

Introduction to random zeros of holomorphic sections: Part 3: large deviation and hole probability

Notes available on webpage
<http://www.mi.uni-koeln.de/~bxliu/>

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§ 0 Result: equidistribution in semi-classical limit

X (connected) complex manifold

ω J -invariant Hermitian metric

$$\implies g^{TX}(\cdot, \cdot) = \omega(\cdot, J\cdot)$$

Riemannian metric

(L, h_L) Hermitian holomorphic line bundle

$$(H) : \begin{cases} \omega = c_1(L, h_L) := \frac{\sqrt{-1}}{2\pi} R^h > 0 \\ g^{TX} \text{ (or } \omega) \text{ is complete.} \\ Ric_\omega \geq -c\omega \text{ for some } c > 0 \end{cases}$$

Semi-classical setting

$$H_{(2)}^0(X, L^p)$$

dimension $d_p \in \mathbb{N} \cup \{\infty\}$
 $d_p \gg 0$ as $p \uparrow$

In Theorem 3 (Lecture 2), we have

$$B_p(X) = p^n + b_1(X) p^{n-1} + \dots$$

Bergman kernel function

\implies Equidistribution results: $\forall \varphi \in \Omega_0^{n-1, n-1}(X)$
 in Theorem 4.

$$\left\langle \frac{1}{p} [Z(S(L^p))], \varphi \right\rangle \xrightarrow{\text{a.s.}} \langle c_1(L, h_L), \varphi \rangle$$

In fact, we have shown more results on Bergman kernels

$$|B_p(x, y)| \sim p^n \exp\left(-\frac{\pi}{2} d(x, y)^2\right)$$

when x, y are nearby

\implies refine the above results!

§1 Large deviation estimates and hole probability

Def: $U \subset X$, β a current on X , for $\alpha > 0$

$$\|\beta\|_{U, -\alpha} := \sup_{\substack{\varphi \text{ test form} \\ \text{supp } \varphi \subset U \\ |\varphi|^\alpha \leq 1}} |\langle \beta, \alpha \rangle|$$

$$\frac{\alpha=2}{\Delta}$$

For $S_p \subset H^0(X, L^p) \quad \forall U \subset X$

$$\| \underbrace{[Z(S_2)]}_{\substack{\partial \bar{\partial} \log |S_2| \\ \int_U}} \|_{U, -2} < +\infty$$

Thm 9: $n = \dim_{\mathbb{C}} X \quad \forall \delta > 0, U \subset X, X \supset U$
 $\exists C = C(\delta, U) > 0$ s.t.

$$\mathbb{P} \left(\left\| \frac{1}{p} [Z(S(p))] - c_2(L, h_L) \right\|_{U, -2} > \delta \right) \leq e^{-C p^{n+1}}$$

LDE concentration of measure

Large deviation estimates

$$\Rightarrow \mathbb{P} \left(\lim_{p \rightarrow +\infty} \left\| \frac{1}{p} [Z(S(p))] - \underline{c_2(L, h_L)} \right\|_{U, -2} = 0 \right) = 1.$$

(A bit stronger than Thm 4)

Rk: On \mathbb{C}_{Δ} or \mathbb{D}_{Δ} , \mathbb{C}_{Δ}^n , GAF, by Offord, Sodin, Zelditch
 for cpt Kähler, Shiffman-Zelditch-Zeier 2008
Non-cpt, Druzbog-L. - Marinov 2023

$$Y \subset X \quad U \subset\subset X$$

codom 1

$$\begin{cases} \text{Vol}_{2n-2}^L(W \cap Y) = \int_{W \cap Y} \frac{c_1(L, h_U)^{n-1}}{(n-1)!} \\ \text{Vol}_{2n}^L(W) = \int_W \frac{c_1(L, h_U)^n}{n!} \end{cases}$$

$$\text{Vol}_{2n-2}^L(Z(S_p) \cap W) = \sum_{\substack{V \subset X \\ \text{hypersurface} \\ V \subset Z(S_p)}} \text{ord}_V(S_p) \text{Vol}_{2n-2}^L(V \cap W)$$

\uparrow vanishing order

Cor 2: $U \subset\subset X$, ∂U measure zero w.r.t. $dV = \frac{\omega^n}{n!}$

$$\mathbb{P} \left(\left| \frac{1}{p} \text{Vol}_{2n-2}^L(Z(S(U^p)) \cap W) - n \text{Vol}_{2n}^L(W) \right| > \delta \right) \leq e^{-C_{\delta, U} p^{n+1}}$$

Take $\delta = n \text{Vol}_{2n}^L(W)$

$$e^{-C_{\delta, U} p^{n+1}} \leq \mathbb{P} \left(\underbrace{Z(S(U^p)) \cap W = \emptyset}_{\text{Hole Probability}} \right) \leq e^{-C_{\delta, U} p^{n+1}}$$

$\left\{ \begin{array}{l} C_{\delta, U} \text{ Zebiel} \\ \text{opt Kähler} \end{array} \right.$

the lower bounds exists in special cases or with more assumptions.

Rk: (C, D) hole probability can be explicitly computed

Peres-Vorag 2005 Acta
Nishry 2010 IMRN

idea: ∂U zero measure

$$\exists \text{ in } \varphi_2 \leq \chi_U \leq \varphi_1$$

\uparrow indicator function

s.t.

$$\left| \int_{X \setminus U} \frac{c_1(L, h_U)^n}{n!} - \text{Vol}_{2n}^L(U) \right| < \frac{\delta}{2}$$

Put

$$\varphi_p \frac{C(L, h_L)^{n-1}}{J^{(n-1)!}}$$

test forms on LDE

#

§ 2

A key intermediate result to prove Thm 6

For $U \subset \subset X$ $0 \neq s_p \in H^0(X, L^p)$

local sup norm

$$M_p^U(s_p) := \sup_{B \in \mathcal{U}} \|s_p(B)\|_{h_p} > 0$$

Prop 1: $\forall \delta > 0, \exists C = C(U, \delta) > 0$ s.t.

$$\mathbb{P} \left(\left| \log M_p^U(s_p) \right| > \delta p \right) \leq e^{-C p^{n+1}}$$

sketched proof of Thm 6 via Prop 1.

By Prop 1 $\exists E_p(\delta, U) \subset \Omega$ (probability space)

$$\mathbb{P}(E_p(\delta, U)) \leq e^{-C p^{n+1}}$$

on $E_p(\delta, U)^c$: we can always find $x \in U$ s.t.

$$\left| \log \|s_p(x)\|_{h_p} \right| \leq \delta p$$

hence $\Rightarrow \log \|s_p(x)\|_{h_p} \geq -\delta p$
or $\leq \delta p$

A consequence:

$$\mathbb{P} \left(\int_U \left| \log \|s_p(z)\|_{h_p} \right| d\nu(z) > \delta p \right) \leq e^{-C p^{n+1}}$$

To get this:

$$\textcircled{1} \begin{cases} \log = \log^+ - \log^- \\ |\log| = \log^+ + \log^- \end{cases}$$

$$\log^+ |S(L^p(x))|_{hp} \leq \log |M_p^u(S(L^p))|$$

$$\rightarrow \mathbb{P} \left(\int_{\Delta} \log^+ > \delta p \right) \leq e^{-cp^{n+1}}$$

$$\log^- = \log^+ - \log$$

② We need to prove

$$(*) \quad \mathbb{P} \left(- \int_{\Delta} \log |S(L^p(x))|_{hp} d\nu(x) \geq K \delta p \right) \leq e^{-cp^{n+1}}$$

large number > 0

we use sub-mean inequality:

$\log f$ subharmonic on \mathbb{C}^n

$$P_r(\xi, z) = \frac{r^{2n-2}}{\Delta} \frac{r^2 - |\xi|^2}{|\xi - z|^{2n}} \quad \begin{array}{l} |z| = r \\ |z| < r \end{array}$$

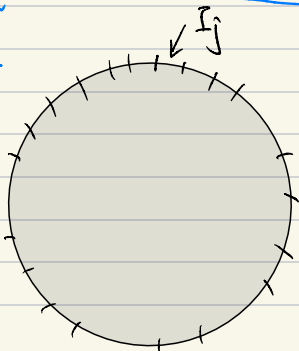
Poisson kernel for $\mathbb{D}(0, r) \subset \mathbb{C}^n$

$$\forall |\xi| < r$$

1st estimate $= \log f(\xi) \leq \int_{|z|=r} P_r(\xi, z) \log f(z) dG_r(z)$

These two are "bounded" by Prop 1

$$\leq \frac{2}{\Delta} |\log f|_{\text{local sep}} - \int_{|z|=r} P_r(\xi, z) |\log f(z)| dG_r(z)$$



$$\begin{aligned} \cup I_j &\sim \partial \mathbb{D}(0, r) \\ \text{diameter}(I_j) &\sim \delta^{2n+2} \\ \int_{G_r(I_j)} &\sim 1 \end{aligned}$$

$$\exists c > 0, \forall \xi_j \quad d(\xi_j, \underset{\Delta}{(r-\delta)} \underset{\Delta}{I_j}) < \delta^{2n+2}$$

2nd estimate

$$-\int_{|z|=r} \log f(z) dG_r(z) \leq - \sum_j \underbrace{G_r(I_j)}_{\leq \frac{\sum G_r(I_j) \delta p}{1}} \log f(\xi_j) + c\delta \int_{|z|=r} |\log f(z)| dG_r(z)$$

Consider

$$f(z) = \underset{\Delta}{S(L^p)} + \underbrace{C^\infty}_{\text{fit}}$$

$\leq Kp$
by the 1st estimate

$$\Rightarrow \text{except an even} \leq \sum_{\text{finite}} e^{-c p^{2n+1}} \sim e^{-c' p^{2n+1}}$$

we have $\forall j$

$$\begin{cases} \int_{I_j} \log |S(L^p)(z_j)|_{h_p} \geq -\delta p & \Rightarrow (*) \\ \int_{\cup} |\log |S(L^p)(z)|_{h_p}| d\nu(z) \leq Kp \end{cases}$$

Using (*)

and Poincaré-Lelong formula: $\text{supp } \varphi \subset \cup \quad \varphi \in \mathcal{H}^{(n-1, n-1)}$

$$\begin{aligned} & \left| \langle \frac{1}{p} [Z(S(L^p))] - G_{(L, h_L)}, \varphi \rangle \right| \\ &= \left| \frac{1}{p\pi} \int_X \log |S(L^p)(z)|_{h_p} \partial \bar{\partial} \varphi \right| \\ &\leq \frac{|\varphi|_2}{p\pi} \int_{\cup} |\log |S(L^p)(z)|| d\nu(z) \end{aligned}$$

$$|\partial \bar{\partial} \varphi| \leq |\varphi|_2 \cdot d\nu(z)$$

$$\leq \delta \cdot p \text{ except an even of probability } \leq e^{-c p^{2n+1}}$$

$$\Rightarrow \mathbb{P} \left(\left\| \frac{1}{p} [Z(S(L^p))] - c(L, h_U) \right\|_{W, 2} > \frac{\delta}{c} \right) \leq e^{-c p^{n+1}}$$

$\delta > 0$ arbitrary.

§ 3 Proof of Prop 1: normalized Bergman kernel.

$$\left\{ \begin{array}{l} \textcircled{A} \quad \mathbb{P} \left(M_p^W(S(L^p)) \geq e^{\delta p} \right) \leq e^{-c p^{n+1}} \\ \textcircled{B} \quad \mathbb{P} \left(M_p^W(S(L^p)) \leq e^{-\delta p} \right) \leq e^{-c p^{n+1}} \end{array} \right.$$

①: local C^0 -norm \approx local L^2 -norm

$$M_p^W(S(L^p))^{2p^n} \leq C_p^{p^n} \left[\int_U |S(L^p)(z)|_{h_p}^2 d\mu(z) \right]^{p^n}$$

U very small

e_L hol. frame on (U)

$$\sup_U |e_{\pm}|_{h_L} = 1$$

$$\textcircled{2)} = \inf_U |e_{\pm}|_{h_L} \leq 1$$

1 if U sufficiently small

$$C_p \approx \frac{C_U}{\gamma^{2p}}$$

$$\mathbb{E} \left[\left\| S(L^p) \right\|_{L^2(U)}^{2p^n} \right] \leq \frac{1}{\gamma^{2p^{n+1}}} \exp(C p^n \log p)$$

$C > 0$

$\forall x \in U$
 $S(L^p)(x) \in L_x \approx \mathbb{C}$
 Gaussian r.v.
 of variance $B_p(x)$

$$\mathbb{E} \left[\left\| S(L^p) \right\|_{L^2(U)}^{2p^n} \right] \approx |U|^{p^n} \int_U \mathbb{E} \left[|S(L^p)(z)|^{2p^n} \right] d\mu(z)$$

$(2p^n)$ -th moments of Gaussian random variable

$$\begin{aligned}
 \mathbb{P} \left(\underbrace{M_p^W(S(L^p))}^{2p^n} \geq e^{2\delta p^{n+1}} \right) &\leq e^{-2\delta p^{n+1}} \mathbb{E} \left[M_p^W(S(L^p))^{2p^n} \right] \\
 &\leq e^{-2\delta p^{n+1}} \underbrace{C_W^{p^n}}_{\underbrace{p^{2p^{n+1}}}} \mathbb{E} \left[\|S(L^p)\|_{L^2(\Omega)}^{2p^n} \right] \\
 &\leq e^{-2\delta p^{n+1} - 6 \log v p^{n+1} + C p^n \log p}
 \end{aligned}$$

When W is small, we can take $-6 \log v < \delta$ $\Rightarrow e^{-\delta p^{n+1} + C p^n \log p} \approx \dots \Rightarrow \textcircled{A}$

\textcircled{B} Def: normalized Bergman kernel

$$N_p(x, y) = \frac{|B_p(x, y)| / h_p(x) \bar{h}_p(y)^*}{\sqrt{|B_p(x, x)|} \sqrt{|B_p(y, y)|}} \in [0, 1]$$

$S(L^p)$ holomorphic Gaussian field.

N_p correlation function of $S(L^p)$

Thm 7: $W \subset \subset X$, for $k \geq 1$, $b \geq \sqrt{6k/\pi}$

\triangle we have for $p \gg 0$

$$N_p(x, y) = \underbrace{O(p^{-k})}$$

$$\underbrace{d(x, y)}_{x, y \in W} \geq \underbrace{b \sqrt{\frac{\log p}{p}}}$$

$$\underbrace{(1 + R_p(x, y)) \exp\left(-\frac{p\pi}{2} d(x, y)^2\right)}$$

$$d(x, y) \leq \underbrace{b \sqrt{\frac{\log p}{p}}}$$

$$\sup_{\substack{x, y \in W \\ d(x, y) \leq b \sqrt{\frac{\log p}{p}}}} |R_p(x, y)| \rightarrow 0 \text{ as } p \rightarrow +\infty$$

$$d(x, y) \leq b \sqrt{\frac{\log p}{p}}$$

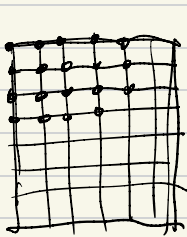
the proof will be given in next lecture

How to prove (B) from Thm 7.

$$F_p \subset W \quad \text{"subset"}$$

$$\mathbb{P}(M_p^H(\mathbb{S}(L^p)) \leq e^{-sp}) \leq \mathbb{P}\left(\sup_{x \in F_p} |S(L^p)\omega|_{L_p} \leq e^{-sp}\right)$$

$$\subset W \quad (B) \approx \leq e^{-cp^{n+1}}$$



$\subset \mathbb{R}^n$ local chart

$$F_p = \left\{ x_p(v) = \frac{c_0}{d^p} v : v = (v_1, \dots, v_{2n}) \right. \\ \left. \begin{array}{l} v_j \leq \epsilon d^p \\ \epsilon \text{ small} > 0 \end{array} \right\} \in \mathbb{Z}^{2n}$$

to be determined

$$\#F_p \sim p^n$$

$$\text{Def: } \xi(x_p(v)) = \frac{\langle S(L^p)x_p(v), \lambda_p(x_p(v)) \rangle_{L_p}}{\sqrt{B_p(x_p(v))}} \sim N_{\mathbb{C}}(0, 1)$$

$\leftarrow \|\cdot\|_{L_p} = 1$

$$\left\{ \xi(x_p(v)) \right\}_{x_p(v) \in F_p}$$

Covariance

$$\Delta_p(u, v) = \mathbb{E}[\xi(x_p(u)) \xi(x_p(v))]$$

$$= N_p(x_p(u), x_p(v))$$

$$= \begin{cases} \leq 2 \exp\left(-\frac{C_0^2}{2} |u-v|^2\right) & \frac{C_0}{d^p} |u-v| \leq b \sqrt{\frac{\log p}{p}} \\ \Theta(p^{-n-1}) & \frac{C_0}{d^p} |u-v| \geq b \sqrt{\frac{\log p}{p}} \end{cases}$$

$$k = n+1$$

$$b = \sqrt{\frac{6 \ln 4}{\pi}}$$

Take $\underline{C_0 \gg 0}$

$$\Delta_p(u, u) = 1$$

$$\sum_{v \neq u} |\Delta_p(u, v)| \leq \frac{1}{2}$$

$\Delta_p =$
 Δ_p is a
 positive invertible
 matrix

$$\begin{bmatrix} 1 & & & & \\ & \ddots & & & \\ & & \ddots & & \\ & & & \ddots & \\ & & & & 1 \end{bmatrix}$$

~ 0

$\#F_p =$ the length
of this
vector

max spec $(\Delta_p^{-1}) \leq 2$

$$\Delta_p^{-1/2} (\xi_{\alpha_p(u)})_{\alpha_p(u) \in F_p} = (\eta_1, \dots, \eta_{\#F_p}) \rightsquigarrow \text{i.i.d. } \sim \mathcal{N}(0, 1)$$

↑
Gaussianity

$$\mathbb{P} \left(\max_{F_p} |\xi_{\alpha_p(u)}| \leq e^{-\delta p} \right) \stackrel{\substack{\sqrt{2} \cdot \sqrt{\#F_p} \\ \text{SS}}}{\sim} \mathbb{P} \left(\max_j |\eta_j| \leq \sqrt{2p^n} e^{-\delta p} \right)$$

$$\leq \mathbb{P} \left(\max_j |\eta_j| \leq \sqrt{2p^n} e^{-\delta p} \right)$$

$$\approx \left[\pi \left(\sqrt{2p^n} e^{-\delta p} \right)^2 \right]^{\#F_p}$$

$$\approx C e^{-C' \delta p^{n+1}} \rightarrow \textcircled{B}$$

$$\frac{\#F_p}{\#} \sim p^n$$

