

Geometry and temperature chaos in some spherical spin glasses

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Based on an ongoing work with Gerard Ben Arous and Ofer Zeitouni

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$$H_{N,p}(\mathbf{x}) = \sqrt{N} \sum_{i_1, \dots, i_p=1}^N J_{i_1, \dots, i_p} x_{i_1} x_{i_2} \cdots x_{i_p}, \quad \mathbf{x} \in \mathbb{S}^N.$$

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- **Mixed models:** $\gamma_p \geq 0$ – real parameters, $\nu(t) = \sum_p \gamma_p^2 t^p$.

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- **Covariance:** $\mathbb{E}\{H_{N,\nu}(\mathbf{x})H_{N,\nu}(\mathbf{x}')\} = N\nu(\langle \mathbf{x}, \mathbf{x}' \rangle)$.

The Gibbs measure – general results

Gibbs meas.: $G_{N,\beta}(A) = \frac{1}{Z_{N,\beta}} \int_A e^{\beta H_N(\mathbf{x})} d\mathbf{x}, \quad A \subset \mathbb{S}^N, \beta \geq 0.$

Partition function: $Z_{N,\beta} = \int_{\mathbb{S}^N} e^{\beta H_N(\mathbf{x})} d\mathbf{x}.$

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Ultrametricity: (Panchenko '13)

$$\forall \epsilon > 0 : G_{N,\beta}^{\otimes 3} \{d(\mathbf{x}_1, \mathbf{x}_3) \leq d(\mathbf{x}_1, \mathbf{x}_2) \vee d(\mathbf{x}_2, \mathbf{x}_3) + \epsilon\} \xrightarrow[N \rightarrow \infty]{P} 1.$$

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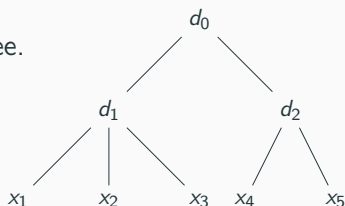
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Remark: any finite ultrametric space $\{x_1, \dots, x_k\}$ can be represented by a tree.



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Cluster decomposition: (Talagrand '10, Jagannath '14)

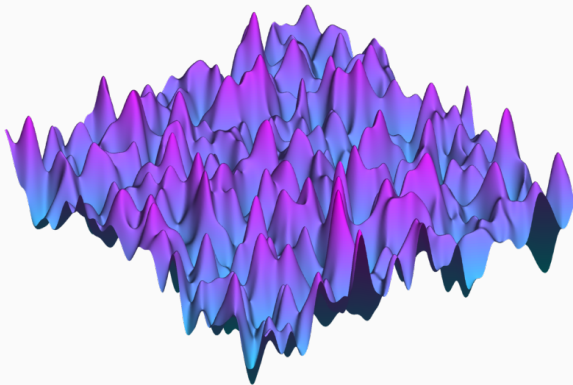
There exist disjoint random $A_i \subset \mathbb{S}^N$ such that:

1. $G_{N,\beta}\{\cup A_i\} \rightarrow 1,$
2. $\{G_{N,\beta}\{A_i\}\}_i$ converges in distribution to some $W_i > 0,$
3. $\exists t_{ij}$ random, s.t conditional on $\mathbf{x}_1 \in A_i, \mathbf{x}_2 \in A_j,$

$$G_{N,\beta}^{\otimes 2} \{|d(\mathbf{x}_1, \mathbf{x}_2) - t_{ij}| > \epsilon\} \rightarrow 0.$$

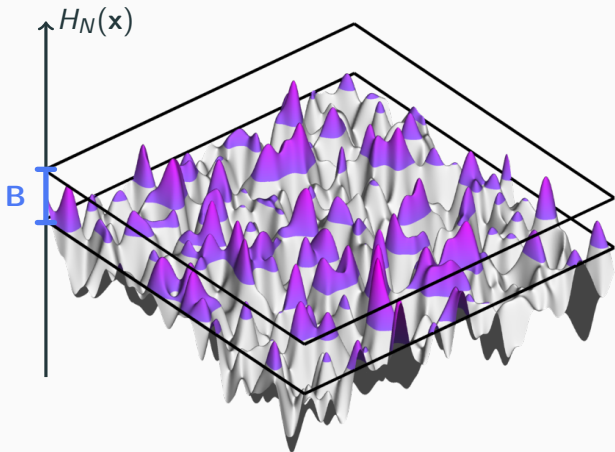
Critical points and the Gibbs measure, pure models

Critical points



Definition. $\mathbf{x} \in \mathbb{S}^N$ is critical if $\nabla H_N(\mathbf{x}) = 0$.

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For $B \subset \mathbb{R}$, $\text{Crit}(B) := \#$ critical values of $H_N(\mathbf{x})$ in B .

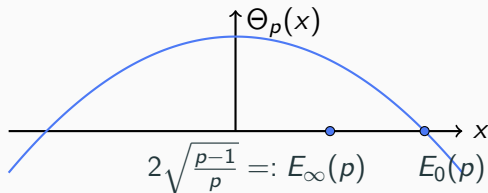
Critical points

Theorem (Auffinger–Ben Arous–Černý '13)

$$\lim_{N \rightarrow \infty} \frac{1}{N} \log (\mathbb{E} \{ \text{Crit}(NB) \}) = \sup_{x \in B} \Theta_p(x).$$

Theorem (S. '15)

For $B \subset (E_\infty, E_0)$, $\frac{\text{Crit}(NB)}{\mathbb{E} \{ \text{Crit}(NB) \}} \rightarrow 1$, in prob.



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Lemma (S. '15)

$B_N \subset (E_\infty, E_0)$, $|B_N| \rightarrow 0 \implies \mathbb{P} \{ |\langle \mathbf{x}_1, \mathbf{x}_2 \rangle| > \epsilon \} \rightarrow 0$,
if $\mathbf{x}_1 \neq \mathbf{x}_2$ are random (unif.) crt. pts. such that $H_N(\mathbf{x}_i) \in NB_N$.

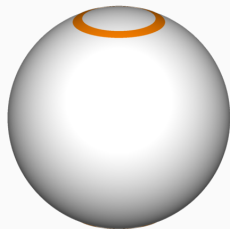
Theorem (S.–Zeitouni '16)

The critical values near $\mathbb{E} \max_{\mathbf{x}} H_N(\mathbf{x})$ converge to $\text{PPP}(e^{c_p x} dx)$.

The Gibbs measure

- Fix some sequence $N^{-\frac{1}{2}} \ll t_N \ll 1$.
 $q_* = q_*(\beta)$ explicit.

$$\text{Band}(\mathbf{x}_0) := \left\{ \mathbf{x} \in \mathbb{S}^N : q_* - t_N \leq \langle \mathbf{x}, \mathbf{x}_0 \rangle \leq q_* + t_N \right\}.$$



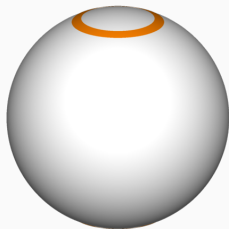
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- Enumerate the crt. pts. \mathbf{x}_0^i , $i \geq 1$,
in decreasing order:

$$H_N(\mathbf{x}_0^i) \geq H_N(\mathbf{x}_0^{i+1}).$$



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1. Asymptotic support: $\forall \epsilon > 0$, for large k, N ,

$$\mathbb{P} \left\{ G_{N,\beta} \left(\cup_{i \leq k} \text{Band}(\mathbf{x}_0^i) \right) > 1 - \epsilon \right\} \geq 1 - \epsilon.$$

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2. Suppose $\mathbf{x}_1, \mathbf{x}_2$ drawn independently from the Gibbs measure.

Then, with probability going to 1, for small $\epsilon > 0$

$$|\langle \mathbf{x}_1, \mathbf{x}_2 \rangle| < \epsilon \iff \mathbf{x}_1, \mathbf{x}_2 \in \text{different bands},$$

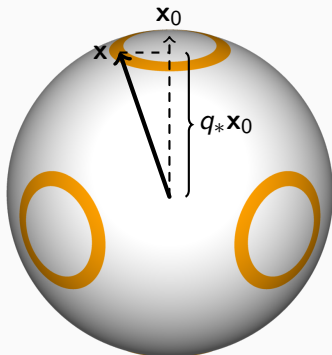
$$|\langle \mathbf{x}_1, \mathbf{x}_2 \rangle - q_*^2| < \epsilon \iff \mathbf{x}_1, \mathbf{x}_2 \in \text{same band}.$$

Overlap distribution

- Given a sample $\mathbf{x} \in \text{Band}(\mathbf{x}_0)$,

$$\mathbf{x} \approx q_* \mathbf{x}_0 + \mathbf{x}^\perp,$$

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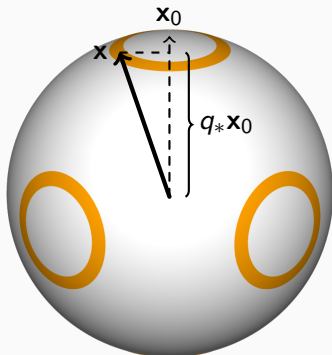
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- For samples \mathbf{x}, \mathbf{y} w.h.p

$$\begin{aligned} \langle \mathbf{x}, \mathbf{y} \rangle &\approx \langle q_* \mathbf{x}_0, q_* \mathbf{y}_0 \rangle \\ &= q_*^2 \langle \mathbf{x}_0, \mathbf{y}_0 \rangle. \end{aligned}$$



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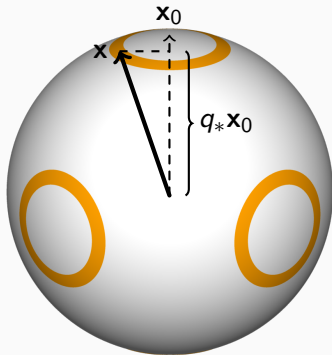
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$$= q_*^2 \langle \mathbf{x}_0, \mathbf{y}_0 \rangle.$$

- Recall: deep crt. pts. are orthogonal w.h.p. Thus,

$$\langle \mathbf{x}_0, \mathbf{y}_0 \rangle \approx 0 \text{ or } 1.$$



Mixed models – work in progress (\w Ben Arous, Zeitouni)

- Reminder: $H_{N,\nu}(\mathbf{x}) = \sum_p \gamma_p H_{N,p}(\mathbf{x}), \quad \nu(x) = \sum \gamma_p x^p.$

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- All the results about critical points hold as in the pure case.
- For large β , the Gibbs measure $G_{N,\beta}$ concentrates on bands around critical points.
- However, the relevant critical points depend on β !
- Compared to the pure case, more complicated structure on bands.
For two samples from the same band $\langle \mathbf{x}_1, \mathbf{x}_2 \rangle > q_*^2$.
But still, for points from different bands $\langle \mathbf{x}_1, \mathbf{x}_2 \rangle \approx 0$.

Temperature chaos

- Let $\beta_1 \neq \beta_2$ be two inverse-temperatures.
- With the disorder $H_N(\mathbf{x})$ fixed, let \mathbf{x}_1 and \mathbf{x}_2 be independent samples from G_{N,β_1} and G_{N,β_2} . (recall: $dG_{N,\beta} \propto e^{\beta H_N(\mathbf{x})} d\mathbf{x}$)

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- Small change in temperature results in a drastic change of the Gibbs measure.

Theorem (Chen–Panchenko '16)

For **even generic models** (under some conditions) temperature chaos **occurs**.

Even: p odd $\implies \gamma_p = 0$, generic: $\sum p^{-1} \mathbf{1}\{\gamma_p \neq 0\} = \infty$.

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Let $p \geq 4$ even, $p' \neq p$, $a \in (0, \frac{1}{4})$. For

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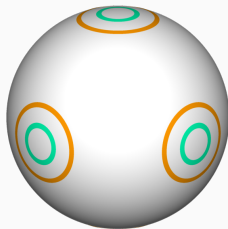
For the **pure spherical models**, $p \geq 3$, with β_1, β_2 large enough, temperature chaos **does not occur**.

Sketch of proof. For large K , both G_{N,β_1} and G_{N,β_2} are essentially supported on the set of bands around $\mathbf{x}_0^1, \dots, \mathbf{x}_0^K$ with $q_*(\beta_1) \neq q_*(\beta_2)$.

As before, if $\mathbf{x} \in \text{Band}_{\beta_1}(\mathbf{x}_0^i)$, $\mathbf{x}' \in \text{Band}_{\beta_2}(\mathbf{x}_0^j)$, then

$$\langle \mathbf{x}, \mathbf{x}' \rangle \approx q_*(\beta_1)q_*(\beta_2)\langle \mathbf{x}_0^i, \mathbf{x}_0^j \rangle.$$

$$\begin{aligned} \implies G_{N,\beta_1} \times G_{N,\beta_2} \{ \langle \mathbf{x}, \mathbf{x}' \rangle \approx q_*(\beta_1)q_*(\beta_2) \} \\ = \sum_i G_{N,\beta_1}(\text{Band}_{\beta_1}(\mathbf{x}_0^i)) G_{N,\beta_2}(\text{Band}_{\beta_2}(\mathbf{x}_0^i)). \end{aligned}$$



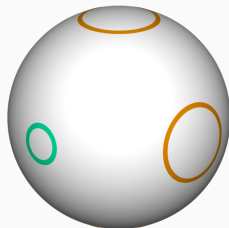
Temperature chaos

On the other hand, for the mixed models (close to pure), the critical points G_{N,β_1} and G_{N,β_2} are supported correspond to different heights $NE_*^{\beta_1}$ and $NE_*^{\beta_2}$ of the Hamiltonian and they are orthogonal.

Therefore, typically for two samples $\mathbf{x} \in \text{Band}_{\beta_1}(\mathbf{x}_0^i)$, $\mathbf{x}' \in \text{Band}_{\beta_2}(\mathbf{x}_0^j)$,

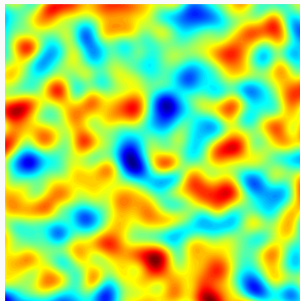
$$\langle \mathbf{x}, \mathbf{x}' \rangle \approx 0,$$

and temperature chaos occurs.



Conditional models on bands

- The Gibbs measure $dG_{N,\beta} \propto e^{\beta H_N(\mathbf{x})} d\mathbf{x}$ mainly charges high values of $H_N(\mathbf{x})$.



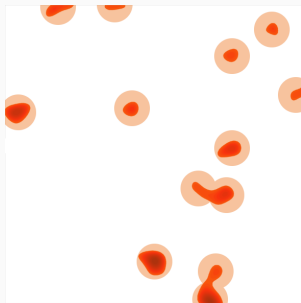
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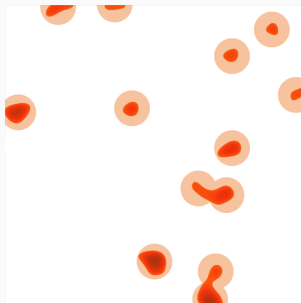
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- The complement is covered by nbhds $\{\mathbf{x} : \langle \mathbf{x}, \mathbf{x}_0 \rangle \geq q_t\}$ of critical pts \mathbf{x}_0 with $H_N(\mathbf{x}_0) \geq NE_t$.



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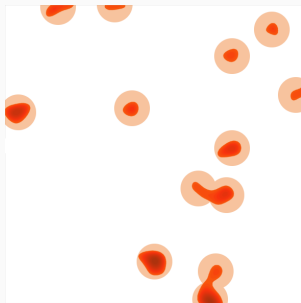


- The latter set is covered by the union, going over all heights $E > E_t$ and 'overlaps' $q \in (q_t, 1)$, of the bands:

$$\text{Band}(\mathbf{x}_0, q) := \{\mathbf{x} : \langle \mathbf{x}, \mathbf{x}_0 \rangle \approx q\}, \quad \mathbf{x}_0 \text{ crit.}, \quad H_N(\mathbf{x}_0) \approx NE.$$

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- To prove that $G_{N,\beta}$ concentrates on bands, we show that the weight of those bands is maximized at log. scale for specific E_*, q_* .

Conditional models on bands

For given height E and 'overlap' q :

- Number of bands $\approx \exp\{N\Theta_\nu(E)\}$.
- Each band has volume $\approx (1 - q^2)^{\frac{N}{2}}$.
- For each critical \mathbf{x}_0 , associate the weight

$$\int e^{\beta H_N|_q(\mathbf{x})} d\mathbf{x},$$

where

$$H_N|_q : \mathbb{S}^{N-1} \rightarrow \mathbb{R}, \quad H_N|_q(\mathbf{x}) = H_N \circ f(\mathbf{x})$$

is the restriction of H_N to the sub-sphere $\{\mathbf{x} : \langle \mathbf{x}, \mathbf{x}_0 \rangle = q\}$, up to a change of coordinates.

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Simplification: assume those weights are i.i.d., with law obtained from conditioning on $\nabla H_N(\mathbf{x}_0) = 0$, $H_N(\mathbf{x}_0) = NE$.

Conditional models on bands – pure case

- For the pure p -spin model,

$$\begin{aligned} H_{N,p|q}(\mathbf{x}) &\stackrel{d}{=} \alpha_{p,0}(q)H_{N,p}(\mathbf{x}_0) \\ &\quad + \alpha_{p,1}(q)\langle \nabla H_{N,p}(\mathbf{x}_0), \mathbf{x} \rangle \\ &\quad + \sum_{k=2}^p \alpha_{p,k}(q)H_{N,k}^{(p)}(\mathbf{x}), \end{aligned}$$

$H_{N,k}^{(p)}$ are copies of pure k -spin models,

$\alpha_{p,k}(q) \in \mathbb{R}$ explicit,

all summands are independent.

- Upon conditioning on $H_{N,p}(\mathbf{x}_0) = NE$, $\nabla H_{N,p}(\mathbf{x}_0) = 0$,

$$H_{N,p|q}(\mathbf{x}) \stackrel{d}{=} \alpha_{p,0}(q)NE + \sum_{k=2}^p \alpha_{p,k}(q)H_{N,k}^{(p)}(\mathbf{x}).$$

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- The weight corresponding to a single band is

$$\int e^{\beta H_{N,p|q}(\mathbf{x})} d\mathbf{x} = e^{N\alpha_{p,0}(q)\beta E} \int e^{\beta \sum_{k=2}^p \alpha_{p,k}(q)H_{N,k}^{(p)}(\mathbf{x})} d\mathbf{x}.$$

- Deterministic, only term that depends on E .
- Distributed like $e^{N(C+W)}$, for $w \neq 0$, $\mathbb{P}\{W \approx w\} = e^{-N^2 J(w)}$, C deterministic – essentially can be replaced by mean.
- For large β always best to choose $E = E_0$, lose number of points and gain in the first term above. $\implies E_* = E_0$.
- Optimize over q to find q_* .

Conditional models on bands – mixed case

- In the mixed case $\left(H_{N,\nu}(\mathbf{x}) = \sum_{p \geq 2} \gamma_p H_{N,p}(\mathbf{x}) \right)$,

$$\begin{aligned} H_{N,\nu}|_q(\mathbf{x}) &\stackrel{d}{=} \sum_{p \geq 2} \gamma_p \alpha_{p,0}(q) H_{N,p}(\mathbf{x}_0) \\ &\quad + \left\langle \sum_{p \geq 2} \gamma_p \alpha_{p,1}(q) \nabla H_{N,p}(\mathbf{x}_0), \mathbf{x} \right\rangle \\ &\quad + \sum_{p \geq 2} \sum_{k=2}^p \gamma_p \alpha_{p,k}(q) H_{N,k}^{(p)}(\mathbf{x}). \end{aligned}$$

- Upon conditioning on

$$\begin{aligned} H_{N,\nu}(\mathbf{x}_0) &= \sum_{p \geq 2} \gamma_p H_{N,p}(\mathbf{x}_0) = NE, \\ \nabla H_{N,\nu}(\mathbf{x}_0) &= \sum_{p \geq 2} \gamma_p \nabla H_{N,p}(\mathbf{x}_0) = 0, \end{aligned}$$

non of the terms becomes deterministic.

- The weight of one band $\approx e^{N(C+W)}$,
 $C > 0$ deterministic, for $w \neq 0$, $\mathbb{P}\{W \approx w\} = e^{-NJ(w)}$.
- However, there are exponentially many points if $E < E_0$, leading to a large deviation type problem.
- The optimal energy E_* turns out to be strictly smaller than E_0 and β dependent.

Disorder chaos – pure models

- Let $H'_N(\mathbf{x})$ be an i.i.d. copy of $H_N(\mathbf{x})$ and for $t \in (0, 1)$ set

$$H_{N,t}(\mathbf{x}) := (1 - t)H_N(\mathbf{x}) + \sqrt{2t - t^2}H'_N(\mathbf{x}).$$

- Let \mathbf{x}_1 and \mathbf{x}_2 be independent samples from $G_{N,\beta}$ and $G_{N,t,\beta}$.

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Theorem (Chen–Hsieh–Hwang–Sheu '15)

Disorder chaos occurs for all spherical models and $t \in (0, 1)$.

- Now, consider $t_N = o(1)$ depending on N .

Disorder chaos

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Theorem (S.–Zeitouni '16)

There exists a random permutation $\sigma : \mathbb{N} \rightarrow \mathbb{N}$ such that for fixed K , for any $i < K$:

1. Location hardly changes: $\|\mathbf{x}_0^i - \mathbf{x}_0^{\sigma(i)}(t_N)\| = o(1)$.

2. Change in value: $H_{N,t_N}(\mathbf{x}_0^{\sigma(i)}(t_N)) = H_N(\mathbf{x}_0^i) + \Delta_i$,

$$\Delta_i := -Nt_N C + \sqrt{Nt_N} \frac{H'_N(\mathbf{x}_0^i)}{\sqrt{N}} + o(1).$$

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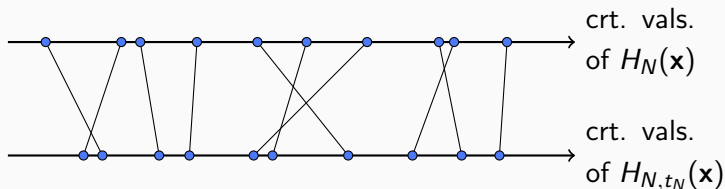
- Bands approximately remain at same position, weights change.

$$\Delta_i = -Nt_N C + \sqrt{Nt_N} \frac{H'_N(\mathbf{x}_0^i)}{\sqrt{N}} + o(1)$$

- For $t_N = c/N$, for any $i \leq K$, w.h.p. $\Delta_i = O(1)$.

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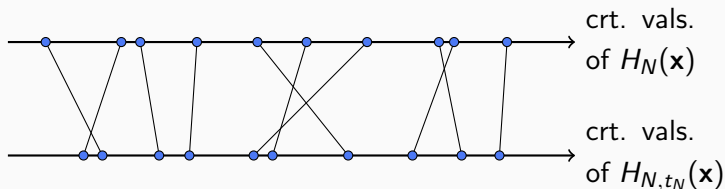
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Disorder chaos

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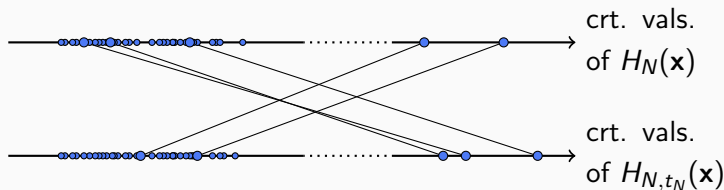
- No disorder chaos.

$$\Delta_i = -Nt_N C + \sqrt{Nt_N} \frac{H'_N(\mathbf{x}_0^i)}{\sqrt{N}} + o(1)$$

- If $t_N = c_N/N$ with $c_N \rightarrow \infty$, for $i \leq K$,

$$Nt_N C \gg \left| \sqrt{Nt_N} \frac{H'_N(\mathbf{x}_0^i)}{\sqrt{N}} \right|.$$

- But $H_N(\mathbf{x}) \stackrel{d}{=} H_{N,t_N}(\mathbf{x})$, not all points are washed away by shift.



- Disorder chaos occurs.

Thank You!