

# Boltzmann Machines on the Nishimori line

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# Outline

- 1 High Dimensional Inference vs Statistical Mechanics
- 2 Towards multipartite models
- 3 Boltzmann Machines on the NL

# Wigner Spiked model (WS)

We face the task of recovering the components of a binary  $\pm 1$  signal  $(\sigma_i^*)_{i \in \Lambda}$ , with  $|\Lambda| = N$  set of indices, *iid* with probabilities 1/2 through the observations

$$y_{ij}(\sigma^*) = \sqrt{\frac{\mu}{2N}} \sigma_i^* \sigma_j^* + z_{ij}, \quad \tilde{y}_i(\sigma^*) = \sqrt{h} \sigma_i^* + \tilde{z}_i, \quad z_{ij}, \tilde{z}_i \stackrel{\text{iid}}{\sim} \mathcal{N}(0, 1)$$

The problem is high dimensional since we have a large number of components  $N$  to recover. The Bayes posterior measure is

$$\begin{aligned} P(\sigma^* = \sigma | y, \tilde{y}) &= \exp \left[ -\frac{1}{2} \sum_{i,j \in \Lambda} \left( y_{ij} - \sqrt{\frac{\mu}{2N}} \sigma_i \sigma_j \right)^2 - \frac{1}{2} \sum_{i \in \Lambda} \left( \tilde{y}_i - \sqrt{h} \sigma_i \right)^2 \right] \frac{1}{Z(y, \tilde{y})} \\ &= \frac{\exp[-H_N(\sigma, \sigma^*, z)]}{Z(\sigma, \sigma^*, z)} \end{aligned}$$

The posterior can thus be rewritten as a Gibbs random measure, the randomness being in the quenched variables  $\sigma^*, z$ . We can treat this problem as a disordered model.

The Hamiltonian is

$$H_N(\sigma, \sigma^*, z) = - \sum_{i,j \in \Lambda} \left[ \sqrt{\frac{\mu}{2N}} z_{ij} \sigma_i \sigma_j + \frac{\mu}{2N} \sigma_i \sigma_j \sigma_i^* \sigma_j^* \right] - \sum_{i \in \Lambda} \left[ \sqrt{h} \tilde{z}_i \sigma_i + h \sigma_i \sigma_i^* \right]$$

This Hamiltonian has a  $\mathbb{Z}_2$ -gauge symmetry:

$$z_{ij} \mapsto z_{ij} \sigma_i^* \sigma_j^*, \quad \sigma_i \sigma_j \mapsto \sigma_i \sigma_i^* \sigma_j \sigma_j^*$$

that allows us to get rid of the ground truth in the Hamiltonian:

$$\begin{aligned} H_N(\sigma, \sigma^*, z) &= - \sum_{i,j \in \Lambda} \left[ \sqrt{\frac{\mu}{2N}} z_{ij} \sigma_i \sigma_j + \frac{\mu}{2N} \sigma_i \sigma_j \right] - \sum_{i \in \Lambda} \left[ \sqrt{h} \tilde{z}_i \sigma_i + h \sigma_i \right] \\ &\stackrel{D}{=} - \sum_{i,j \in \Lambda} J_{ij} \sigma_i \sigma_j - \sum_{i \in \Lambda} h_i \sigma_i, \quad J_{ij} \stackrel{\text{iid}}{\sim} \mathcal{N} \left( \frac{\mu}{2N}, \frac{\mu}{2N} \right), h_i \stackrel{\text{iid}}{\sim} \mathcal{N}(h, h). \end{aligned}$$

The inference is performed in the Bayesian Optimal Setting, meaning that the prior distribution of the spins is known, and this gives rise to the Nishimori identities that ease computations.

# Thermodynamic quantities

Let  $f$  be a function of 2 *replicas*, then we define the Boltzmann-Gibbs (BG) replicated average for a disordered model as

$$\langle f \rangle_N = \frac{\sum_{\sigma, \tau \in \Sigma_N} f(\sigma, \tau) e^{-H_N(\sigma) - H_N(\tau)}}{(\sum_{\lambda} e^{-H_N(\lambda)})^2}$$

Notice that  $\sigma$  and  $\tau$  are sampled independently w.r.t. to the BG measure, but the disorder in  $H_N$  is the same. The goal is generally the computation of the limiting average quenched pressure per particle

$$\lim_{N \rightarrow \infty} p_N = \lim_{N \rightarrow \infty} \frac{1}{N} \log \sum_{\sigma \in \Sigma_N} e^{-H_N(\sigma)} \stackrel{\text{a.s.}}{=} \lim_{N \rightarrow \infty} \frac{1}{N} \mathbb{E} \log \sum_{\sigma \in \Sigma_N} e^{-H_N(\sigma)}$$

if it exists. The a.s. convergence is guaranteed by the fact that  $p_N$  is a Lipschitz function of standard Gaussian r.v.s.

–  $p_N$  is the mutual information  $I(\sigma^*, (y, \tilde{y}))/N$  of the WS model up to a constant.

# The Nishimori line (NL)

For a spin-glass with couplings  $J_{ij} \stackrel{\text{iid}}{\sim} \mathcal{N}(\mu_{ij}, \Delta_{ij}^2)$  and biases  $h_i \stackrel{\text{iid}}{\sim} \mathcal{N}(\mu_i, \Delta_i^2)$  the Nishimori line is the subregion of the phase space  $(\mu_{ij}, \Delta_{ij}^2; \mu_i, \Delta_i^2)$  where

$$\mu_{ij} = \Delta_{ij}^2, \quad \mu_i = \Delta_i^2 \quad \forall i, j \in \Lambda.$$

Therefore we have just seen that the WS model corresponds to a spin-glass model called SK model on the NL

$$H_N(\sigma) = - \sum_{i,j \in \Lambda} J_{ij} \sigma_i \sigma_j - \sum_{i \in \Lambda} h_i \sigma_i, \quad J_{ij} \stackrel{\text{iid}}{\sim} \mathcal{N}\left(\frac{\mu}{2N}, \frac{\mu}{2N}\right), \quad h_i \stackrel{\text{iid}}{\sim} \mathcal{N}(h, h).$$

As it happens in inference, a model on the Nishimori line (or in the Bayesian Optimal Setting) is greatly constrained and some identities on the correlation functions arise, for instance:

$$\mathbb{E}\langle \sigma_i \rangle_N^2 = \mathbb{E}\langle \sigma_i \tau_i \rangle_N = \mathbb{E}\langle \sigma_i \rangle_N, \quad \mathbb{E}\langle \sigma_i \sigma_j \rangle_N^2 = \mathbb{E}\langle \sigma_i \sigma_j \rangle_N.$$

# Type I and II correlation inequalities on the NL

For a model on the NL we have:

Theorem (General case by Contucci, Morita, Nishimori '05)

$$\frac{\partial \mathbb{E} p_N}{\partial \mu_i} = \frac{1}{2} [1 + \mathbb{E} \langle \sigma_i \rangle_N] > 0$$

$$\frac{\partial^2 \mathbb{E} p_N}{\partial \mu_i \partial \mu_j} = \frac{1}{2} \frac{\partial}{\partial \mu_j} \mathbb{E} \langle \sigma_i \rangle_N = \mathbb{E} [(\langle \sigma_i \sigma_j \rangle_N - \langle \sigma_i \rangle_N \langle \sigma_j \rangle_N)^2] \geq 0.$$

Analogous results hold for  $J_{ij}$ -derivatives and mixed  $\mu_i$ ,  $J_{jk}$ -derivatives.

Fundamental observation:

$$\underline{\mathbb{E} \langle \sigma_i \rangle_N \text{ is non decreasing in the parameters } \mu_j.}$$

Correlation inequalities and the Nishimori identities together are sufficient to force replica symmetry in our model.

# WS and SK compared

Both models can be solved comparing them with decoupled systems by means of an *adaptive interpolation*.

Theorem (J. Barbier, N. Macris '19)

Let the signal prior  $P_0$  be bounded with second moment  $\rho$ . The mutual information for the spiked Wigner model verifies

$$\lim_{n \rightarrow +\infty} \frac{1}{n} I(\mathbf{x}^*; \mathbf{Y}) = \inf_{q \in [0, \rho]} \left\{ \frac{\mu}{4} (q^2 + \rho^2) + \right. \\ \left. - \mathbb{E} \log \int dP_0(x) \exp \left[ \mu q x X^* + \sqrt{\mu q} Z x - \frac{\mu q}{2} x^2 \right] \right\}, \quad Z \sim \mathcal{N}(0, 1).$$

Theorem

The quenched pressure of the planted SK model in the thermodynamic limit is:

$$\lim_{N \rightarrow \infty} \mathbb{E} p_N = \sup_{q \in [0, 1]} \left\{ \mu \frac{(1-q)^2}{4} - \mu \frac{q^2}{2} + \mathbb{E} \log 2 \cosh(z \sqrt{\mu q} + \mu q) \right\}, \quad z \sim \mathcal{N}(0, 1)$$



# Adding structure

The model defined by the Hamiltonian

$$H_N(\sigma) = - \sum_{i,j \in \Lambda} J_{ij} \sigma_i \sigma_j - \sum_{i \in \Lambda} h_i \sigma_i, \quad J_{ij} \stackrel{\text{iid}}{\sim} \mathcal{N} \left( \frac{\mu}{2N}, \frac{\mu}{2N} \right), \quad h_i \stackrel{\text{iid}}{\sim} \mathcal{N}(h, h)$$

fulfills a complete permutation symmetry in distribution among the spins.

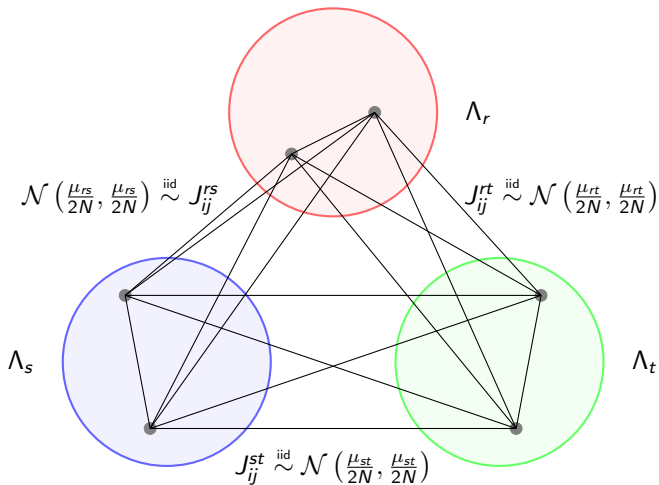
Let us partition  $\Lambda$  into  $K$  disjoint subsets:  $\Lambda = \uplus_{r=1}^K \Lambda_r$ ,  $|\Lambda_r| = N_r = \alpha_r N$ . Now we can set the distribution of the  $J_{ij}$ ,  $h_i$  according to which partition  $i, j$  belong to:

$$H_N(\sigma) := - \sum_{r,s=1}^K \sum_{(i,j) \in \Lambda_r \times \Lambda_s} J_{ij}^{rs} \sigma_i \sigma_j - \sum_{r=1}^K \sum_{i \in \Lambda_r} h_i^r \sigma_i,$$

$$J_{ij}^{rs} \stackrel{\text{iid}}{\sim} \mathcal{N} \left( \frac{\mu_{rs}}{2N}, \frac{\mu_{rs}}{2N} \right), \quad h_i^r \stackrel{\text{iid}}{\sim} \mathcal{N}(h_r, h_r), \quad (i, j) \in \Lambda_r \times \Lambda_s$$

with  $\mu_{rs} = \mu_{sr}$ . The previous is the Hamiltonian of the so called multi-species SK model.

The permutation symmetry is now preserved only inside each partition!



This generalization has added structure to the model making it able to capture the thermodynamic ( $N \rightarrow \infty$ ) behaviour of a system where different populations of particles interact.

# The convex case: the replica symmetric formula

Surprisingly, this disordered model is replica symmetric.

Theorem (Alberici, C., Contucci, Mingione '20)

Let  $\alpha = \text{diag}(\alpha_1, \dots, \alpha_K)$  and  $\Delta := \alpha\mu\alpha$  with  $\mu \geq 0$ . The thermodynamic limit of the pressure  $\bar{p}(\mu, h) := \lim_{N \rightarrow \infty} \bar{p}_N(\mu, h)$  exists and:

$$\bar{p}(\mu, h) = \sup_{\mathbf{x} \in \mathbb{R}_{\geq 0}^K} \bar{p}(\mu, h; \mathbf{x})$$

where

$$\bar{p}(\mu, h; \mathbf{x}) := \frac{(1 - \mathbf{x}, \Delta(1 - \mathbf{x}))}{4} - \frac{(\mathbf{x}, \Delta\mathbf{x})}{2} + \sum_{r=1}^K \alpha_r \psi((\mu\alpha\mathbf{x} + \mathbf{h})_r)$$

$$\psi(h) = \mathbb{E} \log \cosh(z\sqrt{h} + h)$$

with the following stationary condition:

$$\mathbf{x} - \mathbb{E}_z \tanh\left(z\sqrt{\mu\alpha\mathbf{x} + \mathbf{h}} + \mu\alpha\mathbf{x} + \mathbf{h}\right) \in \text{Ker}\Delta, \quad z \sim \mathcal{N}(0, 1)$$

# Phase transition

## Proposition (Alberici, C., Contucci, Mingione '20)

Let  $\mu > 0$ . Denote by  $\rho(A)$  the spectral radius of a matrix  $A$  and by  $\mathcal{H}_{\mathbf{x}}\bar{p}$  the Hessian matrix of  $\bar{p}$ . The following implication holds:

$$\rho(\mu\alpha) = \rho(\alpha^{-1}\Delta) < 1 \quad \Rightarrow \quad \mathcal{H}_{\mathbf{x}}\bar{p}(\mu, h; \mathbf{x}) < 0, \quad \forall \mathbf{x} \in \mathbb{R}_{\geq 0}^K$$

or equivalently  $\bar{p}(\mu, h; \mathbf{x})$  is strictly concave w.r.t.  $\mathbf{x}$ .

Furthermore

$$\mathcal{H}_{\mathbf{x}}\bar{p}(\mu, 0; 0) = \frac{1}{2}\Delta^{1/2} \left[ -\mathbb{1} + \Delta^{1/2}\alpha^{-1}\Delta^{1/2} \right] \Delta^{1/2}$$

therefore whenever  $\mathbf{h} = 0$  and  $\rho(\mu\alpha) < 1$ ,  $\mathbf{x} = 0$  is the unique maximizer.  $\mathbf{x} = 0$  becomes unstable as soon as  $\rho(\mu\alpha) > 1$ .

# Critical exponents for the monopartite case

Proposition (Alberici, C., Contucci, Mingione '20)

Define:

$$\bar{p}_{var}(x; \mu, h) = \mu \frac{(1-x)^2}{4} - \frac{\mu x^2}{2} + \mathbb{E} \log 2 \cosh(z\sqrt{\mu x} + \mu x) .$$

The following hold:

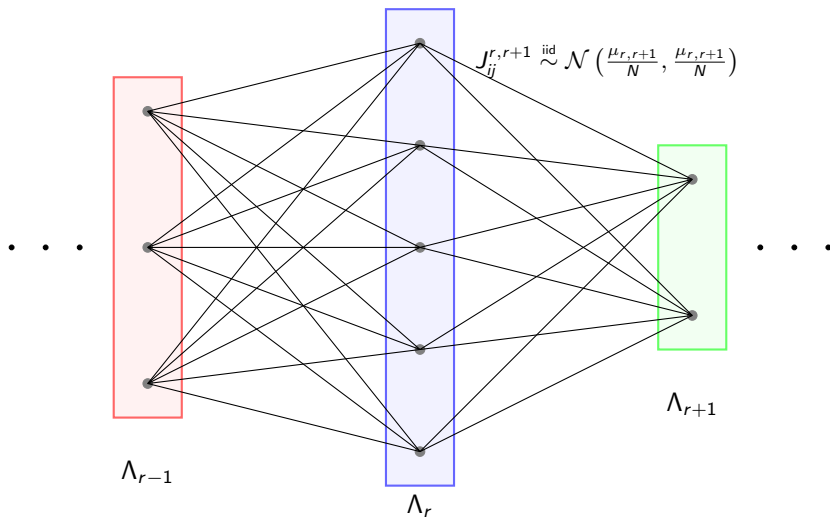
- 1 if  $\mu < 1$  then  $\bar{p}_{var}$  is concave in  $x$  and if further  $h = 0$  then  $x = 0$  the unique maximum point;
- 2 denoting the maximum point by  $\bar{x}(\mu, h)$ :  $\lim_{(\mu, h) \rightarrow (1, 0)} \bar{x}(\mu, h) = 0 = \bar{x}(1, 0)$ ;
- 3 for fixed  $h = 0$  we have  $\bar{x} = (1 + o(1)) \frac{\mu-1}{\mu^2}$  where  $o(1)$  goes to 0 when  $\mu \rightarrow 1_+$ . Therefore the critical exponent  $\beta = 1$ .
- 4 For fixed  $\mu = 1$  and  $h \rightarrow 0_+$  the magnetization behaves as  $\bar{x}^2 = h(1 + o(1))$  where  $o(1) \rightarrow 0$  when  $h \rightarrow 0_+$ . Therefore we have a critical exponent  $\delta = 2$ .

# Deep restricted Boltzmann Machine on the NL

In this model the partitions are rearranged in a consecutive way and only inter-partition interactions are allowed (restricted). It corresponds to the choice

$$\mu = \begin{pmatrix} 0 & \mu_{12} & 0 & \cdots & 0 \\ \mu_{21} & 0 & \mu_{23} & \cdots & 0 \\ 0 & \mu_{32} & 0 & \ddots & 0 \\ \vdots & \vdots & \ddots & \ddots & \mu_{K-1,K} \\ 0 & 0 & 0 & \mu_{K,K-1} & 0 \end{pmatrix}.$$

$\mu$  has eigenvalues with alternating sign (symmetric w.r.t. 0). For centered interactions the solution to this problem is unknown though we have some bounds.



Even the simplest  $K = 2$  bipartite case for centered interactions is still unsolved.

# Replica symmetric formula for the DBM

## Theorem (Alberici, C, Contucci, Mingione '20)

Consider  $J_{ij}^{rs} \stackrel{\text{iid}}{\sim} \mathcal{N}\left(\frac{\mu_{rs}}{2N}, \frac{\mu_{rs}}{2N}\right)$ ,  $h_i^r \stackrel{\text{iid}}{\sim} \mathcal{N}(h_r, h_r)$  for  $(i, j) \in \Lambda_r \times \Lambda_s$ . Denote by  $\mathbf{x}_o$  the odd components of a vector  $\mathbf{x}$ . Similarly for  $\mathbf{x}_e$ .

$$\lim_{N \rightarrow \infty} p_N \stackrel{\text{a.s.}}{=} \lim_{N \rightarrow \infty} \mathbb{E} p_N = \sup_{\mathbf{x}_o} \inf_{\mathbf{x}_e} \bar{p}(\mathbf{x}; \mu, \mathbf{h}),$$

$$\bar{p}(\mathbf{x}; \mu, \mathbf{h}) := \sum_{r=1}^K \alpha_r \psi((\mu \alpha \mathbf{x})_r + h_r) + \sum_{r=1}^K \frac{\Delta_{r,r+1}}{2} [(1 - x_r)(1 - x_{r+1}) - 2x_r x_{r+1}]$$

where

$$\psi(x) := \mathbb{E}_z \log 2 \cosh(z\sqrt{x} + x), \quad z \sim \mathcal{N}(0, 1).$$



# Phase transition

Denote by  $A^{(\circ\circ)}$  the submatrix of  $A$  obtained by erasing its even rows and even columns.

## Theorem

*Let  $K$  be even and  $\mathbf{h} = 0$ . If  $\rho([\mu\alpha]^2)^{(\circ\circ)} < 1$  then  $\mathbf{x} = 0$  is the unique solution to the variational problem. Conversely, if  $\rho([\mu\alpha]^2)^{(\circ\circ)} > 1$  then the solution is a vector  $\mathbf{x} = \bar{\mathbf{x}}(\mu\alpha)$  with strictly positive components satisfying the consistency equation:*

$$x_r = \mathbb{E}_z \tanh \left( z \sqrt{(\mu\alpha\mathbf{x})_r} + (\mu\alpha\mathbf{x})_r \right) \quad \forall r = 1, \dots, K.$$

# Geometry and phase transition

## Proposition

The maximum of the spectral radius of  $[(\mu\alpha)^2]^{(\circ\circ)}$  over  $\alpha_1, \dots, \alpha_K \geq 0$ ,  $\sum_r \alpha_r = 1$ , equals  $\frac{1}{4} \max_r \mu_{r,r+1}^2$  and is reached if and only if:

$$(a) \quad \alpha_{r^*} = \alpha_{r^*+1} = \frac{1}{2} \quad \text{for } r^* \in \arg \max_r \Delta_{r,r+1}, \text{ or:}$$

$$(b) \quad \alpha_{r^*-1} + \alpha_{r^*+1} = \alpha_{r^*} = \frac{1}{2} \quad \text{for } r^*, r^*-1 \in \arg \max_r \Delta_{r,r+1}.$$

## Remark

$$\mu_{r,r+1} < 2 \quad \forall r = 1, \dots, K-1 \quad \Rightarrow \quad \bar{\mathbf{x}}(\mu\alpha) = \lim_{N \rightarrow \infty} \mathbb{E} \langle \mathbf{m} \rangle_N = \lim_{N \rightarrow \infty} \mathbb{E} \langle \mathbf{q}_{12} \rangle_N = 0$$

since there is no way to rearrange the spins into the partitions  $(\alpha_r$ 's) to obtain  $\rho([( \mu\alpha)^2]^{(\circ\circ)}) > 1$ .

