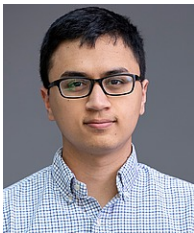


A constructive proof of the spherical Parisi formula

Brice Huang (MIT)

BIRS Workshop in Computational Complexity of Statistical Inference
Joint work with Mark Sellke (Harvard)

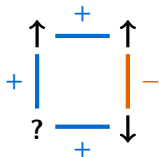


Mean-field spin glasses

- Random polynomial over $\{\pm 1\}^N$ or $S_N = \{\sigma \in \mathbb{R}^N : \|\sigma\| = \sqrt{N}\}$
- Sherrington-Kirkpatrick Model:

$$H_N(\sigma) = \frac{1}{\sqrt{N}} \sum_{i,j=1}^N g_{i,j} \sigma_i \sigma_j \quad \sigma \in \{\pm 1\}^N \quad g_{i,j} \stackrel{i.i.d.}{\sim} \mathcal{N}(0, 1)$$

- Ground state energy: typical max of $H_N(\sigma)/N$
- Random couplings $g_{i,j}$ lead to highly non-trivial behavior



Mixed p -spin models

- p -spin model: (SK: $p = 2$)

$$H_N(\sigma) = \frac{1}{N^{(p-1)/2}} \sum_{\underline{i} \in [M]^p} g_{\underline{i}}(\sigma^{\otimes p})_{\underline{i}} \quad g_{\underline{i}} \stackrel{i.i.d.}{\sim} \mathcal{N}(0, 1)$$

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- (Ising / spherical) mixed p -spin model: linear mixture of above

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- Connections to tensor PCA, random max- k -SAT, high-dim inference, ...
(Ben Arous–Mei–Montanari–Nica 17, Dembo–Montanari–Sen 17, Panchenko 18,
Fan–Mei–Montanari 21)

Free energy, Gibbs measure

- Free energy: softmax version of ground state. Let

$$Z_N = \int_{S_N} e^{\beta H_N(\sigma)} d\sigma \quad (\text{partition function})$$

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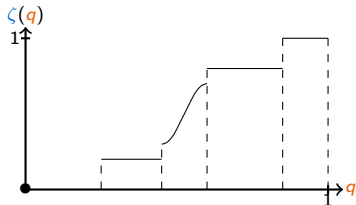
- Gibbs measure:

$$d\mu_N(\sigma) = \frac{1}{Z_N} e^{H_N(\sigma)} d\sigma$$

Limiting behavior of μ_N ? E.g. for replicas $\sigma^1, \sigma^2 \sim \mu_N$, what is the law of the **overlap** $\frac{\langle \sigma^1, \sigma^2 \rangle}{N} \in [-1, 1]$?

Physics predictions (now theorems)

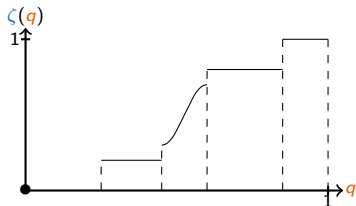
- Order parameter: probability measure ζ on $[0, 1]$; identify with CDF:



Overlap distribution: $\zeta_* = \lim_{N \rightarrow \infty} \text{Law}(\langle \sigma^1, \sigma^2 \rangle / N)$

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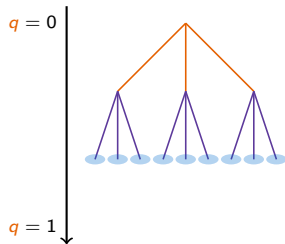
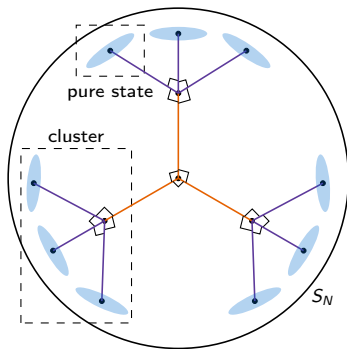
- Parisi formula:**

$$\lim_{N \rightarrow \infty} \frac{1}{N} \log Z_N = \min_{\zeta} \mathcal{P}(\zeta; \xi) = \mathcal{P}(\zeta_*; \xi)$$

where \mathcal{P} = Parisi functional, and recall ξ = model

Ultrametricity

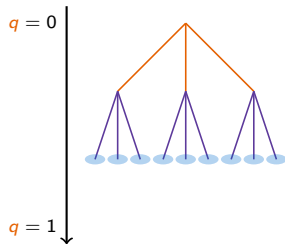
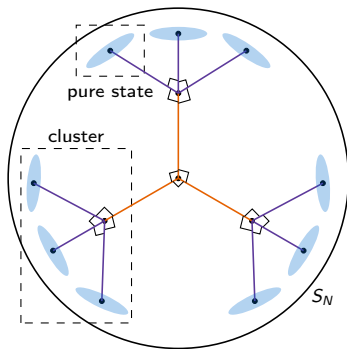
- μ_N convex combination of **pure states** with **hierarchical clustering**
Parisi 83 (heuristic), Panchenko 13 (rigorous)



Ultrametricity

- μ_N convex combination of **pure states** with **hierarchical clustering**

Parisi 83 (heuristic), Panchenko 13 (rigorous)



- Tree branches at radii in $\text{supp } \zeta_*$ (cts support \rightarrow cts branching, **full RSB**)

History of rigorous results

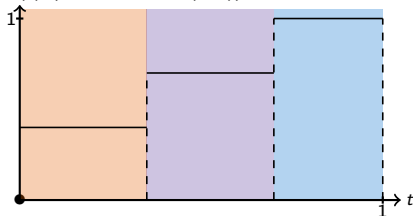
For both Ising, spherical mixed p -spin models:

- Guerra 03: $\limsup_{N \rightarrow \infty} \frac{1}{N} \log Z_N \leq \mathcal{P}(\zeta_*; \xi)$ by interpolation argument
- Talagrand 06: matching lower bound, proves Parisi formula
 - **Analytic proof**, self-bounds the error in Guerra's upper bound
- Panchenko 13: ultrametricity of asymptotic Gibbs measures μ_N
 - \Rightarrow New proof of Parisi formula
 - **Inductive proof on # spins**, focuses on understanding Gibbs measure precisely
- Jagannath 17, Subag 18, Chatterjee–Slooman 21: approximate ultrametricity for finite N
- Jagannath–Tobasco 18: ζ_* finitely many atomic & continuous pieces, in spherical models

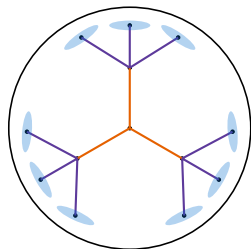
Main result

- New proof of Parisi lower bound for spherical mixed p -spin models:
$$\liminf_{N \rightarrow \infty} \frac{1}{N} \log Z_N \geq \mathcal{P}(\zeta_*; \xi)$$
- **Geometric approach:** directly constructs ultrametric tree of pure states in accordance with Parisi ansatz

$\zeta_*(t)$ (minimizer of $\mathcal{P}(\zeta; \xi)$)



\Rightarrow



Spherical mixed p -spin model

- Recall model: $\xi = (\beta_p)_{p \geq 1}$,

$$H_N(\sigma) = \sum_{p \geq 1} \frac{\beta_p}{N^{(p-1)/2}} \sum_{i \in [N]^p} g_i(\sigma^{\otimes p})_i \quad g_i \stackrel{i.i.d.}{\sim} \mathcal{N}(0, 1)$$

- Domain: $S_N = \{\sigma \in \mathbb{R}^N : \|\sigma\| = \sqrt{N}\}$
- Free energy:

$$\frac{1}{N} \log \int_{S_N} e^{H_N(\sigma)} d\sigma$$

Existence of ultrametric tree

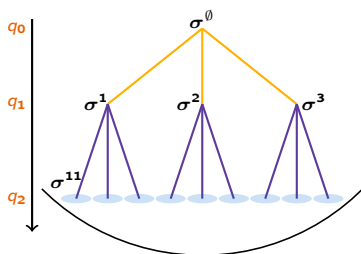
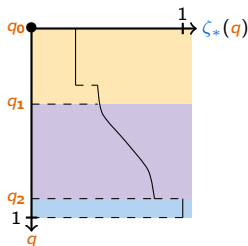
Theorem (H.-Sellke 23)

Let $q_0 < \dots < q_D \in \text{supp } \zeta_*$, $q_D = \max \text{supp } \zeta_*$.

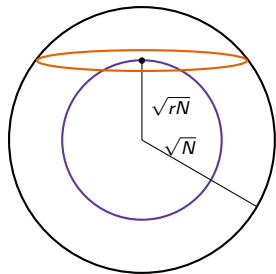
$\mathbb{T}_{k,D}$ k -ary tree of depth D . Whp, exists ultrametric $\{\sigma^u : u \in \mathbb{T}_{k,D}\}$ such that each σ^u , u leaf, is the center of a band B with

$$\frac{1}{N} \log \int_B e^{H_N(\sigma)} d\sigma \geq \mathcal{P}(\zeta_*; \xi) - \varepsilon.$$

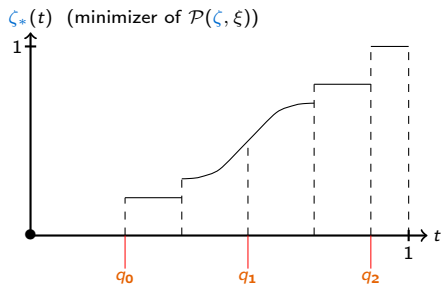
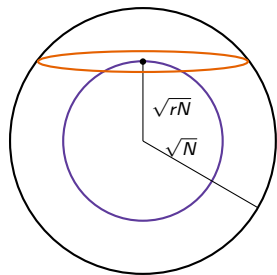
\Rightarrow New proof of Parisi LB!



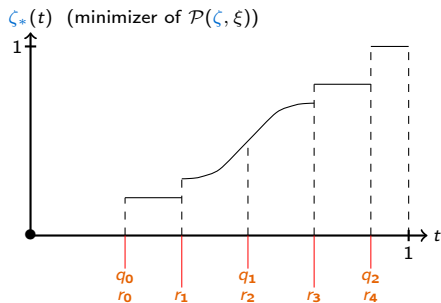
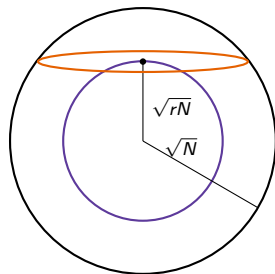
Proof idea 1: Decomposition into fundamental types



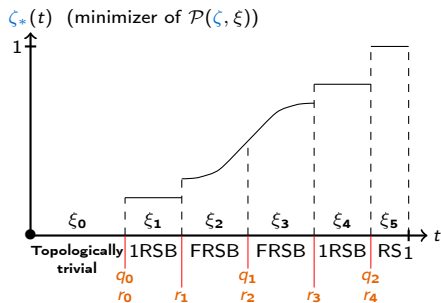
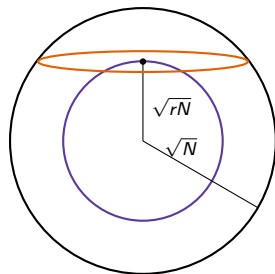
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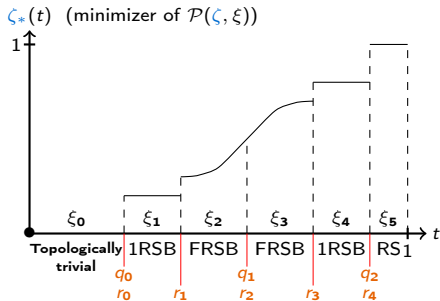
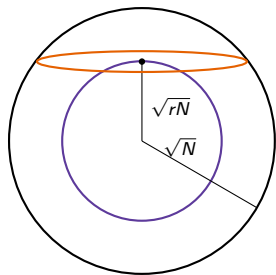
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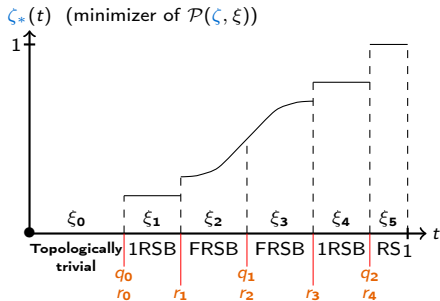
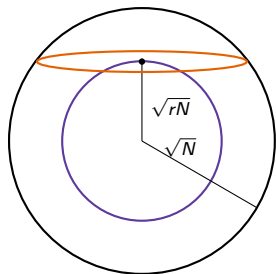
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By following maxima of successive sub-models, can show: (Subag 18)

$$FE(\xi) \geq GSE(\xi_0) + \dots + GSE(\xi_{D-1}) + FE(\xi_D)$$

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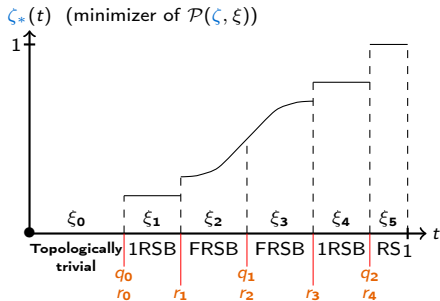
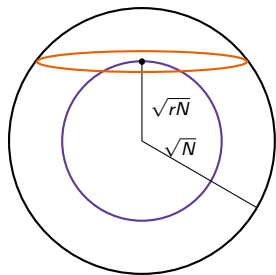
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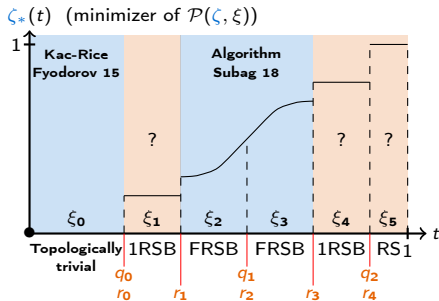
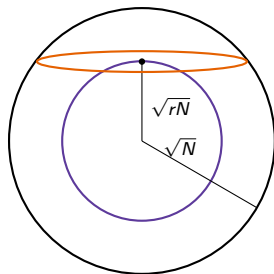
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Proof idea 2: Truncated 2nd moment method

- For RS models, Parisi LB amounts to showing

$$Z_N = \int_{S_N} e^{H_N(\sigma)} d\sigma \geq e^{-o(N)} \mathbb{E} Z_N$$

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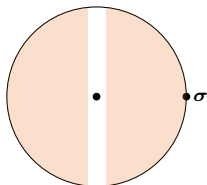
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Truncation: σ **typical** if $\{\rho : \frac{1}{N} |\langle \rho, \sigma \rangle| \geq \delta\}$ accounts for $\leq e^{-\varepsilon N}$ of Z_N



Do 2nd moment on truncation: $\tilde{Z}_N = \int_{\sigma \text{ typical}} e^{H_N(\sigma)} d\sigma$

Proof idea 2: Truncated 2nd moment method (details)

- Truncated partition fn $\tilde{Z}_N = \int_{\sigma \text{ typical}} e^{H_N(\sigma)} d\sigma$
 $\mathbb{E}[\tilde{Z}_N^2] \approx \mathbb{E}[Z_N]^2$ automatic; main work is to show $\mathbb{E}[\tilde{Z}_N] \approx \mathbb{E}[Z_N]$

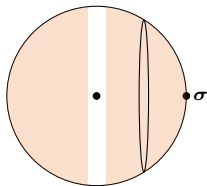
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Guerra's upper bound controls FE of each non-equatorial band:



\Rightarrow conditional on $H_N(\sigma) \approx E_* N$, σ is whp typical

Conclusion

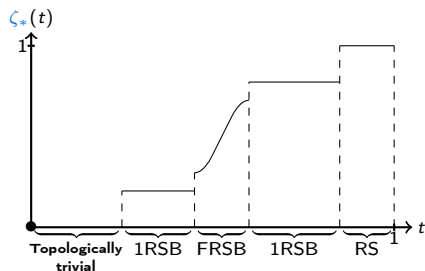
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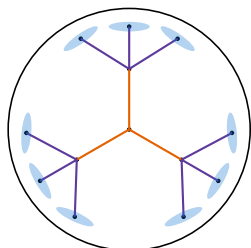
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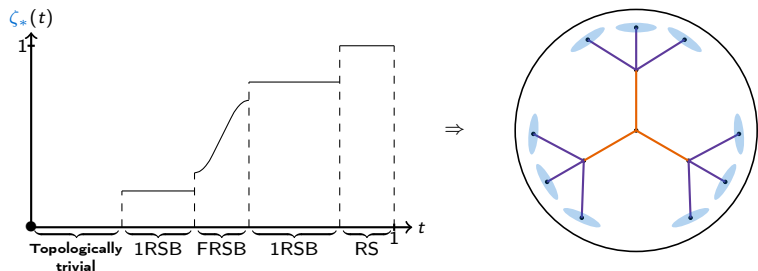


\Rightarrow



Conclusion

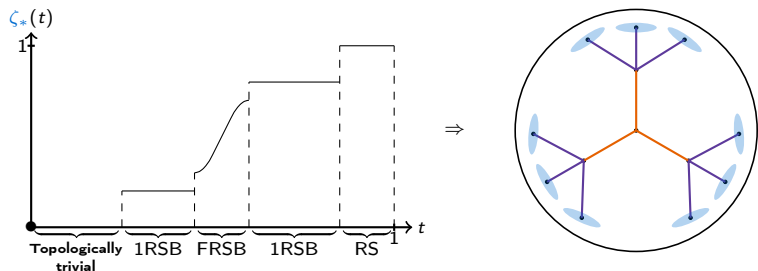
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Thanks!