

Exact separation phenomenon for the eigenvalues of Information-Plus-Noise type models and application to spiked models

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"Information plus noise" type model

$\sigma > 0$.

$n \in \mathbb{N}$, $N \in \mathbb{N}$, $n \leq N$.

$\{X_{ij}, i \in \mathbb{N}^*, j \in \mathbb{N}^*\}$ independent random complex variables with mean 0 and variance 1.

X_N : a $n \times N$ matrix such that $(X_N)_{ij} = X_{ij}$.

A_N : a $n \times N$ deterministic matrix.

$$M_N = \left(\sigma \frac{X_N}{\sqrt{N}} + A_N\right) \left(\sigma \frac{X_N}{\sqrt{N}} + A_N\right)^*$$

Dozier-Silverstein (cas i.i.d) (2007), Xie (2012) proved that under the assumptions:

- $n/N \rightarrow c \in]0, 1]$ when $N \rightarrow +\infty$,
- $\lim_{N \rightarrow +\infty} \frac{1}{N^2} \sum_{1 \leq i \leq n; 1 \leq j \leq N} \mathbb{E} \left(|X_{ij}|^2 \mathbf{1}_{|X_{ij}| \geq \eta \sqrt{N}} \right) = 0$,
- $\mu_{A_N A_N^*} = \frac{1}{n} \sum_{i=1}^n \delta_{\lambda_i(A_N A_N^*)} \xrightarrow{N \rightarrow +\infty} \nu$ weakly,

almost surely,

$$\mu_{M_N} = \frac{1}{n} \sum_{i=1}^n \delta_{\lambda_i(M_N)} \xrightarrow{N \rightarrow +\infty} \mu_{\sigma, \nu, c} \text{ weakly.}$$

When $\nu = \delta_0$, $\mu_{\sigma, \delta_0, c}$ is the Marchenko-Pastur distribution:

$$\frac{d\mu_c}{dx}(x) = \frac{1}{2\pi cx} \sqrt{(b-x)(x-a)} 1_{[a,b]}(x)$$

$$a = (1 - \sqrt{c})^2, \quad b = (1 + \sqrt{c})^2.$$

$\mu_{\sigma,\nu,c}$ is deterministic and characterized by its Stieltjes transform: $\mathfrak{g}_{\mu_{\sigma,\nu,c}}$ defined for any $z \in \mathbb{C} \setminus \mathbb{R}$ by

$$\mathfrak{g}_{\mu_{\sigma,\nu,c}}(z) = \int_{\mathbb{R}} \frac{d\mu_{\sigma,\nu,c}(x)}{z - x},$$

which is the unique solution m among the Stieltjes transforms of probability measures on $[0; +\infty[$, of the equation: $\forall z \in \mathbb{C}^+$,

$$m(z) = \int \frac{1}{(1 - \sigma^2 cm(z))z - \frac{t}{1 - \sigma^2 cm(z)} - \sigma^2(1 - c)} d\nu(t).$$

Characterization of $\text{supp}(\mu_{\sigma,\nu,c})$

[Dozier-Silverstein 2007][C.2013]

$$\Phi_{\sigma,\nu,c} : \begin{cases} \mathbb{R} \setminus \text{supp}(\nu) \rightarrow \mathbb{R} \\ x \mapsto x(1 + c\sigma^2 g_\nu(x))^2 + \sigma^2(1 - c)(1 + c\sigma^2 g_\nu(x)) \end{cases} .$$

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$$\mathcal{E}_{\sigma,\nu,c} := \left\{ u \in \mathbb{R} \setminus \text{supp}(\nu), \Phi'_{\sigma,\nu,c}(u) > 0, g_\nu(u) > -\frac{1}{\sigma^2 c} \right\} .$$

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$$\mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c}) = \Phi_{\sigma,\nu,c}(\mathcal{E}_{\sigma,\nu,c})$$

$$\omega_{\sigma,\nu,c} : \begin{cases} \mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c}) \rightarrow \mathbb{R} \\ x \mapsto x(1 - \sigma^2 c g_{\mu_{\sigma,\nu,c}}(x))^2 - \sigma^2(1 - c)(1 - \sigma^2 c g_{\mu_{\sigma,\nu,c}}(x)) \end{cases}$$

$$\mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c}) \begin{matrix} \xrightarrow{\omega_{\sigma,\nu,c}} \\ \xleftarrow{\Phi_{\sigma,\nu,c}} \end{matrix} \mathcal{E}_{\sigma,\nu,c} .$$

For any $y > x$ in $\mathcal{E}_{\sigma,\nu,c}$, $\Phi_{\sigma,\nu,c}(y) > \Phi_{\sigma,\nu,c}(x)$.

Proposition (C. 2013)

Assume that the support of ν is compact and has a finite number of connected components then there exists $p \in \mathbb{N}^*$ and $u_1 < v_1 < u_2 < \dots < u_p < v_p$ (depending on σ, ν, c) such that

$${}^c\mathcal{E}_{\sigma, \nu, c} = \bigcup_{l=1}^p [u_l; v_l].$$

We have moreover:

$$\text{supp}(\nu) \subset \bigcup_{l=1}^p [u_l; v_l],$$

$$\forall l \in \{1, \dots, p\}, [u_l; v_l] \cap \text{supp}(\nu) \neq \emptyset.$$

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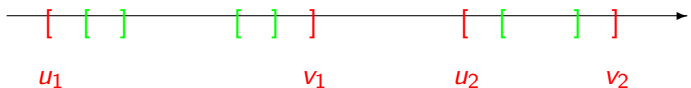
$$\forall l \in \{1, \dots, p\}, [u_l; v_l] \cap \text{supp}(\nu) \neq \emptyset.$$

$$\mathbb{R} \setminus \text{supp}(\mu_{\sigma, \nu, c}) = \Phi_{\sigma, \nu, c}(\mathcal{E}_{\sigma, \nu, c})$$

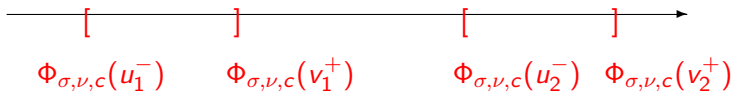
For any $y > x$ in $\mathcal{E}_{\sigma, \nu, c}$, $\Phi_{\sigma, \nu, c}(y) > \Phi_{\sigma, \nu, c}(x)$.

$$\implies \text{supp}(\mu_{\sigma, \nu, c}) = \cup_{l=1}^p [\Phi_{\sigma, \nu, c}(u_l^-); \Phi_{\sigma, \nu, c}(v_l^+)].$$

${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν



support $\mu_{\sigma,\nu,c}$



$$\Phi_{\sigma,\nu,c} : \begin{cases} \mathbb{R} \setminus \text{supp}(\nu) \rightarrow \mathbb{R} \\ x \mapsto x(1 + c\sigma^2 g_\nu(x))^2 + \sigma^2(1 - c)(1 + c\sigma^2 g_\nu(x)) \end{cases} .$$

$$\mathcal{E}_{\sigma,\nu,c} := \left\{ u \in \mathbb{R} \setminus \text{supp}(\nu), \Phi'_{\sigma,\nu,c}(u) > 0, g_\nu(u) > -\frac{1}{\sigma^2 c} \right\} .$$

Aim of this talk: to describe almost surely for large N the spectrum of

$$M_N = \left(\sigma \frac{X_N}{\sqrt{N}} + A_N \right) \left(\sigma \frac{X_N}{\sqrt{N}} + A_N \right)^* .$$

•

Presentation of the results under the following assumptions:

$$n/N \rightarrow c \in]0, 1],$$

X_{ij} i.i.d , with mean 0, variance 1 and finite fourth moment.

$$A_N = \begin{pmatrix} a_1(N) & & (0) \\ & (0) & \\ & \ddots & (0) \\ (0) & & a_n(N) & (0) \end{pmatrix} \quad (1)$$

$\sup_N \|A_N\| < +\infty$, $\mu_{A_N A_N^*} = \frac{1}{n} \sum_{i=1}^n \delta_{\lambda_i(A_N A_N^*)} \xrightarrow{N \rightarrow +\infty} \nu$ weakly, the support of ν is compact and has a finite number of connected components.

(actually the results are obtained under more general technical assumptions)

A deterministic probability measure playing a fundamental role in the study of the spectrum of M_N : $\mu_{\sigma, \mu_{A_N A_N^*}, c_N}$, ($c_N := \frac{n}{N}$) characterized by its Stieltjes transform:

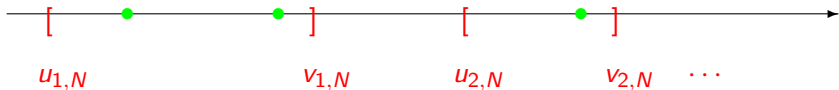
$$z \in \mathbb{C} \setminus \mathbb{R}, \quad g_{\mu_{\sigma, \mu_{A_N A_N^*}, c_N}}(z) = \int_{\mathbb{R}} \frac{d\mu_{\sigma, \mu_{A_N A_N^*}, c_N}(x)}{z - x},$$

unique solution m among Stieltjes transforms of probability measures on $[0; +\infty[$, of the equation: $\forall z \in \mathbb{C}^+$,

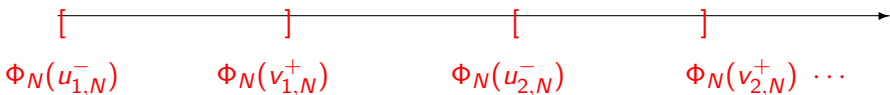
$$m(z) = \int \frac{1}{(1 - \sigma^2 c_N m(z))z - \frac{t}{1 - \sigma^2 c_N m(z)} - \sigma^2(1 - c_N)} d\mu_{A_N A_N^*}(t).$$

We have also $\mu_{\sigma, \mu_{A_N A_N^*}, c_N} \xrightarrow{N \rightarrow +\infty} \mu_{\sigma, \nu, c}$ weakly.

$c\mathcal{E}_{\sigma, \mu_{A_N A_N^*}, c_N}$ support $\mu_{A_N A_N^*}$



support $\mu_{\sigma, \mu_{A_N A_N^*}, c_N}$

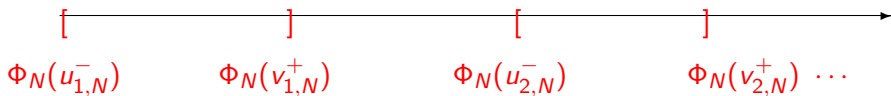


$$\Phi_N(x) := \Phi_{\sigma, \mu_{A_N A_N^*}, c_N}(x)$$

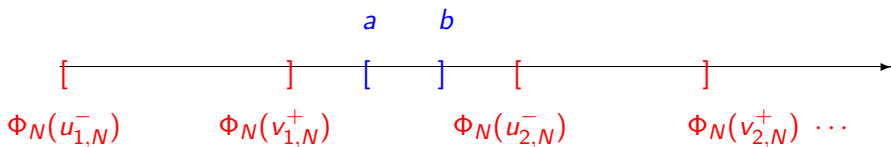
$$= x(1 + c_N \sigma^2 g_{\mu_{A_N A_N^*}}(x))^2 + \sigma^2(1 - c_N)(1 + c_N \sigma^2 g_{\mu_{A_N A_N^*}}(x))$$

$$\mathcal{E}_{\sigma, \mu_{A_N A_N^*}, c_N} = \left\{ u \in \mathbb{R} \setminus \text{supp}(\mu_{A_N A_N^*}), \Phi'_{\sigma, \mu_{A_N A_N^*}, c_N}(u) > 0, g_{\mu_{A_N A_N^*}}(u) > -\frac{1}{\sigma^2 c_N} \right\}$$

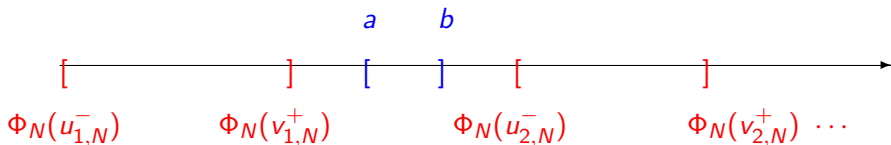
support $\mu_{\sigma, \mu_{A_N A_N^*}, c_N$



support $\mu_{\sigma, \mu_{A_N A_N^*}, c_N$



support $\mu_{\sigma, \mu_{A_N A_N^*}, c_N$



Theorem (Gaussian case: Vallet-Loubaton-Mestre (2010);
non-gaussian case: Bai-Silverstein (2012))

$0 < a < b$. If there exists $0 < \delta < a$ such that for all large N ,
 $]a - \delta; b + \delta[\subset \mathbb{R} \setminus \text{supp}(\mu_{\sigma, \mu_{A_N A_N^*}, c_N})$, then

$$\mathbb{P}[\text{for } N \text{ large enough, } \text{spect}(M_N) \subset \mathbb{R} \setminus [a, b]] = 1.$$

Actually, in the non-gaussian case when A_N is not "diagonal", Bai and Silverstein has an extra technical condition (ii):

Theorem (Bai-Silverstein (2012))

Let $[a, b]$ be such that the couple of the following properties is satisfied:

- (i) there exists $0 < \delta < a$ such that for all large N , $]a - \delta; b + \delta[\subset \mathbb{R} \setminus \text{supp}(\mu_{\sigma, \mu_{A_N A_N^*}, c_N})$
- (ii) A_{Nj} denoting the matrix resulting from removing the j -th column from A_N , there exists $0 < \tau < \delta$ and a positive $d < 1$ such that for all N large, the number of j 's with no eigenvalues of $N/(N-1)A_{Nj}A_{Nj}^*$ appearing in $\omega_{\sigma, \mu_{A_N A_N^*}, c_N}(]a - \tau, b + \tau[)$ is greater than $N - N^d$.

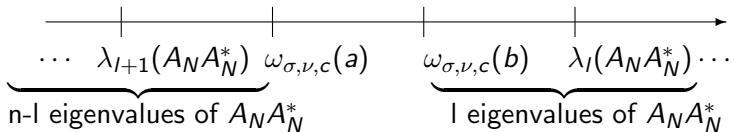
Then

$$\mathbb{P}[\text{for all large } N, \text{spect}(M_N) \subset \setminus [a, b]] = 1.$$

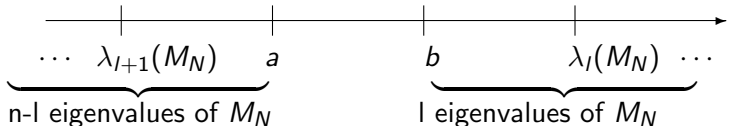
Exact separation phenomenon

$$[a, b] \longleftrightarrow [\omega_{\sigma, \nu, c}(a), \omega_{\sigma, \nu, c}(b)]$$

$$\text{gap in Spect}(M_N) \longleftrightarrow \text{gap in Spect}(A_N A_N^*)$$



Then, almost surely, for large enough N ,



Theorem (gaussian case: Loubaton-Vallet 2010,
non-gaussian case: C. 2013)

For such $[a; b]$ satisfying moreover if $c < 1$, $\omega_{\sigma, \nu, c}(b) > 0$, for all large N ,

$$\omega_{\sigma, \nu, c}([a, b]) = [\omega_{\sigma, \nu, c}(a); \omega_{\sigma, \nu, c}(b)] \subset^c \text{supp}(\mu_{A_N A_N^*}).$$

With the convention $\lambda_0(M_N) = \lambda_0(A_N A_N^*) = +\infty$ and $\lambda_{N+1}(M_N) = \lambda_{N+1}(A_N^*) = -\infty$, let i_N be in $\{0, \dots, n\}$ such that

$$\lambda_{i_N+1}(A_N A_N^*) < \omega_{\sigma, \nu, c}(a) \quad \text{and} \quad \lambda_{i_N}(A_N A_N^*) > \omega_{\sigma, \nu, c}(b).$$

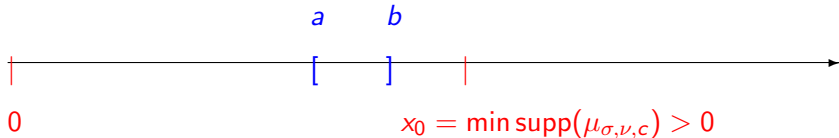
Then

$$P[\text{for all large } N, \lambda_{i_N+1}(M_N) < a \text{ and } \lambda_{i_N}(M_N) > b] = 1.$$

Remark

$c < 1$. Let b be in $\mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c})$ such that $\omega_{\sigma,\nu,c}(b) \leq 0$. Then b is on the left hand side of $\text{supp}(\mu_{\sigma,\nu,c})$.

\implies the restriction $\omega_{\sigma,\nu,c}(b) > 0$ means that we may not be allowed to consider an interval $[a, b]$ such that



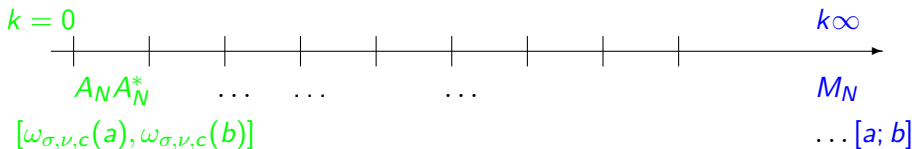
Such an exact separation phenomenon was established:

- sample covariance matrices $X_N \Sigma_N X_N^*$ by Bai-Silverstein 1999,
- deformed Wigner matrices $W_N + A_N$ by Capitaine-Donati-Martin-Féral-Février 2011.

key idea of the proof: To introduce a continuum of matrices between $A_N A_N^*$ and M_N :

$M_N^{(k)} := (\sigma_k \frac{X_N}{\sqrt{N}} + A_N)((\sigma_k \frac{X_N}{\sqrt{N}} + A_N)^*$, $\sigma_0 = 0$, $\sigma_k \uparrow \sigma$ when k goes to infinity and such that $\sigma_{k+1} - \sigma_k$ is small enough.

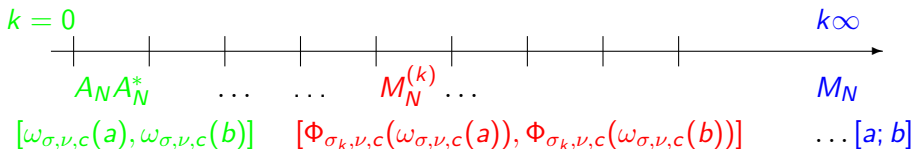
$$\Phi_{\sigma_k, \nu, c}(x) = x(1 + c\sigma_k^2 g_\nu(x))^2 + \sigma_k^2(1 - c)(1 + c\sigma_k^2 g_\nu(x))$$



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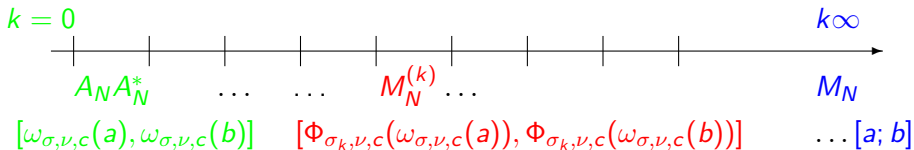
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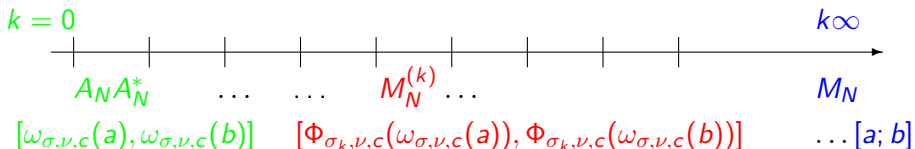


- The intervals $[\Phi_{\sigma_k, \nu, c}(\omega_{\sigma, \nu, c}(a)), \Phi_{\sigma_k, \nu, c}(\omega_{\sigma, \nu, c}(b))]$ split the spectra of $M_N^{(k)}$ exactly in the same way.

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- For k large enough, the interval

$[\Phi_{\sigma_k, \nu, c}(\omega_{\sigma, \nu, c}(a)), \Phi_{\sigma_k, \nu, c}(\omega_{\sigma, \nu, c}(b))]$ splits the spectrum of $M_N^{(k)}$ as $[a; b]$ splits the spectrum of M_N .

Lemma

There exists $m_{a,b} > 0$ such that for all k ,

$$\Phi_{\sigma_k, \nu, c}(\omega_{\sigma_k, \nu, c}(b)) - \Phi_{\sigma_k, \nu, c}(\omega_{\sigma_k, \nu, c}(a)) \geq m_{a,b}.$$

Tool:

Lemma

Let B and C be two $n \times N$ complex matrices, $n \leq N$. For any pair of integers j, k such that $1 \leq j, k \leq n$ and $j + k \leq n + 1$, we have

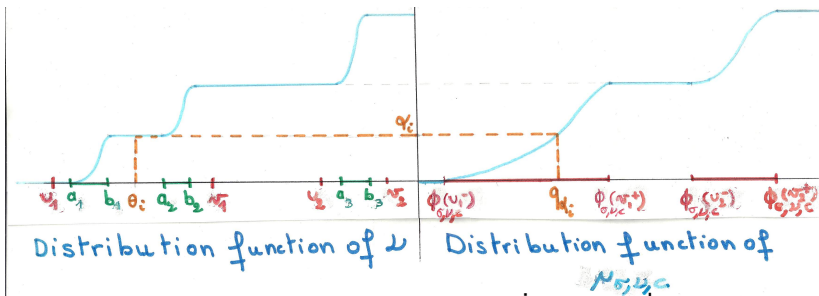
$$\sqrt{\lambda_{j+k-1}[(B+C)(B+C)^*]} \leq \sqrt{\lambda_j(BB^*)} + \sqrt{\lambda_k(CC^*)}.$$

Consequence

Corollary

$$\forall l \in \{1, \dots, p\},$$

$$\mu_{\sigma, \nu, c}([\Phi_{\sigma, \nu, c}(u_l^-); \Phi_{\sigma, \nu, c}(v_l^+)]) = \nu([u_l; v_l])$$



Distribution function of ν

Distribution function of $\mu_{\sigma, \nu, c}$

continuous and strictly increasing on $[\Phi_{\sigma, \nu, c}(u_l^-); \Phi_{\sigma, \nu, c}(v_l^+)]$

When $A_N := 0$,

Theorem

(Geman 1980) (Bai-Yin-Krishnaiah 1988) (Bai-Silverstein-Yin 1988)

$$\lambda_1\left(\frac{X_N X_N^*}{\sqrt{N}}\right) \rightarrow (1 + \sqrt{c})^2 \text{ a.s. when } N \rightarrow +\infty.$$

$$\lambda_n\left(\frac{X_N X_N^*}{N}\right) \rightarrow (1 - \sqrt{c})^2 \text{ a.s. when } N \rightarrow +\infty.$$

\implies Almost surely, for all large N , no eigenvalues of $\frac{X_N X_N^*}{\sqrt{N}}$ outside a neighborhood of the support of the limiting spectral measure μ_c .

EXISTENCE OF OUTLIERS WHEN $A_N \neq 0$?

The equation satisfied by the Stieltjes transform of $\mu_{\sigma,\nu,c}$ leads to

$$\begin{aligned} czg_{\mu_{\sigma,\nu,c}}(z)^2 + (1-c)g_{\mu_{\sigma,\nu,c}}(z) \\ = c\omega_{\sigma,\nu,c}(z)g_{\nu}(\omega_{\sigma,\nu,c}(z))^2 + (1-c)g_{\nu}(\omega_{\sigma,\nu,c}(z)). \end{aligned}$$

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The intuition is that for large N ,

$$\begin{aligned} cxg_{\mu_{M_N}}(x)^2 + (1-c)g_{\mu_{M_N}}(x) \\ \approx cxg_{\mu_{\sigma,\mu_{A_N A_N^*},c}}(x)^2 + (1-c)g_{\mu_{\sigma,\mu_{A_N A_N^*},c}}(x) \end{aligned}$$

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\implies they will be eigenvalues of M_N that separate from the bulk whenever some of the equations $\omega_{\sigma,\nu,c}(x) = \theta_j$, $j = 1, \dots, J$, admits a solution outside the support of $\mu_{\sigma,\nu,c}$.

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$$\mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c}) \begin{array}{c} \xrightarrow{\omega_{\sigma,\nu,c}} \\ \xleftarrow{\Phi_{\sigma,\nu,c}} \end{array} \mathcal{E}_{\sigma,\nu,c}.$$

\Downarrow

$\omega_{\sigma,\nu,c}(x) = \theta_j$ admits a solution outside the support of $\mu_{\sigma,\nu,c}$ if and only if $\theta_j \in \mathcal{E}_{\sigma,\nu,c}$ i.e. $\Phi'_{\nu,\sigma,c}(\theta_j) > 0$, $g_\nu(\theta_j) > -\frac{1}{\sigma^2 c}$.
 There is only one such a solution which is $\Phi_{\sigma,\nu,c}(\theta_j)$.

\implies they will be eigenvalues of M_N that separate from the bulk whenever some of the equations $\omega_{\sigma,\nu,c}(x) = \theta_j, j = 1, \dots, J$, admits a solution outside the support of $\mu_{\sigma,\nu,c}$.

$$\mathbb{R} \setminus \text{supp}(\mu_{\sigma,\nu,c}) \begin{array}{c} \xrightarrow{\omega_{\sigma,\nu,c}} \\ \xleftarrow{\mathcal{E}_{\sigma,\nu,c}} \\ \Phi_{\sigma,\nu,c} \end{array}$$

\Downarrow

$\omega_{\sigma,\nu,c}(x) = \theta_j$ admits a solution outside the support of $\mu_{\sigma,\nu,c}$ if and only if $\theta_j \in \mathcal{E}_{\sigma,\nu,c}$ i.e $\Phi'_{\nu,\sigma,c}(\theta_j) > 0, g_\nu(\theta_j) > -\frac{1}{\sigma^2 c}$.

There is only one such a solution which is $\Phi_{\sigma,\nu,c}(\theta_j)$.

This intuition leads us to introduce the following set

$$\Theta_{\sigma,\nu,c} = \left\{ \theta \in \{\theta_1, \dots, \theta_J\}, \Phi'_{\sigma,\nu,c}(\theta) > 0, g_\nu(\theta) > -\frac{1}{\sigma^2 c} \right\},$$

and for any θ in $\Theta_{\sigma,\nu,c}$, $\rho_\theta = \Phi_{\sigma,\nu,c}(\theta) \notin \text{supp}(\mu_{\sigma,\nu,c})$.

$$\forall \theta \in \Theta_{\sigma, \nu, c},$$

$$\rho_\theta := \Phi_{\sigma, \nu, c}(\theta) \notin \text{supp}(\mu_{\sigma, \nu, c}), \quad (\mathbb{R} \setminus \text{supp}(\mu_{\sigma, \nu, c}) = \Phi_{\sigma, \nu, c}(\mathcal{E}_{\sigma, \nu, c})).$$

$$\mathcal{S} := \text{supp}(\mu_{\sigma, \nu, c}) \cup \{\rho_\theta, \theta \in \Theta_{\sigma, \nu, c}\}.$$

Theorem (C.2013)

- If $c = 1$,
 $\mathbb{P}[\text{for all large } N, \text{spect}(M_N) \subset \{x, \text{dist}(x, \mathcal{S}) \leq \epsilon\}] = 1.$
- If $c < 1$, then, $\forall \epsilon > 0$,
 $\mathbb{P}[\text{for all large } N, \text{spect}(M_N) \subset \{x, \text{dist}(x, \mathcal{S} \cup \{0\}) \leq \epsilon\}] = 1.$
- If $c < 1$ and if moreover $\theta_j > 0$ and $0 \in \mathcal{E}_{\sigma, \nu, c}$, then, $\forall \epsilon > 0$,
 $\mathbb{P}[\text{for all large } N, \text{spect}(M_N) \subset \{x, \text{dist}(x, \mathcal{S}) \leq \epsilon\}] = 1.$

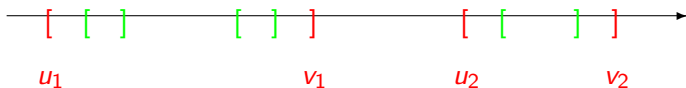
$n_{i-1} + 1, \dots, n_{i-1} + k_i$: descending ranks of θ_i among the eigenvalues $A_N A_N^*$.

Complete description of the convergence of

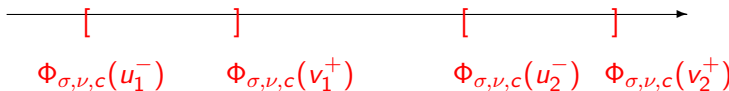
$\lambda_{n_{i-1}+1}(M_N), \dots, \lambda_{n_{i-1}+k_i}(M_N)$

depending on the location of θ_i with respect to $\mathcal{E}_{\sigma, \nu, c}$ and the connected components of support ν .

${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν



support $\mu_{\sigma,\nu,c}$

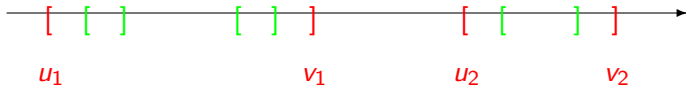


$$\Phi_{\sigma,\nu,c} : \begin{cases} \mathbb{R} \setminus \text{supp}(\nu) \rightarrow \mathbb{R} \\ x \mapsto x(1 + c\sigma^2 g_\nu(x))^2 + \sigma^2(1 - c)(1 + c\sigma^2 g_\nu(x)) \end{cases} .$$

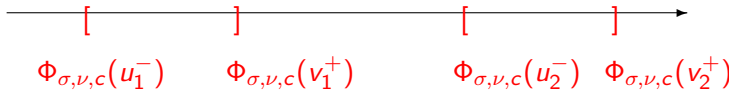
$$\mathcal{E}_{\sigma,\nu,c} := \left\{ u \in \mathbb{R} \setminus \text{supp}(\nu), \Phi'_{\sigma,\nu,c}(u) > 0, g_\nu(u) > -\frac{1}{\sigma^2 c} \right\} .$$

Convergence of outliers

${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν

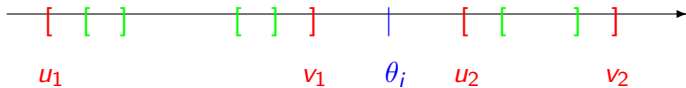


support $\mu_{\sigma,\nu,c}$

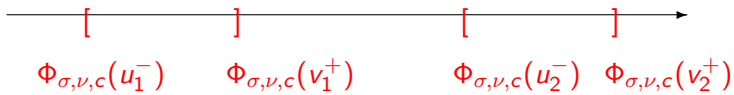


Convergence of outliers

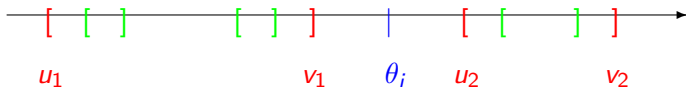
${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν



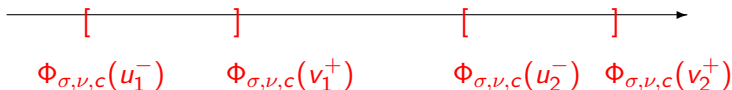
support $\mu_{\sigma,\nu,c}$



${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν



support $\mu_{\sigma,\nu,c}$



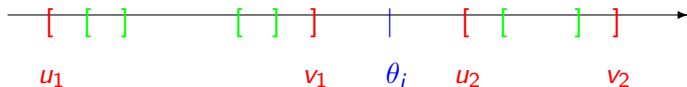
Theorem (C.2013)

$n_{i-1} + 1, \dots, n_{i-1} + k_i$: descending ranks of θ_i among the eigenvalues of $A_N A_N^*$.

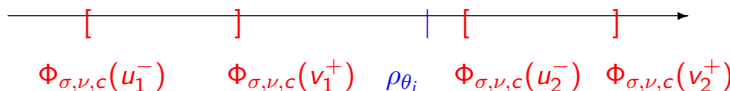
If $\theta_i \in \mathcal{E}_{\sigma,\nu,c}$ and if $c < 1$ $\theta_i \neq 0$, the k_i eigenvalues $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s outside the support of $\mu_{\sigma,\nu,c}$ towards

$$\rho_{\theta_i} = \Phi_{\sigma,\nu,c}(\theta_i) = \theta_i(1 + c\sigma^2 g_\nu(\theta_i))^2 + \sigma^2(1 - c)(1 + c\sigma^2 g_\nu(\theta_i)).$$

${}^c\mathcal{E}_{\sigma, \nu, c}$ support ν



support $\mu_{\sigma, \nu, c}$



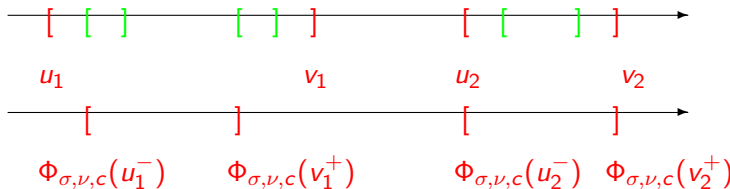
Theorem (C.2013)

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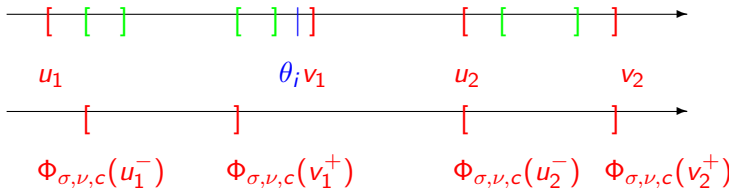
If $\theta_i \in \mathcal{E}_{\sigma, \nu, c}$ and if $c < 1$ $\theta_i \neq 0$, the k_i eigenvalues $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s outside the support of $\mu_{\sigma, \nu, c}$ towards

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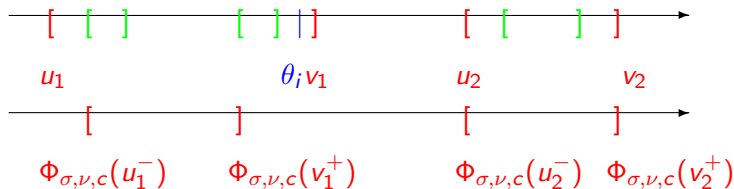
Convergence of outliers



Convergence of outliers



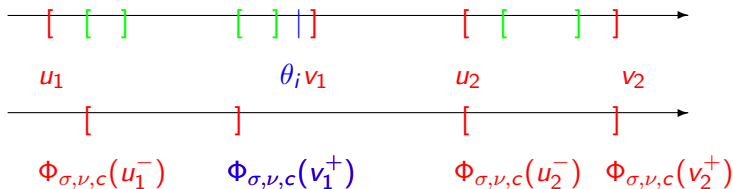
Convergence of outliers



Theorem (C. 2013)

${}^c\mathcal{E}_{\sigma, \nu, c} = \bigcup_{l=1}^p [u_l, v_l]$. Assume $\theta_i \in {}^c\mathcal{E}_{\sigma, \nu, c}$, $\theta_i \in [u_l, v_l]$ and θ_i is on the right hand side of any connected component of $\text{supp}(\nu)$ which is included in $[u_l, v_l]$. Let $n_{i-1} + 1, \dots, n_{i-1} + k_i$ be the descending ranks of θ_i among the eigenvalues of $A_N A_N^*$. Then $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s towards $\Phi_{\sigma, \nu, c}(v_l)$ which is a boundary point of the support of $\mu_{\sigma, \nu, c}$.

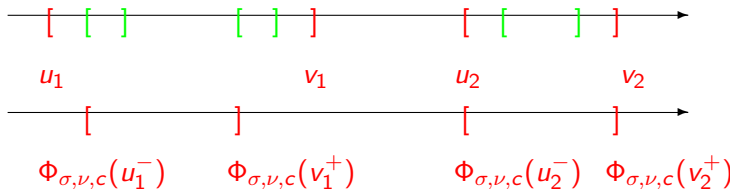
Convergence of outliers



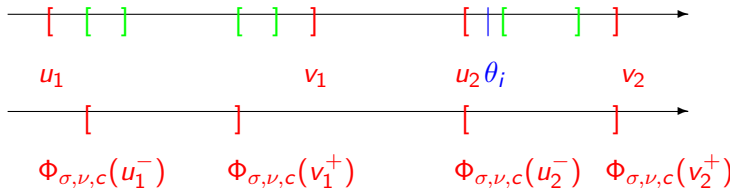
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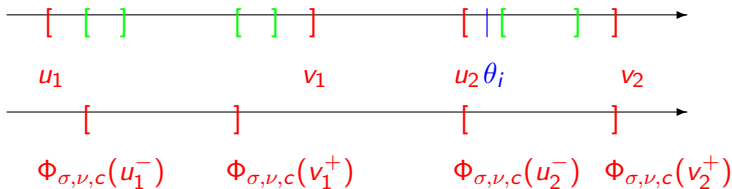
Convergence of outliers



Convergence of outliers



Convergence of outliers

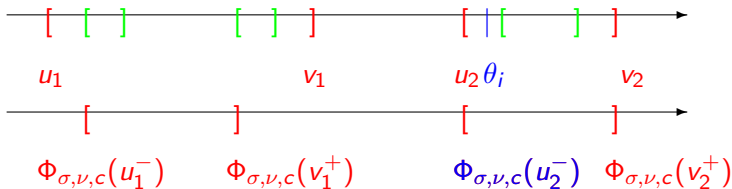


Theorem (C. 2013)

${}^c\mathcal{E}_{\sigma, \nu, c} = \bigcup_{l=1}^p [u_l, v_l]$. Assume that $\theta_i \in {}^c\mathcal{E}_{\sigma, \nu, c}$, $\theta_i \in [u_{l_i}, v_{l_i}]$ and θ_i is on the left hand side of any connected component of $\text{supp}(\nu)$ which is included in $[u_{l_i}, v_{l_i}]$. Let $n_{i-1} + 1, \dots, n_{i-1} + k_i$ be the descending ranks of θ_i among the eigenvalues of $A_N A_N^*$.

- If $u_{l_i} > 0$ then $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s towards $\Phi_{\sigma, \nu, c}(u_{l_i}^-)$
- If $l_i = 1$ and $\Phi_{\sigma, \nu, c}(u_1^-) = 0$ then $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s towards 0.

Convergence of outliers

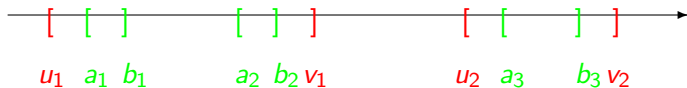


Theorem (C. 2013)

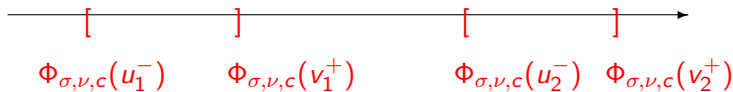
${}^c\mathcal{E}_{\sigma, \nu, c} = \bigcup_{l=1}^p [u_l, v_l]$. Assume that $\theta_i \in {}^c\mathcal{E}_{\sigma, \nu, c}$, $\theta_i \in [u_{l_i}, v_{l_i}]$ and θ_i is on the left hand side of any connected component of $\text{supp}(\nu)$ which is included in $[u_{l_i}, v_{l_i}]$. Let $n_{i-1} + 1, \dots, n_{i-1} + k_i$ be the descending ranks of θ_i among the eigenvalues of $A_N A_N^*$.

- If $u_{l_i} > 0$ then $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s towards $\Phi_{\sigma, \nu, c}(u_{l_i}^-)$
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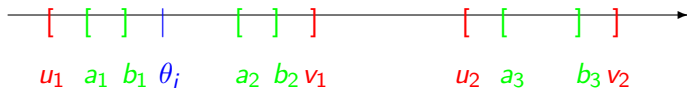
$\mathcal{E}_{\sigma, \nu, c}$ support ν



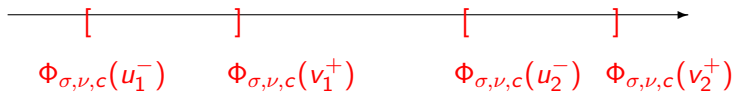
support $\mu_{\sigma, \nu, c}$



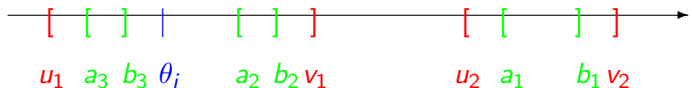
$\mathcal{E}_{\sigma, \nu, c}$ support ν



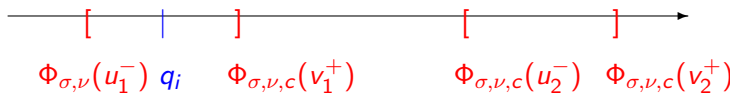
support $\mu_{\sigma, \nu, c}$



${}^c\mathcal{E}_{\sigma,\nu,c}$ support ν



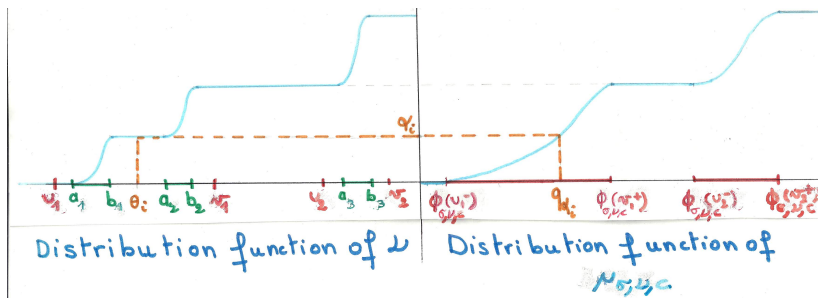
support $\mu_{\sigma,\nu,c}$



Theorem (C. 2013)

If θ_i is between two connected components of $\text{supp}(\nu)$ which are included in $[u_i; v_i]$ then the k_i eigenvalues $(\lambda_{n_{i-1}+j}(M_N), 1 \leq j \leq k_i)$ converge a.s towards q_i defined by $\mu_{\sigma,\nu,c}([-\infty, q_i]) = \nu([-\infty, \theta_i])$.

Convergence of outliers



continuous and
strictly increasing on
 $] \Phi_{\sigma, \nu, c}(u_l^-); \Phi_{\sigma, \nu, c}(u_l^+)[$

Rectangular free convolution

- (a) A and B are independent $n \times p$ matrices,
- (b) A or B is invariant, in law, under multiplication, on the right and on the left, by any unitary matrix,
- (c) there exists μ_1 and μ_2 , two laws on \mathbb{R}^+ such that, for the weak convergence in probability, we have

$$\frac{1}{n} \sum_{s \text{ sing. val. of } A} \delta_s \rightarrow \mu_1, \quad \frac{1}{n} \sum_{s \text{ sing. val. of } B} \delta_s \rightarrow \mu_2, \text{ as } n, p \text{ go to infinity with } n/p \rightarrow c.$$

Theorem (Benaych-Georges)

Under Hypotheses (a), (b) and (c) above, there is a non random law μ on \mathbb{R}^+ , depending only on μ_1 , μ_2 and c , such that for the weak convergence in probability, we have

$$\frac{1}{n} \sum_{s \text{ sing. val. of } A+B} \delta_s \rightarrow \mu. \text{ The law } \mu \text{ denoted by } \mu_1 \boxplus_c \mu_2 \text{ is}$$

*called **the rectangular free convolution with ratio c of the laws μ_1 and μ_2 .***

Choosing Gaussian entries for X_N , we can deduce that

$$\sqrt{\mu_{\sigma, \nu, c}} = \sqrt{\nu} \boxplus_c \sqrt{\sigma^2 \mu_c}$$

where μ_c is the Marchenko-Pastur distribution:

$$\frac{d\mu_c}{dx}(x) = \frac{1}{2\pi cx} \sqrt{(b-x)(x-a)} 1_{[a,b]}(x)$$

$a = (1 - \sqrt{c})^2$, $b = (1 + \sqrt{c})^2$, and for any probability measure on \mathbb{R}^+ and any $\alpha > 0$, $\sqrt{\alpha\tau}$ denotes the pushforward of τ by the map $x \mapsto \sqrt{\alpha x}$.

(F. Benaych-Georges): rectangular R-transform with ratio c

τ probability measure on \mathbb{R}^+ ; $c \in]0; 1]$.

$$M_\tau(z) = \int_{\mathbb{R}^+} \frac{t^2 z}{1 - t^2 z} d\tau(t).$$

$$H_\tau^{(c)}(z) := z(cM_\tau(z) + 1)(M_\tau(z) + 1).$$

$$C_\tau^{(c)}(z) = T^{(c)-1} \left(\frac{z}{H_\tau^{(c)-1}(z)} \right), \quad \text{for } z \neq 0;$$

$$C_\tau^{(c)}(0) = 0.$$

$$T^{(c)}(z) = (cz + 1)(z + 1).$$

Rectangular subordination

Theorem (Belinschi-Benaych-Georges-Guionnet)

Assume that the rectangular R -transform $C_\tau^{(c)}$ of τ extends analytically to $\mathbb{C} \setminus \mathbb{R}^+$; this happens for example if τ is \boxplus_c infinitely divisible. Then there exist two unique meromorphic functions ω_1, ω_2 on $\mathbb{C} \setminus \mathbb{R}^+$ so that

$$H_\tau^{(c)}(\omega_1(z)) = H_\nu^{(c)}(\omega_2(z)) = H_{\tau \boxplus_c \nu}^{(c)}(z),$$

$\omega_j(\bar{z}) = \overline{\omega_j(z)}$ and $\lim_{x \uparrow 0} \omega_j(x) = 0$, $j \in \{1; 2\}$.

The equation satisfied by the Stieltjes transform of $\mu_{\sigma,\nu,c}$ leads to

$$\begin{aligned} czg_{\mu_{\sigma,\nu,c}}(z)^2 + (1-c)g_{\mu_{\sigma,\nu,c}}(z) \\ = c\omega_{\sigma,\nu,c}(z)g_{\nu}(\omega_{\sigma,\nu,c}(z))^2 + (1-c)g_{\nu}(\omega_{\sigma,\nu,c}(z)). \end{aligned}$$

The equation satisfied by the Stieltjes transform of $\mu_{\sigma,\nu,c}$ leads to

$$\begin{aligned} czg_{\mu_{\sigma,\nu,c}}(z)^2 + (1-c)g_{\mu_{\sigma,\nu,c}}(z) \\ = c\omega_{\sigma,\nu,c}(z)g_{\nu}(\omega_{\sigma,\nu,c}(z))^2 + (1-c)g_{\nu}(\omega_{\sigma,\nu,c}(z)). \end{aligned}$$

Since $H_{\sqrt{\tau}}^{(c)}(z) = \frac{c}{z}g_{\tau}(\frac{1}{z})^2 + (1-c)g_{\tau}(\frac{1}{z})$, this may be rewritten

$$H_{\sqrt{\mu_{\sigma,\nu,c}}}^{(c)}\left(\frac{1}{z}\right) = H_{\sqrt{\nu}}^{(c)}\left(\frac{1}{\omega_{\sigma,\nu,c}(z)}\right) \quad \text{and then} \quad \omega_2(z) = \frac{1}{\omega_{\sigma,\nu,c}(1/z)}.$$

The equation satisfied by the Stieltjes transform of $\mu_{\sigma,\nu,c}$ leads to

$$\begin{aligned} czg_{\mu_{\sigma,\nu,c}}(z)^2 + (1-c)g_{\mu_{\sigma,\nu,c}}(z) \\ = c\omega_{\sigma,\nu,c}(z)g_{\nu}(\omega_{\sigma,\nu,c}(z))^2 + (1-c)g_{\nu}(\omega_{\sigma,\nu,c}(z)). \end{aligned}$$

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Results in the lineage of various works on the convergence of outliers of "spiked Hermitian models":

Sample covariance matrices:

Baik-BenArous-Péché (2005) (gaussian case, finite rank perturbation)

Bai-Yao (2008), Rao-Silverstein (2010).

Deformed Wigner matrices:

Péché (2006) (gaussian case, finite rank perturbation),

Féral-Péché (2007) (perturbation of rank 1) ,

Capitaine-Donati-Martin-Féral (2009) (perturbation of finite rank),

Capitaine-Donati-Martin-Féral-Février (2011)

Pizzo, Renfrew, Soshnikov (2011-2012) (perturbation of finite rank)

Deformed unitarily invariant models: Benaych-Georges-Rao (2011) (perturbation of finite rank),

Belinschi-Bercovici-Capitaine-Février (2012).

Information-plus-noise:

Loubaton-Vallet (2010) (gaussian case , perturbation of finite rank)

As to deformations of non Hermitian models:

Bounded rank perturbation of i.i.d matrices Tao (2010)

Low rank perturbations of large elliptic random matrices O'Rourke-Renfrew (2013)

Bounded rank perturbation of a bi-unitarily invariant model
Benaych-Georges-Rochet (2013)

Perturbation of i.i.d matrices forthcoming paper with Charles Bordenave