

Outliers in the Spectrum of Spiked Deformations of Unitarily Invariant Random Matrices

Random Matrices and their Applications

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- Results in the sample covariance setting: from Baik-Ben Arous-Péché (2005) to Bai-Yao (2012).

Historical overview - 2

- Definition of an additive analogue (Péché 2006):
$$W + \text{Diag}(\underbrace{\theta_1, \dots, \theta_1}_{k_1}, \dots, \underbrace{\theta_J, \dots, \theta_J}_{k_J}, 0, \dots, 0).$$

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- Results for the additive analogue: from Péché (2006) to Renfrew-Soshnikov (preprint 2012).

Two works

Largest eigenvalues of finite rank perturbations of unitarily invariant random matrices.

Theorem (Benaych-Georges and Nadakuditi 2009)

Almost surely,

$$\lambda_j \rightarrow_{N \rightarrow +\infty} \begin{cases} G_\nu^{-1}(1/\theta_j) & \text{if } \theta_j > 1/\lim_{z \downarrow b} G_\nu(z), \\ b & \text{otherwise,} \end{cases}$$

while for each fixed $j > r$, almost surely, $\lambda_j \rightarrow_{N \rightarrow +\infty} b$. Here,

$$G_\nu: \mathbb{C} \setminus \text{supp}(\nu) \rightarrow \mathbb{C}, \quad G_\nu(z) = \int_{\mathbb{R}} \frac{d\nu(t)}{z - t},$$

is the Cauchy-Stieltjes transform of the limit distribution ν , and b is the maximum of its support.

Two works

Eigenvalues of full rank perturbations of Wigner matrices.

Theorem (Capitaine-Donati-Martin-Féral and F. 2010)

Let $H(z) := z + \sigma^2 G_\mu(z)$, then there are k_j eigenvalues converging almost surely to $H(\theta_j)$ iff $H'(\theta_j) > 0$, where μ is the limit distribution of the perturbation, and σ^2 is the variance of the entries of the Wigner matrix.

Model

$$X_N = U_N^* B_N U_N,$$

- $B_N = \text{Diag}(\beta_1^{(N)}, \dots, \beta_N^{(N)})$,
- U_N is a random $N \times N$ unitary matrix distributed according to Haar measure.

Model

$$X_N = A_N + U_N^* B_N U_N,$$

- $A_N = \text{Diag}(\underbrace{\theta_1, \dots, \theta_1}_{k_1}, \dots, \underbrace{\theta_J, \dots, \theta_J}_{k_J}, \alpha_1^{(N)}, \dots, \alpha_{N-r}^{(N)}),$
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Question: Spectrum of $X_N = A_N + U_N^* B_N U_N$?

Assumptions

- $B_N = \text{Diag}(\beta_1^{(N)}, \dots, \beta_N^{(N)});$

$$\mu_{B_N} := \frac{1}{N} \sum_{i=1}^N \delta_{\beta_i^{(N)}} \Rightarrow \nu \in \mathcal{P}_c(\mathbb{R}),$$

$$\max_{1 \leq j \leq N} \text{dist}(\beta_j^{(N)}, \text{supp}(\nu)) \rightarrow_{N \rightarrow \infty} 0.$$

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$$\mu_{A_N} := \frac{1}{N} \sum_{i=1}^N \delta_{\lambda_i(A_N)} \Rightarrow \mu \in \mathcal{P}_c(\mathbb{R}),$$

$$\max_{1 \leq j \leq N-r} \text{dist}(\alpha_j^{(N)}, \text{supp}(\mu)) \rightarrow_{N \rightarrow \infty} 0,$$

$\theta_j \notin \text{supp}(\mu)$ (the so-called **spikes**).

Global behaviour

We will use the usual notation:

$$\mu_{X_N} := \frac{1}{N} \sum_{\lambda \in \text{sp}(X_N)} \delta_\lambda.$$

Asymptotic freeness (Voiculescu 91, Speicher 93)

Under these assumptions,

$$\mu_{X_N} \xrightarrow[N \rightarrow +\infty]{\text{a.s.}} \mu \boxplus \nu.$$

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What is this \boxplus operation?

Free convolution of measures

Given $\tau \in \mathcal{P}_c(\mathbb{R})$, one defines:

Stieltjes transform

$$G_\tau(z) = \int_{\mathbb{R}} \frac{d\tau(t)}{z - t}, \quad z \notin \mathbb{R}.$$

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Definition

One calls free convolution of μ and ν the probability measure $\mu \boxplus \nu \in \mathcal{P}_c(\mathbb{R})$ characterized by:

$$R_{\mu \boxplus \nu}(z) = R_\mu(z) + R_\nu(z).$$

Subordination

Theorem (Voiculescu 93, Biane 98)

There is a unique analytic map $\omega : \mathbb{C}^+ \rightarrow \mathbb{C}^+$ such that:

$$\forall z \in \mathbb{C}^+, G_{\mu \boxplus \nu}(z) = G_{\mu}(\omega(z)).$$

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Lemma

The map ω has an extension to \mathbb{C} so that:

- (a) ω is continuous on $\mathbb{C}^+ \cup \mathbb{R}$;
- (b) $\omega(\{\infty\} \cup \mathbb{R} \setminus \text{supp}(\mu \boxplus \nu)) \subseteq \{\infty\} \cup \mathbb{R} \setminus \text{supp}(\mu)$;
- (c) $\forall z \in \mathbb{C} \setminus \mathbb{R}, \overline{\omega(z)} = \omega(\bar{z})$;
- (d) ω is meromorphic on $\mathbb{C} \setminus \text{supp}(\mu \boxplus \nu)$.

A definition

Definition

For each $j \in \{1, \dots, J\}$, define O_j the set of solutions in $\mathbb{R} \setminus \text{supp}(\mu \boxplus \nu)$ of the equation

$$\omega(\rho) = \theta_j, \quad (1)$$

and

$$O = \bigcup_{1 \leq j \leq J} O_j.$$

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Theorem (Collins-Male 2011)

If $r = 0$ (no spikes), then almost surely,

$$\forall \eta > 0, \exists N_0 \in \mathbb{N}, \forall N \geq N_0, \text{sp}(X_N) \subseteq K_\eta,$$

where $K_\eta := \{x \in \mathbb{R} \mid d(x, \text{supp}(\mu \boxplus \nu)) \leq \eta\}$.

Main result

In the general case, one proves:

Theorem

The following results hold almost surely:

- for each $\rho \in O_j$, for all small enough $\varepsilon > 0$, for all large enough N ,

$$\text{card}\{\text{sp}(X_N) \cap]\rho - \varepsilon; \rho + \varepsilon[\} = k_j;$$

- for almost all $\eta > 0$, for all small enough $\varepsilon > 0$, for large enough N ,

$$\text{sp}(X_N) \cap \mathbb{C} \setminus K_\eta \subset \bigcup_{\rho \in O \cap \mathbb{C} \setminus K_\eta}]\rho - \varepsilon; \rho + \varepsilon[.$$

Generalization

Remark

Actually, our result holds for

$$\tilde{X}_N = \tilde{A}_N + \tilde{B}_N,$$

where \tilde{A}_N and \tilde{B}_N are independent random Hermitian matrices, provided the distribution of \tilde{B}_N is invariant by conjugation by unitary matrices.

Comments

In the particular case of a finite rank deformation A_N , one recovers the result of Benaych-Georges and Nadakuditi (BGN 2009) on the convergence of the largest eigenvalues:

Theorem (Benaych-Georges and Nadakuditi 2009)

Almost surely,

$$\lambda_j \rightarrow_{N \rightarrow +\infty} \begin{cases} G_\nu^{-1}(1/\theta_j) & \text{if } \theta_j > 1/\lim_{z \downarrow b} G_\nu(z), \\ b & \text{otherwise,} \end{cases}$$

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while for each fixed $j > r$, almost surely, $\lambda_j \rightarrow_{N \rightarrow +\infty} b$.

Indeed, in that case, $\mu = \delta_0$ and $\omega(z) = \frac{1}{G_\nu(z)}$.

Comments

In the case of a full rank deformation of a GUE, one recovers the result of Capitaine, Donati-Martin, Féral and F. (CDFFF 2010).

Theorem (Capitaine-Donati-Martin-Féral and F. 2010)

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Let $H(z) := z + \sigma^2 G_\mu(z)$, then there are k_j eigenvalues converging almost surely to $H(\theta_j)$ iff $H'(\theta_j) > 0$.

Indeed, in that case, ν is semicircular, ω is invertible with inverse H .

Comments

Remark

This result illustrates that the free probabilistic interpretation of outliers, discovered in (CDFF 2010) generalizing the one in (BGN 2009), is a general principle.

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Remark

It is noteworthy that, in this situation, a **simple** spike may create **several** outliers.

Sketch of proof-1

We use the following decomposition:

$$A_N = A'_N + A''_N,$$

$$A'_N = \text{Diag}(\alpha, \dots, \alpha, \alpha_1^{(N)}, \dots, \alpha_{N-r}^{(N)}),$$

$$A''_N = {}^t P \Theta P,$$

where P is the $r \times N$ matrix defined by

$$P = (I_r | 0_{r \times (N-r)}),$$

Θ is the $r \times r$ matrix

$$\Theta = \text{Diag}(\underbrace{\theta_1 - \alpha, \dots, \theta_1 - \alpha}_{k_1}, \dots, \underbrace{\theta_J - \alpha, \dots, \theta_J - \alpha}_{k_J}),$$

and $\alpha \in \text{supp}(\mu)$.

Sketch of proof-2

$$\det(\lambda I_N - X_N) = \det(\lambda I_N - (A'_N + U_N^* B_N U_N)) \det(I_N - R_N(\lambda)^t P \Theta P),$$

where

$$R_N(\lambda) = (\lambda I_N - (A'_N + U_N^* B_N U_N))^{-1}. \quad (2)$$

Using that, for rectangular matrices $X \in M_{N,r}(\mathbb{C})$, $Y \in M_{r,N}(\mathbb{C})$, one has $\det(I_N - XY) = \det(I_r - YX)$, one obtains:

$$\det(\lambda I_N - X_N) = \det(\lambda I_N - (A'_N + U_N^* B_N U_N)) \det(I_r - P R_N(\lambda)^t P \Theta).$$

Hence, the outliers of X_N are precisely the zeros of $\det(M_N)$ outside the support of $\mu \boxplus \nu$, where

$$M_N := I_r - P R_N^t P \Theta. \quad (3)$$

Sketch of proof-3

Key point

Using Hurwitz's theorem, the zeros of $\det(M_N)$ will cluster towards those of $\det(M)$, where M is the almost sure uniform limit of M_N .

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Key point

Using Hurwitz's theorem, the zeros of $\det(M_N)$ will cluster towards those of $\det(M)$, where M is the almost sure uniform limit of M_N .

- By concentration arguments, $M_N - I_r - P\mathbb{E}(R_N)^t P\Theta$ tends to 0 as N goes to infinity.
- It is known that $\mathbb{E}(R_N)$ is diagonal (Kargin 2011). Actually, it is a polynomial in A'_N .
In particular, $P\mathbb{E}(R_N)^t P$ is a scalar matrix.

Sketch of proof-4

Define ω_N so that:

$$P\mathbb{E}(R_N)^t P = \frac{1}{\omega_N - \alpha} I_r.$$

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Define ω_N so that:

$$P\mathbb{E}(R_N)^t P = \frac{1}{\omega_N - \alpha} I_r.$$

Then $(\omega_N)_{N \in \mathbb{N}}$ is a normal sequence of analytic functions, whose limit points l shall satisfy the subordination equation:

$$G_{\mu \boxplus \nu}(z) = G_{\mu}(l(z)),$$

which has the subordination map ω as a unique solution.

Sketch of proof-5

So M_N almost surely uniformly converges to:

$$M := I_r - \frac{1}{\omega - \alpha} \Theta. \quad (4)$$

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And z such that $\det(M(z)) = 0$ are precisely solutions of $\omega(z) = \theta_j$ for some j , concluding the proof.

Thank you for your attention!