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# Statistics 291: Lecture 8 (February 15th, 2024)

## Kac-Rice IV: topological trivialization

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### 1 Topological trivialization

Before everything, we recap the determinant bounds for  $\text{GOE}(N)$  matrices,

$$\begin{aligned} \frac{1}{N} \log |\det(\text{GOE}(N) - tI_N)| &\approx \psi(t) := \int \log |u - t| d\nu_{sc}(t) \\ &= \frac{t^2}{4} - \frac{1}{2} + \mathbb{1}_{\{|t| \geq 2\}} \left( -\frac{t + \sqrt{t^2 - 4}}{4} + \log \left( \frac{\sqrt{t^2 - 4} + |t|}{2} \right) \right). \end{aligned}$$

**Natural Langevin dynamics.** The concept of topological trivialization relates to the question of

*“When are spin glasses actually glassy?”*

To motivate this point, consider the natural Langevin dynamics, which initialize from  $x_0 \in S_N$  (uniform or independent of  $H_{N,p}$ ) and proceed by

$$dx_t = P_x^\perp dB_t + \left( \beta \nabla_{\text{sph}} H_{N,p}(x_t) + \begin{bmatrix} \text{another drift term by Ito} \\ \text{correction for curvature} \end{bmatrix} \right) dt.$$

where  $B_t$  is  $N$ -dimensional Brownian motion. The stationary law of this continuous-time Markov chain is the Gibbs measure  $\mu_\beta$ . But at very low temperature (large  $\beta$ ), these dynamics can be very complex with exponential mixing time. An “aging” phenomenon is anticipated at low temperature, which can be informally stated in the context of  $1 \ll t \ll N$ ,

$$\lim_{t \rightarrow \infty} \lim_{N \rightarrow \infty} R(x_t, x_{\xi_t}) = f(\xi), \quad f(\alpha) \approx \alpha^{-c}. \quad (1)$$

Intuitively, if there exist “many critical points”, the dynamics easily get stuck and tend to be complex.

**Topological trivialization.** This concept refers to the case of  $O(1)$  critical points (2 critical points in particular). It is likely to happen when the Hamiltonian is added with an external field,

$$H_N = [\text{disordered term e.g. } H_{N,p}] + [\text{simple “signal” term}].$$

One specific example is tensor PCA, where for large SNR  $\lambda > 0$ , the Hamiltonian takes the form of

$$H_N = H_{N,p} + N\lambda R(x, \sigma)^p.$$

It then suggests that given a warm start  $x_0$  which is correlated with  $\sigma$ , it would be easy to find the MLE (maximizer of  $H_N$ ).

## 2 Spherical $p$ -spin model with random external field

Now we move to our model for today,

$$H_N(x) = H_{N,p}(x) + h\langle \vec{g}, x \rangle, \quad \vec{g} \sim \text{Normal}(0, I_N). \quad (2)$$

As shown later, this model has only 2 critical points with high probability if  $h > \sqrt{p(p-2)}$ . It would need a 2-dimensional variational problem to solve this optimization problem of this Hamiltonian. Similar conclusions also hold for another alternative setup where  $\vec{g} \in S_N$ .

The Hamiltonian  $H_N(x)$  can be seen as a centered Gaussian process with covariance

$$\mathbb{E}H_N(x)H_N(y) = N\xi(R(x, y)), \quad \xi(R) = R^p + h^2R. \quad (3)$$

Under this setup, we have

$$\xi(1) = h^2 + 1, \quad \xi'(1) = h^2 + p, \quad \xi''(1) = p(p-1).$$

Similar results also hold for a mixed  $p$ -spin model in which the Hamiltonian is taken as

$$H_N(x) = \sum_{p=1}^P \gamma_p^2 H_{N,p}(x), \quad \mathbb{E}H_N(x)H_N(y) = N\xi(R(x, y)), \quad \xi(R) = \sum_{p=1}^P \gamma_p^2 R^p.$$

Thus we will work in this generality during the rest of class. In general, it holds that

- If  $\xi'(1) > \xi''(1)$ , there are only 2 critical points.
- If  $\xi'(1) < \xi''(1)$ , it holds that  $\mathbb{E}|\text{Crt}_{S_N}(H_N)| \geq e^{cN}$ .

Our subsequent analysis starts with a joint Gaussian distribution for  $H_N(x)$  and its spherical gradient and tangent Hessian.

**Proposition 2.1** (Lemma 3.2 in [1]). *For fixed  $x \in S_N$ ,*

(a)  $\nabla_{\text{sph}} H_N(x)$  is independent of  $(H_N(x), \nabla_{\text{rad}} H_N(x), \nabla_{\text{tan}}^2 H_N(x))$ , with  $\nabla_{\text{sph}} H_N(x) \sim \text{Normal}(0, \xi'(1) I_{N-1})$ ;

(b)  $\left(\frac{H_N(x)}{N}, \nabla_{\text{rad}} H_N(x)\right)$  is a centered Gaussian with covariance

$$\begin{pmatrix} \xi(1) & \xi'(1) \\ \xi'(1) & \xi'(1) + \xi''(1) \end{pmatrix} = \sum_{p=1}^P \gamma_p^2 \begin{pmatrix} 1 & p \\ p & p^2 \end{pmatrix};$$

(c)  $\nabla_{\text{tan}}^2 H_N(x)$  is independent of  $(H_N(x), \nabla_{\text{rad}} H_N(x))$ , with  $\nabla_{\text{tan}}^2 H_N(x) \sim \sqrt{\xi''(1)} \text{GOE}(N-1)$ .

According to Kac-Rice formula, the main term of expected number of critical points is

$$\det\left(\nabla_{\text{sph}}^2 H_N(x)\right) = \det\left(\nabla_{\text{tan}}^2 H_N(x) - \nabla_{\text{rad}} H_N(x) I_{N-1}\right).$$

Specifically,  $Z = \nabla_{\text{rad}} H_N(x)$  is the most important. Given  $Z$ , the conditional distribution of  $E = \frac{H_N(x)}{N}$  is still Gaussian with

$$\mathbb{E}[E|Z] = \frac{\xi'(1)Z}{\xi'(1) + \xi''(1)}, \quad \text{Var}(E|Z) = \frac{1}{N} \left( \xi(1) - \frac{\xi'(1)^2}{\xi'(1) + \xi''(1)} \right). \quad (4)$$

It suffices to count the number of critical points with respect to each value of  $Z$ , since it follows that

$$\frac{1}{N} \log \mathbb{E}|\text{Crt}(H_N, E \approx a, Z \approx b)| \approx \frac{1}{N} \log \mathbb{E}|\text{Crt}(H_N, Z \approx b)| - \frac{(a - \mathbb{E}[E|Z = b])^2}{2N \text{Var}(E|Z)}.$$

When we fix  $Z$ , Kac-Rice formula still has the following ingredients:

- $\text{Vol}_{N-1}(S_N) = (2\pi e)^{N/2}$ ;
- gradient density at  $\mathbf{0}$ ,  $\varphi_{\nabla_{\text{sph}} H_N(x)}(\mathbf{0}) \approx (2\pi \xi'(1))^{-N/2}$ ;
- density for  $Z$ ,  $\exp\left(-\frac{NZ^2}{2(\xi'(1) + \xi''(1))}\right)$ ;
- determinant

$$\mathbb{E}|\det(\sqrt{\xi''(1)} \text{GOE}(N-1) - Z I_{N-1})| \approx \xi''(1)^{N/2} \exp\left(N\psi\left(\frac{Z}{\sqrt{\xi''(1)}}\right)\right).$$

Therefore, in conclusion, we should have

$$\frac{1}{N} \log \mathbb{E}|\text{Crt}(H_N, Z)| \approx \Phi_{\text{rad}}(Z) := \frac{1}{2} \left( 1 + \log \frac{\xi''(1)}{\xi'(1)} - \frac{Z^2}{\xi''(1) + \xi'(1)} \right) + \psi\left(\frac{Z}{\sqrt{\xi''(1)}}\right). \quad (5)$$

Since  $\psi(0) = -1/2$ , we would have  $\Phi_{\text{rad}}(0) = \frac{1}{2} \log \frac{\xi''(1)}{\xi'(1)}$ . Hence if  $\xi''(1) > \xi'(1)$ , we already have an exponentially large number of critical points in expectation from  $Z \approx 0$ , justifying the definition of the non-trivial phase. We focus on the latter case  $\xi''(1) < \xi'(1)$  subsequently and show the number of total expected critical points is  $e^{o(N)}$ .

Besides the value at 0, since  $\Phi_{\text{rad}}$  is even, we only need to investigate its property on the positive halfline. Due to

$$\psi'(t) = \text{Re} \left( \frac{t - \sqrt{t^2 - 4}}{2} \right) = \begin{cases} t/2, & 0 \leq t \leq 2, \\ \frac{t - \sqrt{t^2 - 4}}{2}, & t \geq 2, \end{cases}$$

we can also obtain its derivative by

$$\Phi'_{\text{rad}}(Z) = \begin{cases} Z \left( \frac{1}{2\xi''(1)} - \frac{1}{\xi'(1) + \xi''(1)} \right), & 0 \leq Z \leq 2\sqrt{\xi''(1)}, \\ Z \left( \frac{1}{2\xi''(1)} - \frac{1}{\xi'(1) + \xi''(1)} \right) - \frac{\sqrt{Z^2 - 4\xi''(1)}}{2\xi''(1)}, & Z \geq 2\sqrt{\xi''(1)}. \end{cases}$$

Similarly, the second derivative is (now discontinuous at  $2\sqrt{\xi''(1)}$  and):

$$\Phi''_{\text{rad}}(Z) = \begin{cases} \left( \frac{1}{2\xi''(1)} - \frac{1}{\xi'(1) + \xi''(1)} \right), & 0 \leq Z < 2\sqrt{\xi''(1)}, \\ \left( \frac{1}{2\xi''(1)} - \frac{1}{\xi'(1) + \xi''(1)} \right) - \frac{Z}{2\xi''(1)\sqrt{Z^2 - 4\xi''(1)}}, & Z > 2\sqrt{\xi''(1)}. \end{cases}$$

Recall that  $\Phi_{\text{rad}}(Z) = \Phi_{\text{rad}}(-Z)$ . Since  $\xi''(1) < \xi'(1)$ , it is **convex** on  $[-2\sqrt{\xi''(1)}, 2\sqrt{\xi''(1)}]$ . Since  $\sqrt{Z^2 - 4\xi''(1)} < |Z|$ , it is **concave** outside  $[-2\sqrt{\xi''(1)}, 2\sqrt{\xi''(1)}]$ . Thus (see picture), if we find  $Z_*$  with

$$\Phi_{\text{rad}}(Z_*) = \Phi'_{\text{rad}}(Z_*) = 0$$

we will have shown that

$$\max_{Z \in \mathbb{R}} \Phi_{\text{rad}}(Z) = 0.$$

Moreover the maximum will be achieved exactly at  $\pm Z_*$ , and we must have  $Z_* > 2\sqrt{\xi''(1)}$ . From this we will conclude (a weak form of) topological trivialization.

The solution turns out to be:

$$Z_* = \frac{\xi'(1) + \xi''(1)}{\sqrt{\xi'(1)}} = \sqrt{\xi''(1)} \cdot \left( a + \frac{1}{a} \right) > 2\sqrt{\xi''(1)}$$

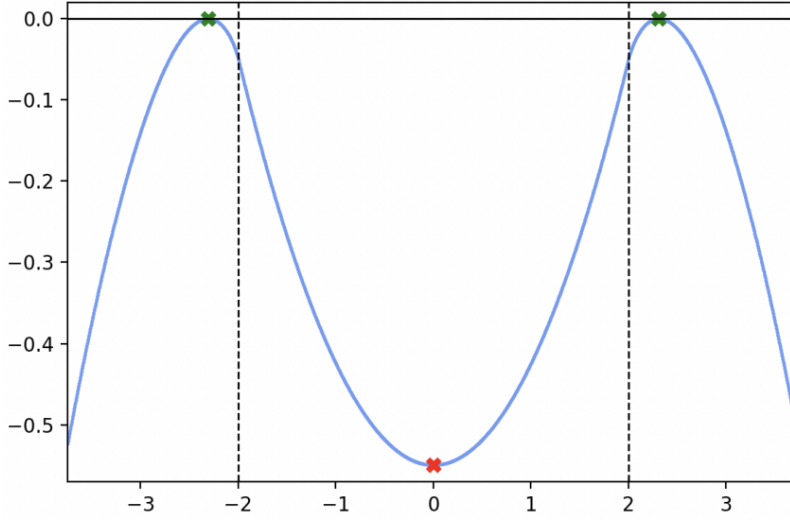


Figure 1: A typical diagram of  $\Phi_{\text{rad}}$ , rescaled so  $\xi''(1) = 1$ .

for  $a = \sqrt{\frac{\xi'(1)}{\xi''(1)}} > 1$ . Note that

$$\sqrt{\left(a + \frac{1}{a}\right)^2 - 4} = a - \frac{1}{a} \implies \sqrt{Z^2 - 4\xi''(1)} = \sqrt{\xi''(1)} \left(a - \frac{1}{a}\right) = \frac{\xi'(1) - \xi''(1)}{\sqrt{\xi'(1)}}. \quad (6)$$

Using this, it is easy to check that  $\Phi'_{\text{rad}}(Z_*) = 0$ .

To check  $\Phi_{\text{rad}}(Z_*) = 0$  involves more miraculous cancellations (skipped in class). Let  $Y_* = Z_* / \sqrt{\xi''(1)}$ . Recalling (1), note that (6) gives

$$\log\left(\frac{\sqrt{Y_*^2 - 4 + |Y_*|}}{2}\right) = \log\left(\sqrt{\frac{\xi'(1)}{\xi''(1)}}\right).$$

This cancels the log term in (5). The main remaining cancellation comes down to

$$\frac{(\xi'(1) - \xi''(1))^2}{4\xi'(1)\xi''(1)} - \frac{1}{2} - \frac{(\xi'(1) + \xi''(1)) \cdot (\xi'(1) - \xi''(1))}{4\xi'(1)\xi''(1)} = \frac{Z_*^2}{2(\xi'(1) + \xi''(1))} = \frac{\xi'(1) + \xi''(1)}{2\xi'(1)}.$$

In any case, we end up with the following theorem.

**Theorem 2.2.** *If  $\xi''(1) < \xi'(1)$ , then the following hold*

- (a)  $\mathbb{E}|\text{Crt}_{S_N}(H_N)| = e^{o(N)}$ ;
- (b) for any  $\epsilon > 0$ , expected number critical points with  $Z \notin [-Z_* - \epsilon, -Z_* + \epsilon] \cup [Z_* - \epsilon, Z_* + \epsilon]$  is  $e^{-cN}$ ;
- (c) for any  $\epsilon > 0$ , expected number critical points with  $E \notin [-E_* - \epsilon, -E_* + \epsilon] \cup [E_* - \epsilon, E_* + \epsilon]$  is  $e^{-cN}$  where

$$E_* = \mathbb{E}[E|Z_*] = \sqrt{\xi'(1)}.$$

We end this lecture with following two corollaries.

**Corollary 2.3.** *If  $\xi''(1) < \xi'(1)$ , it holds that*

$$\mathbf{p}\text{-}\lim_{N \rightarrow \infty} \max_{x \in S_N} \frac{H_N(x)}{N} = \sqrt{\xi'(1)}.$$

*Proof.* The maximal value of  $H_N$  must occur at a critical point, so we must have with high probability:

$$\max_{x \in S_N} \frac{H_N(x)}{N} \approx \pm E_*.$$

We have the symmetry in distribution:  $H_N \stackrel{d}{=} -H_N$ . This implies  $\mathbb{E} \max_x \frac{H_N(x)}{N} \geq 0$ . Consequently by concentration, the limiting maximum value must be  $E_*$ .  $\square$

**Corollary 2.4.** *If  $\xi''(1) < \xi'(1)$ , all critical points are local maxima or minima.*

*Proof.* By Theorem 2.2 and sign symmetry, it suffices to show the expected number of saddle points (i.e. critical points which are not local optima) with  $Z \in [Z_* - \epsilon, Z_* + \epsilon]$  is at most  $e^{-c(\xi, \epsilon)N}$ , for small  $\epsilon > 0$ .

We use Proposition 3.3 from last lecture, with  $E$  replaced by  $Z$ . Analogously to (5), this gives the upper bound

$$\frac{1}{N} \log \mathbb{E} |\text{Crt}^{\text{saddle}}(H_N, Z \in [Z_* - \epsilon, Z_* + \epsilon])| \leq \sup_{Z \in [Z_* - \epsilon, Z_* + \epsilon]} \frac{1}{2} \left( 1 + \log \frac{\xi''(1)}{\xi'(1)} - \frac{Z^2}{\xi''(1) + \xi'(1)} \right) + \psi \left( \frac{Z}{\sqrt{\xi''(1)}} \right) - \frac{1}{2N} \log \mathbb{P}[I|Z].$$

Here  $I$  is the event of being a saddle point. Since we have shown trivialization, it suffices to show that

$$\inf_{Z \in [Z_* - \epsilon, Z_* + \epsilon]} \frac{1}{N} \log \mathbb{P}[I|Z] > c(\xi, \epsilon) > 0$$

for small enough  $\epsilon$  depending on  $\xi$ . Indeed, since  $Z_* > 2\sqrt{\xi''(1)}$ , the event  $I$  holding requires the tangential Hessian

$$\nabla_{\text{tan}}^2 H_N() \sim \sqrt{\xi''(1)} \text{GOE}(N-1)$$

to have an outlier eigenvalue, which has probability at most  $e^{-cN}$ . This concludes the proof.  $\square$

## References

- [1] David Belius, Jiri Cerny, Shuta Nakajima, and Marius A Schmidt. Triviality of the geometry of mixed p-spin spherical hamiltonians with external field. *Journal of Statistical Physics*, 186(1):12, 2022. 2