

Gaussian integration by parts method

Let $X \in \mathbb{R}^{n \times n}$ be a Wigner matrix. Let $G_n(z) = (X - zI)^{-1}$ be its resolvent.

Recall the Stieltjes transform of the BSD of X ,

$$m_n(z) = \frac{1}{n} \text{Tr} G_n(z) = \frac{1}{n} \sum_{i=1}^n \frac{1}{\lambda_i(X) - z}.$$

For any $z \in \mathbb{C}^+$:

$$I = (X - zI) \cdot (X - zI)^{-1} = (X - zI) G_n(z).$$

$$\text{Then } G_n(z) = \frac{1}{z} (X G_n(z) - I)$$

$$m_n(z) = \frac{1}{nz} \text{Tr} (X G_n(z) - I) \\ = \frac{1}{nz} \sum_{i,j=1}^n X_{ij} G_n(z)_{ji} - \frac{1}{z}.$$

Here, $G_n(z)_{ji}$ is a function of X_{ij} and other independent variables.

If X_{ij} were Gaussian, integration by parts yields a formula for $\mathbb{E}[X_{ij} G_n(z)_{ji}]$, which may be used to derive a recursion for $\mathbb{E}[m_n(z)]$.

Lemma (integration by parts): Let $\xi \sim N(0,1)$. If $f: \mathbb{R} \rightarrow \mathbb{C}$ is differentiable, with $|f(x) \cdot e^{-\frac{x^2}{2}}| \rightarrow 0$ as $x \rightarrow \pm\infty$, then

$$\mathbb{E}[\xi \cdot f(\xi)] = \mathbb{E}[f'(\xi)].$$

$$\text{Proof: } \mathbb{E}[\xi \cdot f(\xi)] = \int_{-\infty}^{\infty} \underbrace{f(x)}_u \cdot \underbrace{x \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx}_{dv} \\ = \left[-f(x) \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \right]_{-\infty}^{\infty} + \int_{-\infty}^{\infty} f'(x) \cdot \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} dx = \mathbb{E}[f'(\xi)].$$

Cor: Let $\xi \sim N(0, \sigma^2)$. Under the preceding conditions,

$$\mathbb{E}[\xi f(\xi)] = \sigma^2 \mathbb{E}[f'(\xi)].$$

Proof: Write $\xi = \sigma \cdot \zeta$. Apply the preceding to $\zeta \mapsto f(\sigma \zeta)$. Then

$$\mathbb{E}[\xi f(\xi)] = \sigma \mathbb{E}[\zeta \cdot f(\sigma \zeta)] = \sigma^2 \mathbb{E}[f'(\sigma \zeta)] = \sigma^2 \mathbb{E}[f'(\xi)].$$

Lemma: Let $X \sim \text{GOE}(n)$, i.e. $X_{ij} = \frac{1}{\sqrt{n}} z_{ij}$ where

$$z_{ij} \sim \begin{cases} N(0,1) & i \neq j \\ N(0,2) & i = j. \end{cases}$$

Let $G_n(z) = (X - zI)^{-1}$, $m_n(z) = \frac{1}{n} \text{Tr} G_n(z)$. Then for all $z \in \mathbb{C}^+$,

$$\mathbb{E}[m_n(z)] = -\frac{1}{z} - \frac{1}{z} \mathbb{E}[m_n(z)^2] - \frac{1}{nz} \mathbb{E}[\text{Tr} G_n(z)^2].$$

Proof: Write $X = \frac{1}{\sqrt{2n}}(W + W^T)$, where $W \in \mathbb{R}^{n \times n}$ is asymmetric with $N(0,1)$ entries. Note that

$$X - zI = \frac{1}{\sqrt{2n}} \left(\sum_{i,j=1}^n W_{ij} \Delta_{ij} + W_{ji}^T \Delta_{ji} \right) - zI$$

where $\Delta_{ij} \in \mathbb{R}^{n \times n}$ has (i,j) entry 1 and remaining entries 0.

$$\text{Then } \frac{\partial}{\partial W_{ij}} (X - zI) = \frac{1}{\sqrt{2n}} (\Delta_{ij} + \Delta_{ji}^T) \in \mathbb{R}^{n \times n}.$$

• If $A(t) \in \mathbb{C}^{n \times n}$ has entries differentiable in the parameter $t \in \mathbb{R}$, and is invertible at t , then

$$\Rightarrow I = A(t) \cdot A(t)^{-1}$$

$$\Rightarrow 0 = \frac{d}{dt} (A(t) \cdot A(t)^{-1})$$

$$= A'(t) \cdot A(t)^{-1} + A(t) \cdot \frac{d}{dt} (A(t)^{-1})$$

$$\text{So } \frac{d}{dt} (A(t)^{-1}) = -A(t)^{-1} \cdot A'(t) \cdot A(t)^{-1}.$$

Then, applying to $A(W_{ij}) = X - zI$,

$$\frac{\partial}{\partial W_{ij}} (X - zI)^{-1} = -(X - zI)^{-1} \left(\frac{1}{\sqrt{2n}} (\Delta_{ij} + \Delta_{ji}) \right) (X - zI)^{-1}.$$

$$\text{So } \frac{\partial}{\partial W_{ij}} G_n(z) = -\frac{1}{\sqrt{2n}} G_n(z) (\Delta_{ij} + \Delta_{ji}) G_n(z).$$

$$\frac{\partial}{\partial W_{ij}} G_n(z)_{ij} = -\frac{1}{\sqrt{2n}} (G_n(z)_{ii} G_n(z)_{jj} + G_n(z)_{ji} G_n(z)_{ij}).$$

$$\frac{\partial}{\partial W_{ij}} G_n(z)_{ji} = \frac{\partial}{\partial W_{ij}} G_n(z)_{ij} \text{ because } G_n \text{ is symmetric.}$$

Applying integration by parts,

$$\begin{aligned} \mathbb{E}[X_{ij} G_n(z)_{ji}] &= \frac{1}{\sqrt{2n}} \mathbb{E}[(W_{ij} + W_{ji}) G_n(z)_{ji}] \\ &= \frac{1}{\sqrt{2n}} \mathbb{E}\left[\frac{\partial}{\partial W_{ij}} G_n(z)_{ji} + \frac{\partial}{\partial W_{ji}} G_n(z)_{ji}\right] \\ &= -\frac{1}{n} (\mathbb{E}[G_n(z)_{ii} G_n(z)_{jj}] + \mathbb{E}[G_n(z)_{ji}^2]). \end{aligned}$$

$$\begin{aligned} \text{Then } \mathbb{E}[m_n(z)] &= -\frac{1}{z} + \frac{1}{n^2 z} \sum_{i,j=1}^n \mathbb{E}[X_{ij} G_n(z)_{ji}] \\ &= -\frac{1}{z} - \frac{1}{n^2 z} \mathbb{E}\left[\left(\sum_{i=1}^n G_n(z)_{ii}\right)^2\right] - \frac{1}{n^2 z} \mathbb{E}[T_n G_n(z) G_n(z)] \\ &= -\frac{1}{z} - \frac{1}{z} \mathbb{E}[m_n(z)^2] - \frac{1}{n^2 z} \mathbb{E}[T_n G_n(z)^2]. \end{aligned}$$

Prop: $\left| \frac{1}{n} \text{Tr } G_n(z)^2 \right| \leq \frac{1}{(\text{Im } z)^2}$, for $z \in \mathbb{C}^+$.

Proof: Let $X = \sum_{i=1}^n \lambda_i v_i v_i^T$ be the spectral decomposition of X .

Then $(X - zI)^{-1} = \sum_{i=1}^n \frac{1}{\lambda_i - z} v_i v_i^T$ is the spectral decomposition of $G_n(z)$.

$(X - zI)^{-2} = \sum_{i=1}^n \frac{1}{(\lambda_i - z)^2} v_i v_i^T$ that of $G_n(z)^2$.

$$\text{So } \left| \frac{1}{n} \text{Tr } G_n(z)^2 \right| = \left| \frac{1}{n} \sum_{i=1}^n \frac{1}{(\lambda_i - z)^2} \right| \leq \frac{1}{n} \sum_{i=1}^n \frac{1}{|\lambda_i - z|^2} \leq \frac{1}{(\text{Im } z)^2}.$$

Prop: For constants $C, c > 0$, and any $z \in \mathbb{C}^+$,

$$\mathbb{P}[|m_n(z) - \mathbb{E}m_n(z)| \geq t] \leq 4e^{-ct^2(\text{Im } z)^4 n^2}$$

$$\text{and } |\mathbb{E}[m_n(z)^2] - \mathbb{E}[m_n(z)]^2| \leq \frac{C}{n(\text{Im } z)^3}.$$

Proof: Recall for $x, x' \in \mathbb{R}$, $\left| \frac{1}{x-z} - \frac{1}{x'-z} \right| \leq \frac{|x-x'|}{(\text{Im } z)^2}$. So

$x \mapsto \text{Re } \frac{1}{x-z}$ and $x \mapsto \text{Im } \frac{1}{x-z}$ are both $\frac{1}{(\text{Im } z)^2}$ -Lipschitz.

From concentration of spectral measure,

$$\mathbb{P}\left[\left|\int \text{Re } \frac{1}{x-z} d\mu_n(x) - \mathbb{E} \int \text{Re } \frac{1}{x-z} d\mu_n(x)\right| \geq \frac{t}{2}\right] \leq 2e^{-ct^2(\text{Im } z)^4 n^2}.$$

Similarly for $\text{Im } \frac{1}{x-z}$, and combining yields the bound for $m_n - \mathbb{E}m_n$.

For the second statement, let $\bar{m}_n(z) = \mathbb{E}m_n(z)$.

$$\begin{aligned} |\mathbb{E}[m_n(z)^2] - \bar{m}_n(z)^2| &= |\mathbb{E}[(m_n(z) - \bar{m}_n(z))(m_n(z) + \bar{m}_n(z))]| \\ &\leq \mathbb{E}[|m_n(z) - \bar{m}_n(z)| \cdot |m_n(z) + \bar{m}_n(z)|] \\ &\leq \frac{2}{\text{Im } z} \mathbb{E}[|m_n(z) - \bar{m}_n(z)|]. \end{aligned}$$

(We've used $|m_n(z)| \leq \frac{1}{\text{Im } z}$.)

For any non-negative random variable Z ,

$$\mathbb{E}[Z] = \mathbb{E}\left[\int_0^\infty \mathbb{1}_{\{Z \geq t\}} dt\right] = \int_0^\infty \mathbb{P}[Z \geq t] dt.$$

$$\text{So } \mathbb{E}[|m_n(z) - \bar{m}_n(z)|] = \int_0^\infty \mathbb{P}[|m_n(x) - \bar{m}_n(z)| \geq t] dt$$

$$\leq 4 \int_0^\infty e^{-ct^2(\text{Im } z)^4 n^2} dt$$

$$= 2 \int_{-\infty}^\infty e^{-ct^2(\text{Im } z)^4 n^2} dt \leq \frac{C}{n(\text{Im } z)^2}.$$

Proof #3 of semicircle law, GOE case:

By the preceding results,

$$\mathbb{E}[m_n(z)] = -\frac{1}{z} - \frac{1}{z} \mathbb{E}[m_n(z)]^2 + r_n$$

where $|r_n| \leq \frac{C}{n} \left(\frac{1}{(\operatorname{Im} z)^3} + \frac{1}{(\operatorname{Im} z)^4} \right)$. Let $m(z)$ be the root of

$$m(z) = -\frac{1}{z} - \frac{1}{z} m(z)^2 \Leftrightarrow m(z)^2 + z m(z) + 1 = 0$$

with positive imaginary part. Since $\mathbb{E}[m_n(z)]$ has positive imaginary part, this implies $|\mathbb{E}[m_n(z)] - m(z)| \rightarrow 0$.

From the preceding proposition, $|m_n(z) - \mathbb{E}[m_n(z)]| \rightarrow 0$ a.s. by Borel-Cantelli.

So $m_n(z) \rightarrow m(z)$ a.s.

This may be extended to other models. For example:

Let $X \sim \text{GOE}(n)$. Let $A \in \mathbb{R}^{n \times n}$ be deterministic, symmetric, with the ESD of A converging to μ_A weakly. What is the limit ESD of $X+A$?

Denote $G_n(z) = (X+A-zI)^{-1}$, $m_n(z) = \frac{1}{n} \operatorname{Tr} G_n(z)$.

$$I = (X+A-zI) \cdot G_n(z)$$

$$\Rightarrow (zI-A) \cdot G_n(z) = X G_n(z) - I$$

We apply integration by parts without taking the trace:

$$\begin{aligned} \mathbb{E}[(X G_n(z))_{ij}] &= \sum_{k=1}^n \mathbb{E}[X_{ik} G_n(z)_{kj}] \\ &= -\frac{1}{n} \sum_{k=1}^n \mathbb{E}[(G_n(z) (\Delta_{ik} + \Delta_{ki}) G_n(z))_{kj}] \\ &= -\frac{1}{n} \sum_{k=1}^n \mathbb{E}[G_n(z)_{ki} G_n(z)_{kj} + G_n(z)_{kk} G_n(z)_{ij}] \\ &= -\frac{1}{n} \mathbb{E}[(G_n(z)^2)_{ij}] = \mathbb{E}[m_n(z) \cdot G_n(z)_{ij}]. \end{aligned}$$

$$\text{So } \mathbb{E}[X G_n(z)] = -\frac{1}{n} \mathbb{E}[G_n(z)^2] - \mathbb{E}[m_n(z) G_n(z)].$$

$$\text{Writing } G_n(z) = \sum_{i=1}^n \frac{1}{\lambda_i - z} v_i v_i^T, \quad \|G_n(z)\| = \sup_i \frac{1}{|\lambda_i - z|} \leq \frac{1}{\operatorname{Im} z}.$$

Then $\|\frac{1}{n} \mathbb{E}[G_n(z)^2]\| \leq \frac{1}{n(\operatorname{Im} z)^2}$. Same argument shows concentration of $m_n(z)$.

$$\begin{aligned} \text{Then } \|\mathbb{E}[(m_n(z) - \bar{m}_n(z)) G_n(z)]\| \\ \leq \mathbb{E}[|m_n(z) - \bar{m}_n(z)|] \cdot \frac{1}{\operatorname{Im} z} \leq \frac{C}{n(\operatorname{Im} z)^3}. \end{aligned}$$

$$\text{Thus } \mathbb{E}[X G_n(z)] = -\bar{m}_n(z) \cdot \mathbb{E}[G_n(z)] + R_n$$

where $\|R_n\| \leq \frac{1}{n} \cdot \text{poly}\left(\frac{1}{\operatorname{Im} z}\right)$.

$$\text{We get } (zI-A) \cdot \mathbb{E}[G_n(z)] = -\bar{m}_n(z) \cdot \mathbb{E}[G_n(z)] - I + R_n$$

$$\Rightarrow \mathbb{E}[G_n(z)] = (zI + \bar{m}_n(z)I - A)^{-1} (-I + R_n)$$

$$= (A - (z + \bar{m}_n(z))I)^{-1} + \tilde{R}_n$$

where $\|\tilde{R}_n\| \leq \frac{1}{n} \cdot \text{poly}\left(\frac{1}{\operatorname{Im} z}\right)$ also. Now taking $\frac{1}{n} \operatorname{Tr}$,

$$\bar{m}_n(z) = \frac{1}{n} \operatorname{Tr} (A - (z + \bar{m}_n(z))I)^{-1} + O\left(\frac{1}{n}\right)$$

$$\rightarrow m_A(z + \bar{m}_n(z)),$$

where m_A is the Stieltjes transform of μ_A (the limit ESD of A).

One may show, for any μ_A with bounded support, that

$$m(z) = m_A(z + m(z))$$

has a unique root $m(z) \in \mathbb{C}^+$ when $z \in \mathbb{C}^+$. Then $m_n(z) \rightarrow m(z)$ a.s.

Then (Pastur '72): The equation $m(z) = m_A(z + m(z))$ has a unique root $m(z) \in \mathbb{C}^+$ for $z \in \mathbb{C}^+$, which defines the Stieltjes transform of a measure μ on $\mathbb{R}$. For $X \sim \text{GOE}(n)$ and A deterministic, symmetric, if the ESD of A converges

weakly to μ_A w/ Stieltjes transform $m_A(z)$ as $n \rightarrow \infty$, then that of $X+A$ converges weakly a.s. to μ .

Writing μ_{sc} as the semi-circle law on $[-2, 2]$, we write

$\mu = \mu_{sc} \boxplus \mu_A$, the additive free convolution of μ_{sc} and μ_A .

Lindeberg exchange method

We illustrate the idea by proving the following quantitative CLT:

Thm: Let $X_1, \dots, X_n$ be iid random variables with $E[X_i] = 0$, $E[X_i^2] = 1$, $E[|X_i|^3] = m_3$. Let $e: \mathbb{R} \rightarrow \mathbb{R}$ be three-times differentiable, with uniformly bounded third derivative. Then for $\xi \sim N(0, 1)$,

$$|E[e(\frac{X_1 + \dots + X_n}{\sqrt{n}})] - E[e(\xi)]| \leq \frac{m_3 + \sqrt{\frac{8}{\pi}}}{6\sqrt{n}} \cdot \sup_{x \in \mathbb{R}} |e'''(x)|.$$

Proof: Write $\xi = \frac{Z_1 + \dots + Z_n}{\sqrt{n}}$, where $Z_i \sim N(0, 1)$. We "swap" X_i one by one for Z_i . Let

$$S^{(i)} = \frac{X_1 + \dots + X_i + Z_{i+1} + \dots + Z_n}{\sqrt{n}}.$$

Then we bound $|E[e(S^{(i)})] - E[e(S^{(i-1)})]|$

for $i = n, n-1, \dots, 1$, and then sum over i .

Let $T^{(i)} = \frac{X_1 + \dots + X_{i-1} + Z_{i+1} + \dots + Z_n}{\sqrt{n}}$. Then

• $T^{(i)}$ is independent of both X_i and Z_i .

• $S^{(i)} = T^{(i)} + \frac{X_i}{\sqrt{n}}$, $S^{(i-1)} = T^{(i)} + \frac{Z_i}{\sqrt{n}}$.

$$S_0 \quad e(S^{(i)}) = e(T^{(i)}) + e'(T^{(i)}) \cdot \frac{X_i}{\sqrt{n}} + \frac{e''(T^{(i)})}{2} \cdot \frac{X_i^2}{n} + \frac{e'''(T^{(i)})}{6} \cdot \frac{X_i^3}{n^{3/2}}$$

$$e(S^{(i-1)}) = e(T^{(i)}) + e'(T^{(i)}) \cdot \frac{Z_i}{\sqrt{n}} + \frac{e''(T^{(i)})}{2} \cdot \frac{Z_i^2}{n} + \frac{e'''(T^{(i)})}{6} \cdot \frac{Z_i^3}{n^{3/2}}$$

For some (random) $X_i, W_i \in \mathbb{R}$.

Since $T^{(i)}$ is independent of X_i, Z_i , and the first two moments of X_i and Z_i match, $|E[e(S^{(i)})] - E[e(S^{(i-1)})]| = |E[\frac{e'''(X_i)}{6} \cdot \frac{X_i^3}{n^{3/2}} - \frac{e'''(W_i)}{6} \cdot \frac{Z_i^3}{n^{3/2}}]|$

$$\leq \frac{\sup |e'''(x)|}{6n^{3/2}} \cdot \underbrace{E[|X_i|^3 + |Z_i|^3]}_{= m_3 + \sqrt{\frac{8}{\pi}}}$$

Summing over $i = 1, \dots, n$ gives the result.

We now apply this to the Stieltjes transform of a random matrix:

Lemma: Let $X, \tilde{X} \in \mathbb{R}^{n \times n}$ be two Wigner matrices such that

$$\bullet E[X_{ij}] = E[\tilde{X}_{ij}] = 0 \text{ for all } i, j$$

$$\bullet E[X_{ij}^2] = E[\tilde{X}_{ij}^2] = \frac{1}{n} \text{ for all } i \neq j$$

$$\bullet E[|X_{ij}|^3], E[|\tilde{X}_{ij}|^3] \leq \frac{C}{n^{3/2}} \text{ for all } i, j \text{ and}$$

$$E[|X_{ii}|^2], E[|\tilde{X}_{ii}|^2] \leq \frac{C}{n} \text{ for all } i.$$

Let $m_A(z)$ and $\tilde{m}_A(z)$ be the Stieltjes transforms of the ESDs of $X+A$ and $\tilde{X}+A$, for any deterministic, symmetric $A \in \mathbb{R}^{n \times n}$. Then

$$|E[m_A(z)] - E[\tilde{m}_A(z)]| \leq C \left(\frac{1}{\sqrt{n} (\operatorname{Im} z)^4} + \frac{1}{n (\operatorname{Im} z)^3} \right)$$

for all $z \in \mathbb{C}^+$.

Proof: We swap the entries $(X_{ij})_{i \in j}$ for $(\tilde{X}_{ij})_{i \in j}$, one at a time.
 First do the off-diagonal: Order the pairs (i, j) with $i \in j$ arbitrarily.

Define $Y^{(ij)} \in \mathbb{R}^{n \times n}$ as the matrix where

- $Y_{kl}^{(ij)} = X_{kl}$ and $Y_{lk}^{(ij)} = X_{lk}$ for (k, l) preceding (i, j) .

- $Y_{kl}^{(ij)} = \tilde{X}_{kl}$ and $Y_{lk}^{(ij)} = \tilde{X}_{lk}$ for (k, l) following (i, j) .

- $Y_{ij}^{(ij)} = 0$ and $Y_{ji}^{(ij)} = 0$.

Let $X^{(ij)} = Y^{(ij)} + X_{ij}(\Delta_{ij} + \Delta_{ji})$

$\tilde{X}^{(ij)} = Y^{(ij)} + \tilde{X}_{ij}(\Delta_{ij} + \Delta_{ji})$

So $X^{(ij)} \mapsto \tilde{X}^{(ij)}$ swaps the (i, j) and (j, i) entries of X for $\tilde{X}$.

Let $m_n^{(ij)} = \frac{1}{n} \text{Tr} (X^{(ij)} + A - zI)^{-1} = \frac{1}{n} \text{Tr} G_n^{(ij)}$

$\tilde{m}_n^{(ij)} = \frac{1}{n} \text{Tr} (\tilde{X}^{(ij)} + A - zI)^{-1} = \frac{1}{n} \text{Tr} \tilde{G}_n^{(ij)}$

$h_n^{(ij)} = \frac{1}{n} \text{Tr} (Y^{(ij)} + A - zI)^{-1} = \frac{1}{n} \text{Tr} H_n^{(ij)}$

Key identity: $A^{-1} - B^{-1} = A^{-1}(B - A)B^{-1}$

Repeatedly applying this,

$$A^{-1} - B^{-1} = A^{-1}(B - A)B^{-1}$$

$$= A^{-1}(B - A)A^{-1} - A^{-1}(B - A)(A^{-1} - B^{-1})$$

$$= A^{-1}(B - A)A^{-1} - A^{-1}(B - A)A^{-1}(B - A)B^{-1}$$

$$= \dots = \sum_{k=1}^{\ell} (-1)^{k-1} (A^{-1}(B - A))^k B^{-1} \text{ for any } \ell \geq 1.$$

In particular: Since $(X^{(ij)} + A - zI) - (Y^{(ij)} + A - zI) = X_{ij}(\Delta_{ij} + \Delta_{ji}) \equiv D$,

$$h_n^{(ij)} - m_n^{(ij)} = \frac{1}{n} \text{Tr} [H_n^{(ij)} - G_n^{(ij)}]$$

$$= \frac{1}{n} \text{Tr} [H_n^{(ij)} \Delta H_n^{(ij)} - H_n^{(ij)} \Delta H_n^{(ij)} \Delta H_n^{(ij)} + H_n^{(ij)} \Delta H_n^{(ij)} \Delta H_n^{(ij)} \Delta G_n^{(ij)}]$$

Similarly, letting $\tilde{D} = \tilde{X}_{ij}(\Delta_{ij} + \Delta_{ji})$

$$h_n^{(ij)} - \tilde{m}_n^{(ij)} = \frac{1}{n} \text{Tr} [H_n^{(ij)} - \tilde{G}_n^{(ij)}]$$

$$= \frac{1}{n} \text{Tr} [H_n^{(ij)} \tilde{D} H_n^{(ij)} - H_n^{(ij)} \tilde{D} H_n^{(ij)} \tilde{D} H_n^{(ij)} + H_n^{(ij)} \tilde{D} H_n^{(ij)} \tilde{D} H_n^{(ij)} \tilde{D} \tilde{G}_n^{(ij)}]$$

Use now: $H_n^{(ij)}$ is independent of both X_{ij} and $\tilde{X}_{ij}$, and X_{ij} and $\tilde{X}_{ij}$ have matching first and second moments.

Then $|E[m_n^{(ij)} - \tilde{m}_n^{(ij)}]|$

$$\leq |E[\frac{1}{n} \text{Tr} H_n^{(ij)} \tilde{D} H_n^{(ij)} \tilde{D} H_n^{(ij)} \tilde{D} \tilde{G}_n^{(ij)}]| \quad \text{I}$$

$$+ |E[\frac{1}{n} \text{Tr} H_n^{(ij)} \Delta H_n^{(ij)} \Delta H_n^{(ij)} \Delta G_n^{(ij)}]| \quad \text{II}$$

Note that

$$\text{I} = |E[\frac{X_{ij}^3}{n} \cdot \sum_{k=1}^n (H(\Delta_{ij} + \Delta_{ji}) H(\Delta_{ij} + \Delta_{ji}) H(\Delta_{ij} + \Delta_{ji}) G)_{kk}]|$$

and $\sum_{k=1}^n (H \Delta_{ij} H \Delta_{ij} H \Delta_{ij} G)_{kk} = \sum_{k=1}^n H_{ki} H_{ji} H_{ji} G_{jk} = (GH)_{ji} H_{ji} H_{ji}$

Since $\|H\| \leq \frac{1}{\text{Im } z}$, $\|G\| \leq \frac{1}{\text{Im } z}$, this is bounded by $\|GH\| \cdot \|H\| \cdot \|H\| \leq \frac{1}{(\text{Im } z)^4}$.

Then $\text{I} \leq \frac{1}{n(\text{Im } z)^4} E[|X_{ij}|^3] \leq \frac{C}{n^{5/2}(\text{Im } z)^4}$. Similarly for II.

There are $\binom{n}{2}$ such swaps, so summing over (i,j) shows

$$|E[m_n(z)] - E[\tilde{m}_n(z)]| \leq \frac{C}{\sqrt{n}(\operatorname{Im} z)^4}$$

where $\tilde{m}_n(z) = \frac{1}{n} \operatorname{Tr}(\tilde{X} + A - zI)^{-1}$ and $\tilde{X}$ has off-diag entries equal to $\tilde{X}$, diag entries equal to X .

Now do the same for the diagonal: Define $Y^{(i)} \in \mathbb{R}^{n \times n}$ st.

$$Y_{ij}^{(i)} = X_{ij} \text{ for } j \leq i$$

$$Y_{ij}^{(i)} = \tilde{X}_{ij} \text{ for } j > i$$

$$Y_{ii}^{(i)} = 0$$

$$Y_{jk}^{(i)} = \tilde{X}_{jk} \text{ for } j \neq k.$$

$$\text{Set } X^{(i)} = Y^{(i)} + \sum_{j=1}^i \tilde{X}_{ji} \Delta_{ji}, \quad \tilde{X}^{(i)} = Y^{(i)} + \sum_{j=1}^i \tilde{X}_{ji} \tilde{\Delta}_{ji},$$

define analogously $m_n^{(i)}, \tilde{m}_n^{(i)}, h_n^{(i)}$. Then, since first moments match,

$$\begin{aligned} |E m_n^{(i)} - E \tilde{m}_n^{(i)}| &\leq |E[\frac{1}{n} \operatorname{Tr} H_n^{(i)} \Delta H_n^{(i)} \Delta G_n^{(i)}]| \\ &\quad + |E[\frac{1}{n} \operatorname{Tr} H_n^{(i)} \tilde{\Delta} H_n^{(i)} \tilde{\Delta} \tilde{G}_n^{(i)}]| \\ &\leq \frac{1}{n(\operatorname{Im} z)^3} (E[|X_{11}|^2] + E[|\tilde{X}_{11}|^2]) \leq \frac{C}{n^2(\operatorname{Im} z)^3} \end{aligned}$$

Number of diagonal swaps is n , so summing over i ,

$$|E[\tilde{m}_n(z)] - E[\tilde{m}_n(z)]| \leq \frac{C}{n(\operatorname{Im} z)^3}.$$

Combining yields the result.

Proof #3 of semi-circle law, when X satisfies LSI:

By this result, $E[m_n(z)]$ is the same for X as for the GOE.

Then the integration-by-parts argument shows $E[m_n(z)] \rightarrow m(z)$.

Concentration of $m_n(z)$ follows from the LSI, since $x \mapsto \frac{1}{x-z}$ is $\frac{1}{(\operatorname{Im} z)^2}$ -Lipschitz.

So $m_n(z) \rightarrow m(z)$ a.s.

Problem 3: Recall that if $A \in \mathbb{R}^{n \times n}$ has ESD converging to μ_A , with Stieltjes transform $m_A(z) = \int \frac{1}{x-z} d\mu_A(x)$, then the ESD of $X+A$ for Wigner X converges to μ w/ Stieltjes transform satisfying

$$m(z) = m_A(z + m(z)) \quad (*)$$

(a) For fixed $z \in \mathbb{C}^+$, devise a numerical algorithm to solve (*) for $m(z) \in \mathbb{C}^+$, and implement this algorithm.

(b) Write a program to plot an approximation for the probability density function of μ . How close is the ESD of $X+A$ to this density for $n=100$? 500? 2500?

(c) Derive the analogue of (*) for the matrix $\sigma X + A$, for arbitrary scaling $\sigma > 0$, and let the limit ESD be μ_σ .

For fixed A (numerically), how does the density of μ_σ change as σ increases from 0 to ∞ ? What happens to the number of intervals of support of μ_σ ?

Explore this in some settings where μ_A has multiple disjoint intervals of support.