

Circular law for matrices with iid entries

$$A \in M_n(\mathbb{C})$$

$\lambda_1, \dots, \lambda_n$ - eigenvalues $\mu_{A_n} = \frac{1}{n} (\sqrt{\lambda_1} + \dots + \sqrt{\lambda_n})$

s_1, \dots, s_n - singular values (eigenvalues of $\sqrt{AA^*}$) $\nu_{A_n} = \frac{1}{n} (\sqrt{s_1} + \dots + \sqrt{s_n})$

For Z -r.v. taking complex values $\text{Var}(Z) = \mathbb{E}(|Z|^2) - |\mathbb{E}Z|^2$

Our framework $(X_{i,j})_{i,j=1}^n$ infinite table of iid r.v.'s on \mathbb{C} with $\text{Var}(X_{i,j}) = 1$. $X = (X_{i,j})_{i,j=1}^n$ $n^{-2}X$

$\nu_{n^{-2}X} \rightarrow \mathcal{Q}_2$ - quarter circle

$\mu_{n^{-2}X}$

$$\int s^2 d\nu_{n^{-2}X}(s) = \frac{1}{n^2} \sum_{i,j=1}^n |X_{i,j}|^2 \xrightarrow{\text{L.L.N.}} \mathbb{E}(|X_{1,1}|^2)$$

Thm 2.2 Circular law

a.s. $\mu_{n^{-2}X} \xrightarrow{w} \mathcal{C}_1$

$$\frac{\sum_{i=1}^n X_{i,1}}{n} \quad \frac{\sum_{i=1}^n X_{i,1}}{n} = \frac{1}{n} \sum_{i=1}^n X_{i,1}$$

Proof of Thm 2.2

1) Prove for $X_{i,j}$ complex Gaussians

$$U_{\mu}^{(z)} = \int_{\mathbb{C}} \log |z-w| d\mu(w)$$

$$U_{\mu}^{(z)} = -\frac{1}{n} \log \det \sqrt{(A-zI)(A-zI)^*} = -\int_0^{\infty} \log(t) d\nu_{A-zI}(t)$$

f is uniformly integrable (U.I.) for a sequence $(\eta_n)_{n \geq 1}$ when

$$\lim_{t \rightarrow \infty} \sup_{n \geq 1} \int_{\{|f| > t\}} |f| d\eta_n = 0$$

$$\lim_{t \rightarrow \infty} \limsup_n \int_{\{|f| > t\}} |f| d\eta_n = 0$$

$\eta_n \xrightarrow{w} \eta$ and f is u.i. for η_n then
 f is η integrable and $\lim_{n \rightarrow \infty} \int f d\eta_n = \int f d\eta$

Lemma 4.3 (Hermitization). Let $(A_n)_{n \geq 1}$ be a sequence of complex random matrices where A_n is $n \times n$ for every $n \geq 1$. Suppose that there exists a family of (non-random) probability measures $(\nu_z)_{z \in \mathbb{C}}$ on \mathbb{R}_+ such that, for a.a. $z \in \mathbb{C}$, a.s.

- (i) $\nu_{A_n - zI} \xrightarrow{w} \nu_z$ as $n \rightarrow \infty$
- (ii) \log is uniformly integrable for $(\nu_{A_n - zI})_{n \geq 1}$.

Then there exists a probability measure $\mu \in \mathcal{P}(\mathbb{C})$ such that

- (j) a.s. $\mu_{A_n} \xrightarrow{w} \mu$ as $n \rightarrow \infty$
- (jj) for a.a. $z \in \mathbb{C}$,

$$U_\mu(z) = - \int_0^\infty \log(s) d\nu_z(s).$$

Moreover, if the convergence (i) and the uniform integrability (ii) both hold in probability for a.a. $z \in \mathbb{C}$ (instead of for a.a. $z \in \mathbb{C}$, a.s.), then (j-jj) hold with the a.s. weak convergence in (j) replaced by the weak convergence in probability.

Corollary 4.10

For all $z \in \mathbb{C}$ there exists a prob. measure ν_z depending only on z such that a.s.

$$\nu_{n^{-1}X - zI} \xrightarrow{w} \nu_z \text{ as } n \rightarrow \infty$$

Lemma 4.11 (Count of small sing. values)

There exists $c_0 > 0$ and $0 < \gamma < 1$ such that a.s. for $n \gg 1$ and $n^{1-\gamma} \leq i \leq n-1$ and all $M \in M_n(\mathbb{C})$

$$s_{n-i}(n^{-1}X + M) \geq c_0 \frac{i}{n}$$

- There are at most $n^{1-\gamma}$ small singular values

Lemma 4.12

For every $a, d > 0$ there exists $b > 0$ such that if M is a deterministic complex $n \times n$ matrix with $\text{sp}(M) \subseteq n^d$ then

$$\mathbb{P}(s_n(X+M) \leq n^{-b}) \leq n^{-a}$$

In particular there exists $b > 0$ such that a.s. for $n \gg 1$

$$s_n(X+M) \geq n^{-b}$$

$$\int |\log(s)| d\nu_{n^{-1}X - zI}(s) \leq \int \frac{(\log(s))^2}{t} d\nu_{n^{-1}X - zI}(s) \leq \frac{1}{t} \int_0^\infty s^p \nu_{n^{-1}X - zI}(s)$$

$$\int_{|\log(s)| > t} |\log(s)| dV_{n^{-\frac{1}{2}}X - 2I}(s) \leq \int \frac{(\log(s))^2}{t} dV_{n^{-\frac{1}{2}}X - 2I}(s) \leq \frac{1}{t} \int_0^{+\infty} s^p V_{n^{-\frac{1}{2}}X - 2I}(s) \leq \frac{1}{t} \int_0^{+\infty} s^{-p} V_{n^{-\frac{1}{2}}X - 2I}(s)$$

$\frac{|\log(s)|}{t} > 1$

 $\frac{(\log(s))^2}{t} \leq s^p + s^{-p} \quad p > 0$

So it suffices to show that

$$\limsup_{n \rightarrow \infty} \int s^{\pm p} dV_{n^{-\frac{1}{2}}X - 2I}(s) < +\infty$$

Recall that $\max_{1 \leq i \leq n} |s_i(A) - s_i(B)| \leq \rho_1(A - B)$

$$A = (n^{-\frac{1}{2}}X - 2I) \quad B = n^{-\frac{1}{2}}X$$

$$|s_i(n^{-\frac{1}{2}}X - 2I) - s_i(n^{-\frac{1}{2}}X)| \leq |z|$$

$$s_i(n^{-\frac{1}{2}}X - 2I) \leq s_i(n^{-\frac{1}{2}}X) + |z|$$

So $\limsup \int s^p dV_{n^{-\frac{1}{2}}X - 2I}(s) < +\infty$ for all $p \leq 2$

$$\limsup_n \int s^{-p} dV_{n^{-\frac{1}{2}}X - 2I}(s) < +\infty$$

$$\frac{1}{n} \sum_{i=1}^n s_i^{-p} \leq \frac{1}{n} \sum_{i=1}^{n - \lfloor n^{\frac{\gamma}{b'}} \rfloor} s_i^{-p} + \frac{1}{n} \sum_{i=n - \lfloor n^{\frac{\gamma}{b'}} \rfloor + 1}^n s_i^{-p}$$

$$\leq \frac{1}{n} C_0^{-p} \sum_{i=1}^n \left(\frac{n}{i}\right)^p + \frac{1}{n} \frac{n^{1-\gamma} \cdot n^{pb'}}{n^{pb' - \gamma}}$$

fine for $p < 1$

$$s_n \left(\frac{1}{\sqrt{n}} X - 2I \right) = \frac{1}{\sqrt{n}} s_n \left(X - 2\sqrt{n}I \right) \geq \frac{1}{\sqrt{n}} n^{-b} = n^{-\left(b + \frac{\gamma}{b'}\right)}$$

$$pb' - \gamma < 0 \quad p < \frac{\gamma}{b'}$$

for $p < \min\left\{1, \frac{\gamma}{b'}\right\}$ we get $\limsup \int s^{-p} dV_{n^{-\frac{1}{2}}X - 2I}(s) < +\infty$

$$U_\mu(z) = - \int_0^{+\infty} \log(s) dV_z(s)$$

since V_z does not depend on the distr. of $X_{i,j}$ it is as in the gaussian case and

$n < 11, \rightarrow$

a.s. $\mu_{n^{-1}X} \xrightarrow{w} C_1$ as in the gaussian case and \square

Corollary 4.10

For all $Z \in \mathcal{E}$ there exists a prob. measure ν_Z depending only on Z such that a.s. $\nu_{n^{-1}X - Z} \xrightarrow{w} \nu_Z$ as $n \rightarrow \infty$

Pf

Step 1

$$f: \mathbb{R} \rightarrow \mathbb{R}$$

$$\|f\|_{TV} = \sup_{\mathcal{C} \in \mathcal{C}} \sum_{C \in \mathcal{C}} |f(x_{C-1}) - f(x_C)|$$

$$\text{if } f' \in L^1(\mathbb{R}) \quad \|f\|_{TV} = \int |f'(x)| dx$$

Lemma 4.18 (Concentr. of measure for sing. val. emp. measure)
 If M is an $n \times n$ complex random matrix with independent rows then for any $f: \mathbb{R} \rightarrow \mathbb{R}$ going to 0 at $\pm\infty$ with $\|f\|_{TV} \leq 1$ and every $t > 0$

$$\mathbb{P}\left(\left|\int f d\nu_M - \mathbb{E} \int f d\nu_M\right| \geq t\right) \leq 2e^{-2nt^2}$$

Proof

Ljolskii inequality: $A, B \in M_n(\mathbb{C})$ with $\text{rank}(A-B) \leq k$
 then $s_{i+k}(A) \geq s_i(B) \geq s_{i+k}(A)$

$$F_A(\cdot) = \nu_A((-\infty, \cdot])$$

$$F_B(\cdot) = \nu_B((-\infty, \cdot])$$

$$\|F_A - F_B\|_{\infty} \leq \frac{\text{rank}(A-B)}{n}$$