

## Lectures 20 – 22 : Scaling limit of $\beta$ -ensembles

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### 1 Edelman-Sutton Conjecture (cont.)

#### 1.1 Existence of eigenvalue-eigenfunction pair for the smallest eigenvalue of $cH_\beta$

In the last lecture, we proved  $\exists\{f_n\}$  and a nonnegative random variable  $K$  with  $\|f_n\|_2^2 = 1$ ,  $\|f_n\|_*^2 \langle K \rangle a.s. \forall n$ , so that

$$\langle f_n, \mathcal{H}_\beta f_n \rangle \rightarrow \widetilde{\Lambda}_0 = \inf_{\substack{\|f\|_2^2=1 \\ f \in L^*}} \langle f, \mathcal{H}_\beta f \rangle a.s.$$

Also, we proved the **lemma**: If  $\{f_n\}$  is bounded in  $L^*$ , then we can find a subsequence  $\{f_{n_k}\}$  so that

- 1)  $f_{n_k} \rightarrow \widetilde{f}$  in  $L_2$
- 2)  $f'_{n_k} \rightarrow \widetilde{f}'$  weak convergence in  $L^2$
- 3)  $f_{n_k} \rightarrow \widetilde{f}$  uniformly on compacts
- 4)  $f_{n_k} \rightarrow \widetilde{f}$  weakly in  $L^*$

Without loss of generality, we may assume  $\{f_n\}$  itself has the above properties.

We will show that  $\widetilde{\Lambda}_0 = \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle a.s.$  and that  $(\widetilde{\Lambda}_0, \widetilde{f})$  is an eigenvalue-eigenfunction pair with the smallest eigenvalue.

*Proof.* In the previous lecture, we proved  $\exists$  a nonnegative random variable  $C(B)$  :

$$\sup_x \frac{\max(|\bar{B}'(x)|, |B(x) - \bar{B}(x)|)}{\sqrt{\log(2+x)}} < C(B) < \infty.$$

Hence,  $\forall \epsilon > 0$ ,  $\exists$  a random variable  $X$ , so that  $|\bar{B}'(x)| < \epsilon(1+x)$  and  $|B(x) - \bar{B}(x)| < \epsilon\sqrt{1+x}$ ,  $\forall x > X$ . Then,  $\forall f_n$ ,

$$\langle f_n, \mathcal{H}_\beta f_n \rangle = \int_0^\infty [(f'_n)^2 + x f_n^2] dx + \frac{2}{\sqrt{\beta}} \int_0^x f_n^2 \bar{B}' dx + \frac{4}{\sqrt{\beta}} \int_0^x f'_n f_n (\bar{B} - B) dx + E \quad (1)$$

where  $|E| \leq \epsilon C(\beta) \|f_n\|_*^2 \leq \epsilon C(\beta) K$  *a.s.*, by Cauchy-Schwarz inequality, and  $C(\beta)$  is a nonnegative finite constant.

Then, we want to show that

$$\widetilde{\Lambda}_0 = \lim_{n=1} \langle f_n, \mathcal{H}_\beta f_n \rangle = \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle \text{ a.s.}$$

The second term in (1),

$$\frac{2}{\sqrt{\beta}} \int_0^x f_n^2 \bar{B}' dx \rightarrow \frac{2}{\sqrt{\beta}} \int_0^x \widetilde{f}^2 \bar{B}' dx \text{ a.s.}$$

because  $f_n \rightarrow \widetilde{f}$  uniformly on compacts.

We **claim** that the third term in (1),

$$\frac{4}{\sqrt{\beta}} \int_0^x f_n' f_n (\bar{B} - B) dx \rightarrow \frac{4}{\sqrt{\beta}} \int_0^x \widetilde{f}' \widetilde{f} (\bar{B} - B) dx \text{ a.s.}$$

It's enough to show  $\int_0^x |f_n f_n' - \widetilde{f} \widetilde{f}'| dx \rightarrow 0$  *a.s.*

Here we use triangle inequality. For large enough  $n$ , we have

$$\begin{aligned} & \int_0^x |f_n f_n' - \widetilde{f} \widetilde{f}'| dx \\ & \leq \int_0^x |f_n f_n' - \widetilde{f}' f_n| dx + \int_0^x |\widetilde{f}' f_n - \widetilde{f} \widetilde{f}'| dx \\ & \leq \int_0^x |f_n| |f_n' - \widetilde{f}'| dx + \|\widetilde{f}'\|_2 \|\widetilde{f} - f_n\|_2 \\ & \leq \int_0^x (|\widetilde{f}| + \epsilon) |f_n' - \widetilde{f}'| dx + \|\widetilde{f}'\|_2 \|\widetilde{f} - f_n\|_2 \\ & \rightarrow 0 \text{ a.s.} \end{aligned} \tag{2}$$

The last inequality is because of the uniform convergence of  $\{f_n\}$  in  $L^2$ .

Since  $(|\widetilde{f}| + \epsilon)$  is in  $L^2$  and  $f_n' \rightarrow \widetilde{f}'$  weak convergence in  $L^2$ , the first term in (2)  $\rightarrow 0$  *a.s.*.

Since  $\|\widetilde{f}'\|_2$  is finite and  $f_n \rightarrow \widetilde{f}$  in  $L^2$ , the second term in (2)  $\rightarrow 0$  *a.s.*.

Then the **claim** follows.

Finally, we deal with the first term in (1).

By Fatou's lemma,

$$\liminf \int_0^\infty x f_n^2 dx \geq \int_0^\infty x \widetilde{f}^2 dx \text{ a.s.}$$

And

$$\|\widetilde{f}'\|_2^2 = \int_0^\infty \widetilde{f}' \widetilde{f}' dx = \int_0^\infty f_n' \widetilde{f}' dx \leq \liminf \|f_n'\|_2 \|\widetilde{f}'\|_2 \text{ a.s.},$$

which implies

$$\liminf \int_0^\infty (f_n')^2 dx = \liminf \|f_n'\|_2^2 \geq \|\widetilde{f}'\|_2^2 = \int_0^\infty (\widetilde{f}')^2 dx \text{ a.s.}$$

Hence,

$$\begin{aligned}
\widetilde{\Lambda}_0 &= \lim_{n=1} \langle f_n, \mathcal{H}_\beta f_n \rangle \\
&= \lim_{n=1} \left\{ \int_0^\infty [(f'_n)^2 + x f_n^2] dx + \frac{2}{\sqrt{\beta}} \int_0^x f_n^2 \bar{B}' dx + \frac{4}{\sqrt{\beta}} \int_0^x f'_n f_n (\bar{B} - B) dx + E \right\} \\
&\geq \int_0^\infty [(\widetilde{f}')^2 + x \widetilde{f}^2] dx + \frac{2}{\sqrt{\beta}} \int_0^x \widetilde{f}^2 \bar{B}' dx + \frac{4}{\sqrt{\beta}} \int_0^x \widetilde{f}' \widetilde{f} (\bar{B} - B) dx - \epsilon C(\beta) K \\
&\geq \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle - 2\epsilon C(\beta) K \text{ a.s.}
\end{aligned}$$

Since  $\epsilon$  is arbitrary,

$$\widetilde{\Lambda}_0 \geq \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle \text{ a.s.}$$

According to the definition of  $\widetilde{\Lambda}_0$ ,

$$\widetilde{\Lambda}_0 = \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle \text{ a.s.}$$

Next, we prove the second part of the statement.

First, let's go back to the matrices example. Assume that

$$\lambda_0 = \inf_{\|x\|=1} \langle x, Ax \rangle = \min_{\|x\|=1} \langle x_0, Ax_0 \rangle$$

Then, for any  $\epsilon > 0$

$$\begin{aligned}
&\frac{1}{\|x_0 + \epsilon y\|^2} \langle (x_0 + \epsilon y), A(x_0 + \epsilon y) \rangle - \langle x_0, Ax_0 \rangle \\
&= (1 - 2\epsilon \langle x_0, y \rangle + O(\epsilon^2)) (\langle x_0, Ax_0 \rangle + 2\epsilon \langle y, Ax_0 \rangle + O(\epsilon^2)) - \langle x_0, Ax_0 \rangle \\
&= 2\epsilon (\langle y, Ax_0 \rangle - \langle x_0, y \rangle \langle x_0, Ax_0 \rangle) + O(\epsilon^2) \\
&= 2\epsilon (\langle y, Ax_0 - \lambda_0 x_0 \rangle) + O(\epsilon^2)
\end{aligned}$$

Since  $x_0$  is the minimizer, the first term in the above equation must be 0 for all  $y$ . That is,  $\langle y, Ax_0 - \lambda_0 x_0 \rangle = 0$  for all  $y$ . Hence,  $(\lambda_0, x_0)$  is an eigenvalue-eigenvector pair.

Then, we use the same idea for  $\widetilde{f}$  and  $\mathcal{H}_\beta$ . Define that

$$f^{\epsilon, \varphi} = \frac{\widetilde{f} + \epsilon \varphi}{\|\widetilde{f} + \epsilon \varphi\|_2}$$

Then,

$$\begin{aligned}
&\langle f^{\epsilon, \varphi} - \mathcal{H}_\beta \widetilde{f}^{\epsilon, \varphi} \rangle - \langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle \\
&= 2\epsilon \left[ -\langle \widetilde{f}, \mathcal{H}_\beta \widetilde{f} \rangle \int_0^\infty \widetilde{f} \varphi dx + \int_0^\infty (\widetilde{f}' \varphi' + x \widetilde{f} \varphi) dx - \frac{2}{\sqrt{\beta}} \int_0^\infty \bar{B}' \widetilde{f} \varphi dx \right. \\
&\quad \left. - \frac{2}{\sqrt{\beta}} \int_0^\infty (\bar{B} - B) (\widetilde{f} \varphi)' dx \right] + O(\epsilon^2) \\
&= 2\epsilon [\langle \varphi, \mathcal{H}_\beta \widetilde{f} - \widetilde{\Lambda}_0 \widetilde{f} \rangle] + O(\epsilon^2)
\end{aligned}$$

Use the same argument that we used in matrices example,  $(\widetilde{\Lambda}_0, \widetilde{f})$  is an eigenvalue-eigenfunction pair for  $\mathcal{H}_\beta$ . And we also proved the  $\widetilde{\Lambda}_0$  is the smallest eigenvalue.  $\square$

Define  $\mathcal{H}_0$  as the function space of all the eigenfunction corresponding to  $\widetilde{\Lambda}_0$ .

**Claim**  $\mathcal{H}_0$  is finite dimensional.

*Proof.* Suppose  $\exists$  countable infinite  $f_1, f_2, \dots$ , which form an orthonormal( $L_2$ ) basis in  $\mathcal{H}_0$ . Then we can find a random variable  $K$  so that  $\|f_i\|_*^2 \langle K \rangle a.s., \forall i \in \mathbf{N}$ . By the lemma in the previous lecture,  $\exists$  a subsequence  $\{f_{n_k}\}$  of  $\{f_n\}$  that  $f_{n_k} \xrightarrow{L_2} f$  and  $f \in \mathcal{H}_0$ . Then  $\forall n$ , we can find  $k$  such that  $n_k \geq n$ . Since  $f_n \perp \{f_{n_k}, f_{n_{k+1}}, \dots\}$ , we get  $f_n \perp f$ , which is a contradiction to the fact that  $\{f_1, f_2, \dots\}$  is an orthonormal basis. Hence,  $\mathcal{H}_0$  is finite dimensional.  $\square$

We can define (find) the other eigenvalues inductively:  $\widetilde{\Lambda}_0 \leq \widetilde{\Lambda}_1 \leq \widetilde{\Lambda}_2 \leq \dots$ .

## 1.2 Counting function of eigenvalues

First let's look at a much simpler problem:

$$\begin{cases} -\partial_{XX}f = \lambda f \\ f(0) = 0, f(L) = 0 \end{cases}$$

We have

$$f_{k-1}(x) = \sin(k \frac{\pi}{L} x), \quad k = 1, 2, \dots$$

$$\lambda_{(k-1)} = (k \frac{\pi}{L})^2$$

Then,  $k^{th}$  eigenfunction has  $k$  roots in  $(0, L)$ .

For any  $\lambda$ :

$$\begin{cases} f_\lambda : -\partial_{XX}f = \lambda f \\ f(0) = 0 \end{cases}$$

$\lambda$  is an eigenvalue  $\Leftrightarrow f_\lambda(L) = 0$ .

Then,  $f_\lambda$  has  $k$  roots in  $(0, L) \Leftrightarrow \lambda_k \geq \lambda \geq \lambda_{k-1}$ .

This is true in general on  $[0, L]$ :

$$\begin{cases} -\partial_{XX}f + Vf = \lambda f \\ f(0) = 0 \end{cases}$$

And for such kind problem, we can use the Riccati transformation:

$$p = \frac{f'}{f} \Rightarrow p' = -(\frac{f'}{f})^2 + \frac{f''}{f} = p^2 + \frac{f''}{f}$$

Then,

$$\begin{cases} -\partial_{XX}f + Vf = \lambda f \\ f(0) = 0 \end{cases} \Leftrightarrow \begin{cases} p' = -\lambda + V - p^2 \\ p(0) = \infty \end{cases}$$

# of explosions = # of eigenvalues below  $\lambda$ .

This is same on  $[0, \infty)$ :

$$\begin{cases} \mathcal{H}_\beta f = \lambda f \\ f(0) = 0 \end{cases} \Leftrightarrow \begin{cases} -\partial_{XX} f + V f = \lambda f \\ f(0) = 0 \end{cases}$$

where  $V = x + \frac{2}{\sqrt{\beta}} B'(x)$ .

Let  $p = f'/f$ , then

$$\begin{aligned} p'(x) &= \frac{2}{\sqrt{\beta}} B'(x) + x - \lambda - p^2(x) \\ \Leftrightarrow dp &= \frac{2}{\sqrt{\beta}} dB + (x - \lambda - p^2) dx \\ \Leftrightarrow p(x_1) - p(x_0) &= \frac{2}{\sqrt{\beta}} (B(x_1) - B(x_0)) + \int_{x_0}^{x_1} [x - \lambda - p^2(x)] dx \end{aligned}$$

The last equation is a diffusion model which has a general form

$$dX_t = \sigma(X_t, t) dB_t + f(X_t, t) dt.$$

The discrete analogue of the diffusion model is

$$X_{t+\epsilon} = X_t + \eta_t$$

where  $\eta_t$  is asymptotically normal with expectation  $f(X_t, t) \cdot \epsilon$  and variance  $\sigma(X_t, t)^2 \cdot \epsilon$ .

If  $\sigma$  and  $f$  are appropriate, it's solvable. In our case, for the equation

$$\mathcal{H}_\beta f = \lambda f, \quad f(0) = 0,$$

if the solution  $f$  decays fast enough, it can be proved that  $(\lambda, f)$  is an ev-ef pair.

### 1.3 Limit of the matrix form

Recall that

$$M_n^\beta = \begin{bmatrix} a_n & b_{n-1} & & & \\ b_{n-1} & a_{n-1} & b_{n-2} & & \\ & b_{n-2} & \ddots & \ddots & \\ & & \ddots & \ddots & b_1 \\ & & & b_1 & a_1 \end{bmatrix}, \quad H_n^\beta = n^{\frac{1}{6}} (2\sqrt{n} I_n - M_n)$$

where  $a_k \sim \frac{1}{\sqrt{\beta}}N(0, 2)$  and  $b_k \sim \frac{1}{\sqrt{\beta}}\chi_{\beta k}$ , and

$$H_n^\beta = n^{\frac{2}{3}} \begin{bmatrix} 2 & -1 & & & \\ -1 & 2 & -1 & & \\ & -1 & \ddots & \ddots & \\ & & \ddots & \ddots & -1 \\ & & & -1 & 2 \end{bmatrix} + n^{\frac{1}{6}} \begin{bmatrix} 0 & \gamma_1 & & & \\ \gamma_1 & 0 & \gamma_2 & & \\ & \gamma_2 & \ddots & \ddots & \\ & & \ddots & \ddots & \gamma_{n-1} \\ & & & \gamma_{n-1} & 0 \end{bmatrix} + \frac{n^{\frac{1}{6}}}{\sqrt{2\beta}} \begin{bmatrix} 2\rho_0 & \zeta_1 & & & \\ \zeta_1 & 2\rho_1 & \zeta_2 & & \\ & \zeta_2 & \ddots & \ddots & \\ & & \ddots & \ddots & \zeta_{n-1} \\ & & & \zeta_{n-1} & 2\rho_{n-1} \end{bmatrix}$$

where  $\gamma_k = \sqrt{n} - E\chi_{\beta(n-k)} \approx \frac{k}{2\sqrt{n}}$ .

We know that

$$w_{n,1}(x) = n^{-\frac{1}{6}} \sum_{i=1}^{\lfloor x \cdot n^{1/3} \rfloor} \rho_i \Rightarrow B_1(x)$$

$$w_{n,2}(x) = n^{-\frac{1}{6}} \sum_{i=1}^{\lfloor x \cdot n^{1/3} \rfloor} \zeta_i \Rightarrow B_2(x)$$

$$2n^{-\frac{1}{6}} \sum_{i=1}^{\lfloor x \cdot n^{1/3} \rfloor} \gamma_i \rightarrow \frac{x^2}{2}$$

Define  $\|v\|_{2,n}^2 = n^{-\frac{1}{3}} \sum v_k^2$  and

$$\langle v, H_n^\beta v \rangle_{2,n} = n^{\frac{1}{3}} \sum (v_{k+1} - v_k)^2 + 2n^{-\frac{1}{6}} \sum \gamma_k v_k v_{k+1} + n^{-\frac{1}{6}} \sqrt{\frac{2}{\beta}} \sum \rho_k v_k^2 + n^{-\frac{1}{6}} \sqrt{\frac{2}{\beta}} \sum \zeta_k v_k v_{k+1}$$

which are scaled version of  $L_2$  norm and inner product of vectors, and

$$\|v\|_*^2 = n^{\frac{1}{3}} \sum (v_{k+1} - v_k)^2 + n^{-\frac{2}{3}} \sum k v_k^2 + n^{-\frac{1}{3}} \sum v_k^2$$

$$= n^{-\frac{1}{3}} \sum \left( \frac{v_{k+1} - v_k}{n^{-1/3}} \right)^2 + n^{-\frac{1}{3}} \sum (k \cdot n^{-\frac{1}{3}}) v_k^2 + n^{-\frac{1}{3}} \sum v_k^2$$

which is a discrete version of  $L^*$  norm if we view  $v_k$  as the mesh sampled by size of  $n^{-\frac{1}{3}}$ .

**Lemma 1.** *There exists a sequence of random constants  $C_1(n), C_2(n), C_3(n)$  such that*

$$C_1 \|v\|_*^2 - C_2 \|v\|_2^2 \leq \langle v, H_n^\beta v \rangle \leq C_3 \|v\|_*^2.$$

Also  $C_1(n), C_2(n), C_3(n)$  are tight.

Define  $L_{2,n} = \{f \in L_2 : \text{step functions with mesh } n^{-\frac{1}{3}}, f(0) = 0, f(x) = 0, x \geq n^{\frac{2}{3}}\}$ . For each  $f \in L_{2,n}$ , it can be treated as a vector of length  $n$ . Let  $P_n : L_2 \rightarrow L_{2,n}$  be the projection from  $L_2$  to  $L_{2,n}$ , then we can define

$$\langle f, \hat{H}_n^\beta f \rangle_{L_2} = \langle P_n f, H_n^\beta P_n f \rangle_{2,n}.$$

For each  $n$ , let  $\lambda_n$  denote the minimum eigenvalue of  $\hat{H}_n^\beta$  and  $f_n$  denote the corresponding eigenfunction satisfying  $\|f_n\| = 1$ . We have that  $\lambda_n \rightarrow \lambda$  (maybe subsequence) and  $f_n \xrightarrow{L_2} f$ .

**Lemma 2.**

1. If  $f_n \in L_{2,n}$  with  $f_n \rightarrow f$  in  $L_2$  weakly,  $n^{\frac{1}{3}}(f_n(x + n^{-\frac{1}{3}}) - f_n(x)) \rightarrow f'$  in  $L_2$  weakly, then for each  $\varphi \in C_0^\infty$ ,

$$\langle \varphi, \hat{H}_n^\beta f_n \rangle \rightarrow \langle \varphi, \mathcal{H}_\beta f \rangle$$

Notice that the left term is a partial summation.

2. If  $f_n \in L_{2,n}$  is bounded in  $L_{*,n}$ , then there exists a subsequence  $f_{n_k} \rightarrow f$  in  $L_2$  such that

$$\langle \varphi, \hat{H}_n^\beta f_n \rangle \rightarrow \langle \varphi, \mathcal{H}_\beta f \rangle$$

Therefore, for all  $\varphi \in C_0^\infty$ ,

$$\begin{aligned} \lambda \langle \varphi, f \rangle &= \lim \langle \varphi, \lambda_n f_n \rangle \\ &= \lim \langle \varphi, \hat{H}_n^\beta f_n \rangle \\ &= \langle \varphi, \mathcal{H}_\beta f \rangle \end{aligned}$$

which implies  $(\lambda, f)$  is an ev-ef pair of  $\mathcal{H}_\beta$ .

Next, we want to show that  $\lambda$  is the smallest eigenvalue of  $\mathcal{H}_\beta$ . Let  $\Lambda_0$  be the smallest eigenvalue, so there exists minimizing function  $f_0$  for  $\Lambda_0$ . By definition the discretized version

$$\lambda_n \leq \frac{\lim \langle f_{0,n}, \hat{H}_n^\beta f_{0,n} \rangle}{\lim \langle f_{0,n}, f_{0,n} \rangle} \rightarrow \Lambda_0$$

where the left hand side  $\lambda_n \rightarrow \lambda$ . So it implies that  $\lambda \leq \Lambda_0$ , and together with the fact that  $\lambda \geq \Lambda_0$ ,

$$\Rightarrow \lambda = \Lambda_0.$$

We can use the induction to deal with other eigenvalues, then

$$(\lambda_0^{(n)}, \lambda_1^{(n)}, \lambda_2^{(n)}, \dots) \Rightarrow (\Lambda_0, \Lambda_1, \Lambda_2, \dots)$$

**Remark:** we can also obtain the edge scaling of the Laguerre ensemble  $M = AA^T \stackrel{d}{=} \frac{1}{\beta} BB^T$ , where  $B$  is a bidiagonal matrix in the form

$$B = \begin{bmatrix} \chi_{\beta m} & & & \\ \chi_{\beta(n-1)} & \chi_{\beta(m-1)} & & \\ & \ddots & \ddots & \\ & & & \ddots \end{bmatrix}$$

## 2 Description of the limiting process

Recall that the density of eigenvalues are given by

$$P_{\beta,n}(\lambda_1, \dots, \lambda_n) = \frac{1}{Z_{n,\beta}} \prod_{i>j} |\lambda_i - \lambda_j|^\beta e^{-\frac{\beta}{4} \sum \lambda_i^2}, \quad \lambda_1 < \lambda_2 < \dots < \lambda_n.$$

### 2.1 Edge scaling limit

By Tracy-Widom, in the cases  $\beta = 1, 2, 4$ ,  $n^{\frac{1}{6}}(\lambda_n - 2\sqrt{n}) \rightarrow TW_\beta = -\Lambda_0$ .

For any  $a > 0$ ,

$$P(n^{\frac{1}{6}}(\lambda_n - 2\sqrt{n}) \geq a) = \exp\left(-\frac{2}{3}\beta a^{\frac{3}{2}}(1 + o(1))\right) \quad (3)$$

$$P(n^{\frac{1}{6}}(\lambda_n - 2\sqrt{n}) \leq -a) = \exp\left(-\frac{1}{24}\beta a^3(1 + o(1))\right). \quad (4)$$

For  $\beta = 1, 2, 4$ , one can prove this using the explicit formulas for the limits. By Ramirez-Rode-Virag, these asymptotic will hold for all  $\beta > 0$ .

$$P(TW_\beta > a) = P(\Lambda_0 < -a)$$

where

$$\begin{aligned} \Lambda_0 &= \inf_{\substack{f \in L^* \\ \|f\|_2^2=1}} \left\{ \int_0^\infty (f'^2 + x f^2) dx + \frac{2}{\sqrt{\beta}} \int_0^\infty f^2 dB \right\} \\ &\leq \frac{1}{\|g\|_2^2} \left\{ \int_0^\infty (g'^2 + x g^2) dx + \frac{2}{\sqrt{\beta}} \int_0^\infty g^2 dB \right\} \quad \forall g \in L^* \end{aligned}$$

in which  $\frac{2}{\sqrt{\beta}} \int_0^\infty g^2 dB = N(0, 1) \times \frac{2}{\sqrt{\beta}} \left( \int_0^\infty g^4 dx \right)^{\frac{1}{2}}$ . Therefore,

$$\begin{aligned} P(\Lambda_0 < -a) &\geq P\left(\frac{2}{\sqrt{\beta}} \left( \int_0^\infty g^4 dx \right)^{\frac{1}{2}} \times N(0, 1) < -a \int g^2 dx - \int (g'^2 + x g^2) dx\right) \\ &= \Phi\left(\frac{-a \int g^2 dx - \int (g'^2 + x g^2) dx}{\frac{2}{\sqrt{\beta}} \left( \int_0^\infty g^4 dx \right)^{\frac{1}{2}}}\right) \end{aligned}$$

Since it holds for any  $g \in L^*$ , we can choose a particular function  $g = \text{sech}(\sqrt{a}(x-1))$  to obtain the lower bound for up tail in (3). Similarly

$$P(TW_\beta < -a) \leq P(\langle g, \mathcal{H}_\beta g \rangle > a \|g\|_2^2)$$

With a suitable test function, it gives you the upper bound in (4). And the other two inequality can be obtained by diffusion description.



## 2.2 Bulk scaling limit of the $\beta$ -ensemble

We have showed that, for  $|c| < 2$ ,

$$2\pi\sqrt{n}\sigma(c)(\Lambda_n - 2\sqrt{n}) \Rightarrow \text{some point process limit}, \quad \sigma(c) = \frac{1}{2\pi}\sqrt{4-c^2}$$

By V-Viray, for each  $\beta > 0$ , there is a limiting point process(Sine $_\beta$ ) which does not depend on  $c$ .

How can we describe the process Sine $_\beta$ ?

We will describe the counting function  $N(\lambda)$  of Sine $_\beta$ .

$$N(\lambda) = \begin{cases} 0 & \lambda = 0 \\ \#\{\text{points in } [0, \lambda)\} & \lambda > 0 \\ \#\{\text{points in } [-\lambda, 0)\} & \lambda < 0 \end{cases}$$

For any  $a < b$ ,  $\#$  of points in  $[a, b) = N(b) - N(a) \stackrel{d}{=} N(b-a)$  which means it is translation invariant.

Next we will introduce two equivalent description of the process.

### 2.2.1 First description

Let  $Z(t)$  be complex BM.  $Z(t) = B_1(t) + iB_2(t)$  where  $B_1, B_2$  are independent standard BM. Consider the following family of stochastic differential equations.

$$\alpha_\lambda(t), \quad \lambda \in \mathbb{R}, \quad t \in [0, \infty)$$

$$d\alpha_\lambda(t) = \lambda \frac{\beta}{4} e^{-\frac{\beta}{4}t} dt + \text{Re} \left( (e^{-i\alpha_\lambda} - 1) dZ \right), \quad \alpha_\lambda(0) = 0$$

where  $\text{Re} \left( (e^{-i\alpha_\lambda} - 1) dZ \right) = (\cos \alpha_\lambda - 1)dB_1 + \sin \alpha_\lambda dB_2$ .

$$\alpha_\lambda(t) = \int_0^t \lambda \frac{\beta}{4} e^{-\frac{\beta}{4}s} ds + \int_0^t \text{Re} \left( (e^{-i\alpha_\lambda(s)} - 1) dZ \right)$$

**Claim:**

1. This will have a solution.
2.  $\frac{1}{2\pi} \lim_{t \rightarrow \infty} \alpha_\lambda(t)$  exists a.s. and it's an integer. Define

$$N(\lambda) = \frac{1}{2\pi} \lim_{t \rightarrow \infty} \alpha_\lambda(t).$$

3.  $N(\lambda)$  is increasing in  $\lambda$ .  $N(\lambda)$  is an integer-valued step function.
4.  $N(\lambda)$  is the counting function of Sine $_\beta$ .

### 2.2.2 Second description

In this section, we need some knowledge about 2D hyperbolic space. Specifically, we will use the Poincaré disk model on  $\mathbb{U} = \{|z| < 1\}$ .

lines:      1) lines passing 0  
               2) circles  $\perp \{|z| = 1\}$   
 angles:     same as in the euclidean plane  
 distance:    $4 \frac{1}{1 - |z|^2}$

It's conformal invariant. For  $|\alpha| < 1$ , the transformation  $T(z) = e^{iv} \frac{z - \alpha}{1 - \bar{\alpha}z}$  keep distance, angle, ...

Rotation around a point.

- rotation around 0:  $z \rightarrow e^{iv} \cdot z$
- rotation around  $\alpha (|\alpha| < 1)$ : first shift  $T(z) = \frac{z - \alpha}{1 - \bar{\alpha}z}$ , then rotation and transform back  

$$z \rightarrow T^{-1}(e^{iv} \cdot T(z))$$

BM in the hyperbolic plane - scaling limit of a simple random walk in the hyperbolic plane.

$$dB = \frac{1 - |B|^2}{2} dZ$$

where  $B(t)$  is hyperbolic BM and  $Z(t)$  is complex BM.

For each  $\lambda$ , we will follow a point  $Z_\lambda(t) \in \mathbb{U}$  on the boundary  $z_\lambda(0) = 1$ . We rotate (in hyp)  $Z_\lambda(t)$  about the point  $B(t)$  with speed  $\lambda \frac{\beta}{4} e^{-\frac{\beta}{4}t}$ , and

$$\frac{\partial Z_\lambda(t)}{\partial t} = i\lambda \frac{\beta}{4} e^{-\frac{\beta}{4}t} \frac{|z_\lambda - B|^2}{1 - |B|^2}$$

For each  $\lambda$ , we also follow the winding angle of  $z_\lambda$  about  $B(t)$ .

$$\alpha_\lambda(t) = \langle 1, B(t), Z(t) \rangle$$

Define  $N(\lambda)$  as the total winding number of  $Z_\lambda(t)$ .

**Claim:**

1.  $N(\lambda)$  is increasing.
2.  $N(\lambda)$  is the counting function of  $\text{Sine}_\beta$ .