$*

The replica symmetric case



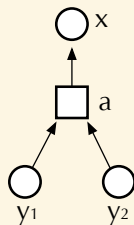
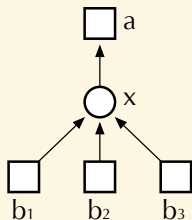
The standard messages

- ▶ obtain $\mathbb{G} - (\partial x \setminus a)$ by removing all $b \in \partial x \setminus a$

$$\mu_{\mathbb{G}, a \rightarrow x}(s) = \mu_{\mathbb{G} - (\partial x \setminus a), x}(s) \quad (s \in \Omega)$$

- ▶ *the marginal of x in $\mathbb{G} - (\partial x \setminus a)$*

The replica symmetric case



Reminder: Belief Propagation

We expect that

$$\mu_{\mathbb{G}, x \rightarrow a}(s) \propto \prod_{b \in \partial x \setminus a} \mu_{\mathbb{G}, b \rightarrow x}(s)$$

$$\mu_{\mathbb{G}, a \rightarrow x}(s) \propto \sum_{\sigma \in \Omega^{\partial a}} \mathbf{1}\{\sigma_x = s\} \psi_a(\sigma) \prod_{y \in \partial a \setminus x} \mu_{\mathbb{G}, y \rightarrow a}(\sigma_y)$$

The replica symmetric case

Theorem

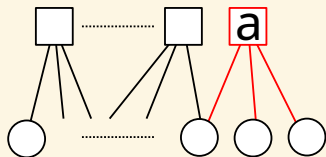
[COP18]

If replica symmetry holds we have for all $s \in \Omega$,

$$\lim_{n \rightarrow \infty} \sum_{a \in \partial x_1} \left| \mu_{\mathbb{G}, x \rightarrow a}(s) - \frac{\prod_{b \in \partial x \setminus a} \mu_{\mathbb{G}, b \rightarrow x}(s)}{\sum_{t \in \Omega} \prod_{b \in \partial x \setminus a} \mu_{\mathbb{G}, b \rightarrow x}(t)} \right| = 0$$
$$\lim_{n \rightarrow \infty} \sum_{a \in \partial x_1} \left| \mu_{\mathbb{G}, a \rightarrow x}(s) - \frac{\sum_{\sigma \in \Omega^{\partial a}} \mathbf{1}\{\sigma_x = s\} \psi_a(\sigma) \prod_{y \in \partial a \setminus x} \mu_{\mathbb{G}, y \rightarrow a}(\sigma_y)}{\sum_{\sigma \in \Omega^{\partial a}} \psi_a(\sigma) \prod_{y \in \partial a \setminus x} \mu_{\mathbb{G}, y \rightarrow a}(\sigma_y)} \right| = 0$$

“the standard messages satisfy the BP equations”

The replica symmetric case

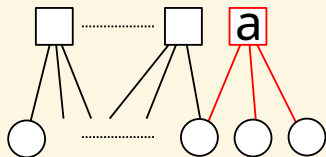


Lemma

Obtain $\mathbb{G} + a$ by adding a single factor node a arbitrarily. Then w.h.p. $\mu_{\mathbb{G}+a}$ is $o(1)$ -extremal and

$$\frac{1}{n} \sum_{i=1}^n d_{\text{TV}}(\mu_{\mathbb{G}+a, x_i}, \mu_{\mathbb{G}, x_i}) = o(1)$$

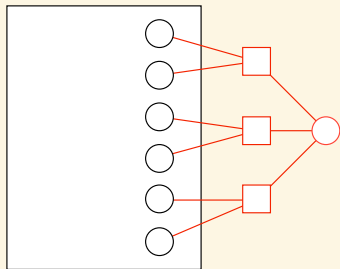
The replica symmetric case



Reminder: cut metric

$$d_{\square}(\mu, \nu) = \frac{1}{n} \min_{\gamma \in \Gamma(\mu, \nu)} \max_{\substack{I \subset \{1, \dots, n\} \\ B \subset \Omega^n \times \Omega^n \\ \omega \in \Omega}} \left| \sum_{i \in I} \sum_{(\sigma, \tau) \in B} \gamma(\sigma, \tau) (\mathbf{1}\{\sigma_i = \omega\} - \mathbf{1}\{\tau_i = \omega\}) \right|$$

The replica symmetric case



Proof of the theorem

- ▶ add a new random variable node along with adjacent factors
- ▶ the attachment points are random
- ▶ due to extremality, their joint distribution factorises
- ▶ we can therefore verify the BP equations

The replica symmetric case

Corollary

[COP18]

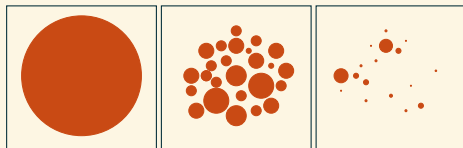
Assume replica symmetry. Then and that the Bethe free entropy $\mathcal{B}(\mathbb{G})$ of the standard messages satisfies

$$\lim_{n \rightarrow \infty} \frac{1}{n} \mathcal{B}(\mathbb{G}) = B \in \mathbb{R} \quad \text{in probability.}$$

Then

$$\lim_{n \rightarrow \infty} \frac{1}{n} \left| \log Z(\mathbb{G}) - B \right| = 0 \quad \text{in probability.}$$

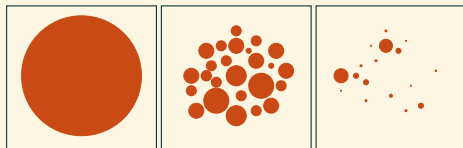
Bethe states



Absence of replica symmetry

- ▶ let us drop the assumption of replica symmetry
- ▶ we expect any number of Bethe states

Bethe states



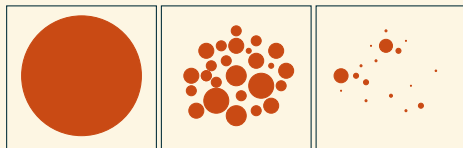
The conditional standard messages

For $S \subset \Omega^n$ let

$$\mu_{\mathbb{G}, x \rightarrow a}(s \mid S) = \mu_{\mathbb{G} - a, x}(s \mid S) \quad (s \in \Omega)$$

$$\mu_{\mathbb{G}, a \rightarrow x}(s \mid S) = \mu_{\mathbb{G} - (\partial x \setminus a), x}(s \mid S) \quad (s \in \Omega)$$

Bethe states



Definition

A set $S \subset \Omega^n$ is an ε -*Bethe state* if

$$\frac{1}{n} \sum_{i=1}^n \sum_{a \in \partial x_i} \left| \mu_{\mathbb{G}, x_i \rightarrow a}(s | S) - \frac{\prod_{b \in \partial x_i \setminus a} \mu_{\mathbb{G}, b \rightarrow x_i}(s | S)}{\sum_{t \in \Omega} \prod_{b \in \partial x_i \setminus a} \mu_{\mathbb{G}, b \rightarrow x_i}(t | S)} \right| < \varepsilon$$
$$\frac{1}{n} \sum_{i=1}^n \sum_{a \in \partial x_i} \left| \mu_{\mathbb{G}, a \rightarrow x_i}(s | S) - \frac{\sum_{\sigma \in \Omega^{\partial a}} \mathbf{1}\{\sigma_{x_i} = s\} \psi_a(\sigma) \prod_y \mu_{\mathbb{G}, y \rightarrow a}(\sigma_y | S)}{\sum_{\sigma \in \Omega^{\partial a}} \psi_a(\sigma) \prod_{y \in \partial a \setminus x_i} \mu_{\mathbb{G}, y \rightarrow a}(\sigma_y | S)} \right| < \varepsilon$$

Bethe states



Theorem

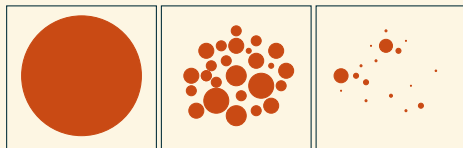
[COP19]

For any $\varepsilon > 0$ there exist $L \geq 1$ and $n_0 > 0$ such that for all $n > n_0$ with high probability there exist $S_1, \dots, S_\ell \subset \Omega^n$, $\ell \leq L$, such that

- (i) S_1, \dots, S_ℓ are ε -Bethe states
- (ii) $\sum_{i=1}^{\ell} \mu_{\mathbb{G}}(S_i) > 1 - \varepsilon$

“any random factor graph model has a Bethe state decomposition”

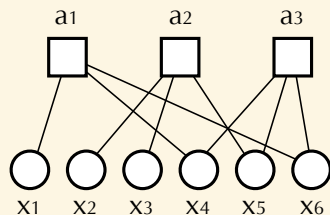
Bethe states



Proof

- ▶ apply pinning repeatedly like in the decomposition theorem
- ▶ to each sub-cube apply a coupling argument as in the RS case
- ▶ **delicate point:** the “edits” shift the relative weights of the Bethe states

Random k -XORSAT



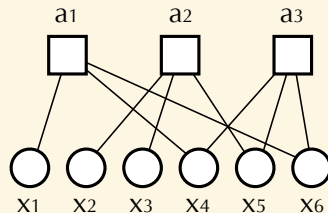
The random k -XORSAT problem

- ▶ variables x_1, \dots, x_n ranging over $\Omega = \mathbb{F}_2$
- ▶ check nodes a_1, \dots, a_m represent k -XOR constraints:

$$x_{i_1} + x_{i_2} + \dots + x_{i_k} = y_i$$

- ▶ with $m \sim \text{Po}(dn/k)$, for what d is it possible to satisfy all constraints?

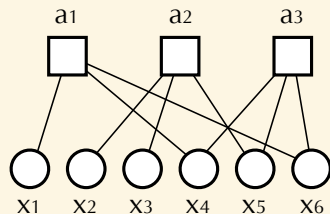
Random k -XORSAT



Equivalent formulation

- ▶ let A be the random (bi)adjacency matrix
- ▶ for what d, k does A have full row rank?
- ▶ equivalently, determine the dimension of the kernel
- ▶ $Z = \#\{\text{solutions to } Ax = 0\}$
- ▶ $\mu_A(\sigma) = \mathbf{1}\{A\sigma = 0\} / Z$

Random k -XORSAT



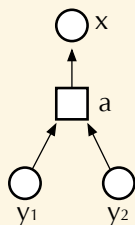
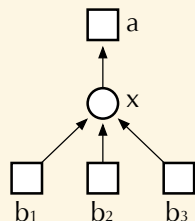
Trivial BP solution

- ▶ set all messages to $1/2$:

$$\mu_{x \rightarrow a}(0) = \mu_{x \rightarrow a}(1) = \frac{1}{2}$$

$$\mu_{a \rightarrow x}(0) = \mu_{a \rightarrow x}(1) = \frac{1}{2}$$

Random k -XORSAT



Trivial BP solution

- ▶ set all messages to $1/2$:

$$\mu_{x \rightarrow a}(0) = \mu_{x \rightarrow a}(1) = \frac{1}{2} \quad \mu_{a \rightarrow x}(0) = \mu_{a \rightarrow x}(1) = \frac{1}{2}$$

- ▶ Bethe free entropy = $n(1 - d/k) \log 2$
- ▶ *Are there other fixed points* (there'd better be)?

Random k -XORSAT

Lemma

For any $m \times n$ -matrix A for a random $\sigma \in \ker A$ we have

$$P[\sigma_i = 0] \in \{1/2, 1\}$$

If $P[\sigma_i = 0] = 1$, call coordinate i *frozen*.

Proof

- ▶ consider a basis ξ_1, \dots, ξ_ℓ of the kernel
- ▶ $\sigma = \omega_1 \xi_1 + \dots + \omega_\ell \xi_\ell$

Random k -XORSAT

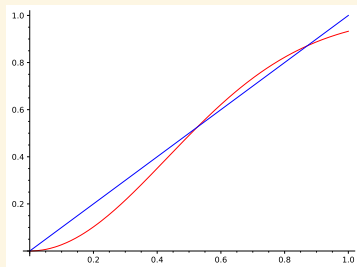
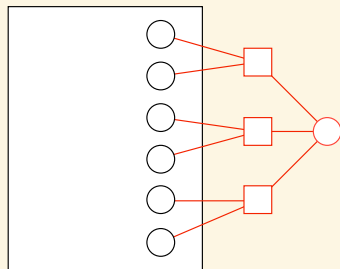
Bethe states of k -XORSAT

- ▶ consider the Bethe states S_1, \dots, S_L
- ▶ for each of them let $\alpha_1, \dots, \alpha_L$ be the fraction of frozen variables:

$$\alpha_j = \frac{1}{n} \sum_{i=1}^n \mathbf{1} \{P[\sigma_i = 0 \mid S_j] = 1\}$$

- ▶ *what values can $\alpha_1, \dots, \alpha_L$ take?*

Random k -XORSAT



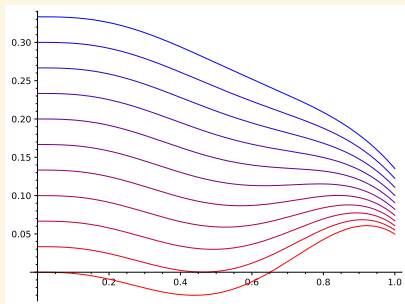
The fixed point property

- ▶ The α_i satisfy the fixed point equation

$$\alpha = 1 - \exp(-d\alpha^{k-1})$$

- ▶ this equation has at most three solutions
- ▶ at most two of them are stable

Random k -XORSAT



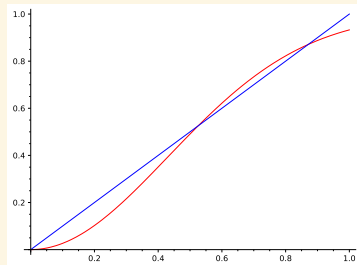
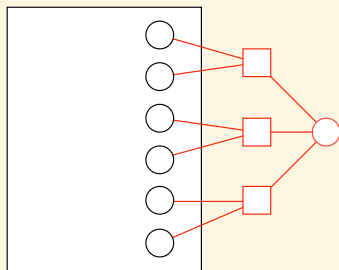
The ensuing Bethe free entropy

- ▶ the fixed points translate into stationary points of the BFE

$$\mathcal{B}(\alpha) = \exp(-d\alpha^{k-1}) - \frac{d}{k} \left(1 - k\alpha^{k-1} + (k-1)\alpha^k \right)$$

- ▶ the stable ones are local maxima
- ▶ from a certain threshold $d^* = d^*(k)$ the positive one dominates

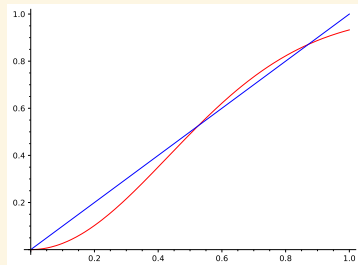
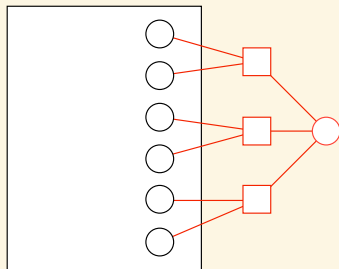
Random k -XORSAT



The method of moments

- ▶ we can finally calculate the *expected* number of solutions to $Ax = 0$ with a certain fraction of frozen coordinates
- ▶ if that fraction is a fixed point, the answers (with suitable conditioning) boils down to $\mathcal{B}(\alpha)$

Random k -XORSAT



Theorem

[DM02,MRTZ02,DGMMPR10,PS16]

The random k -XORSAT satisfiability threshold equals $d^*(k)$.

Random k -XORSAT

Summary

- ▶ random factor graph models possess Bethe states
- ▶ they can be constructed **obliviously** via pinning
- ▶ we can harness Bethe decompositions to derive combinatorial results
- ▶ *Example:* the k -XORSAT threshold