

# Spectrum of large deformed classical Hermitian matrices and free probability theory

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# Aim of this talk:

To show how free probability theory sheds light on spectral properties of deformed matricial models and provides **A UNIFIED UNDERSTANDING** of various phenomena

# Notations

$$B = B^* \in \mathcal{M}_N(\mathbb{C})$$

Eigenvalues of  $B$ :  $\lambda_1(B) \geq \lambda_2(B) \geq \dots \geq \lambda_N(B)$ ,

The empirical distribution of these eigenvalues:

$$\mu_B := \frac{1}{N} \sum_{i=1}^N \delta_{\lambda_i(B)}$$

$\mu$  probability measure on  $\mathbb{C}$ ,  $z \in \mathbb{C} \setminus \text{supp}(\mu)$ ,  $g_\mu(z) = \int \frac{d\mu(x)}{z - x}$

$\mathcal{M}$ : the set of probability measures supported on the real line

$\mathcal{M}^+$ : the set of probability measures supported on  $[0; +\infty[$ .

Free probability theory defines:

- a binary operation on  $\mathcal{M}$ : **the free additive convolution**  $\mu \boxplus \nu$  for  $\mu$  and  $\nu$  in  $\mathcal{M}$ ,
- binary operations on  $\mathcal{M}^+$ : **the free multiplicative convolution**  $\mu \boxtimes \nu$  and **the free rectangular convolution with ratio  $c \in ]0; 1]$**   $\mu \boxplus_c \nu$ , for  $\mu$  and  $\nu$  in  $\mathcal{M}^+$ ,

(cf Voiculescu, Maassen, Bercovici-Voiculescu, and Benaych-Georges)

For several matricial models where  $A_N$  and  $B_N$  are independent  $N \times N$  Hermitian random matrices

(for instance when  $\mathcal{L}(B_N) = \mathcal{L}(U_N B_N U_N^*)$  for any deterministic unitary matrix (“unitarily invariant”)),

free probability provides a good understanding of the asymptotic global behaviour of the spectrum of  $A_N + B_N$  and  $A_N^{\frac{1}{2}} B_N A_N^{\frac{1}{2}}$  ( $A_N \geq 0, B_N \geq 0$ )

$$\mu_{A_N+B_N} \xrightarrow{N \rightarrow +\infty} \mu_a \boxplus \mu_b$$

$$\mu_{A_N^{\frac{1}{2}} B_N A_N^{\frac{1}{2}}} \xrightarrow{N \rightarrow +\infty} \mu_a \boxtimes \mu_b$$

where  $\mu_{A_N} \xrightarrow{N \rightarrow +\infty} \mu_a$  and  $\mu_{B_N} \xrightarrow{N \rightarrow +\infty} \mu_b$ .

Pionnering work 90' of D. Voiculescu extended by several authors

For several matricial models where  $A_N$  and  $B_N$  are independent rectangular  $n \times N$  random matrices such that  $n/N \rightarrow c \in ]0; 1]$ , rectangular free convolution provides a good understanding of the asymptotic global behaviour of the singular values of  $A_N + B_N$  :

$$\frac{1}{n} \sum_{s \text{ sing. val. of } A_N + B_N} \delta_s \rightarrow \nu_a \boxplus_c \nu_b.$$

$$\left( \text{where } \frac{1}{n} \sum_{s \text{ sing. val. of } A_N} \delta_s \rightarrow \nu_a, \quad \frac{1}{n} \sum_{s \text{ sing. val. of } B_N} \delta_s \rightarrow \nu_b \right)$$

(cf work of Benaych-Georges when  $B_N$  is invariant, in law, under multiplication, on the right and on the left, by any unitary matrix: “biunitarily invariant”)

# Additive free subordination property

For a probability measure  $\tau$  on  $\mathbb{R}$ ,  $z \in \mathbb{C} \setminus \mathbb{R}$ ,  $g_\tau(z) = \int_{\mathbb{R}} \frac{d\tau(x)}{z-x}$ .

Theorem (D.Voiculescu (93), P. Biane (98))

Let  $\mu$  and  $\nu$  be two probability measures on  $\mathbb{R}$ , there exists a unique analytic map  $\omega_{\mu,\nu} : \mathbb{C}^+ \rightarrow \mathbb{C}^+$  such that

$$\forall z \in \mathbb{C}^+, g_{\mu \boxplus \nu}(z) = g_\nu(\omega_{\mu,\nu}(z)),$$

$\forall z \in \mathbb{C}^+, \Im \omega_{\mu,\nu}(z) \geq \Im z$  and  $\lim_{y \uparrow +\infty} \frac{\omega_{\mu,\nu}(iy)}{iy} = 1$ .

$\omega_{\mu,\nu}$  is called the **additive subordination map** of  $\mu \boxplus \nu$  with respect to  $\nu$ .

## Multiplicative free subordination property

$$\Psi_{\tau}(z) = \int \frac{tz}{1-tz} d\tau(t) = \frac{1}{z} g_{\tau}\left(\frac{1}{z}\right) - 1,$$

for complex values of  $z$  such that  $\frac{1}{z}$  is not in the support of  $\tau$ .

### Theorem (Biane (98))

Let  $\tau \neq \delta_0$  and  $\nu \neq \delta_0$  be two probability measures on  $[0; +\infty[$ . There exists a unique analytic map  $F_{\tau, \nu}$  defined on  $\mathbb{C} \setminus [0; +\infty[$  such that

$$\forall z \in \mathbb{C} \setminus [0; +\infty[, \Psi_{\nu \boxtimes \tau}(z) = \Psi_{\nu}(F_{\tau, \nu}(z))$$

and

$$\forall z \in \mathbb{C}^+, F_{\tau, \nu}(z) \in \mathbb{C}^+, F_{\tau, \nu}(\bar{z}) = \overline{F_{\tau, \nu}(z)}, \arg(F_{\tau, \nu}(z)) \geq \arg(z).$$

$F_{\tau, \nu}$  is called the **multiplicative subordination map** of  $\tau \boxtimes \nu$  with respect to  $\nu$ .

# Rectangular free subordination property

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

$$M_\tau(z) = \int_{\mathbb{R}^+} \frac{t^2 z}{1 - t^2 z} d\tau(t), \quad H_\tau^{(c)}(z) := z(cM_\tau(z) + 1)(M_\tau(z) + 1).$$

**Theorem (Belinschi&Benaych-Georges&Guionnet (2008))**

*Assume that  $\tau$  is  $\boxplus_c$  infinitely divisible. Then there exist two unique meromorphic functions  $\omega_1, \omega_2$  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  $\lim_{x \uparrow 0} \omega_j(x) = 0$ ,  $j \in \{1; 2\}$ .

# Standard models

- **Wigner matrices**

$$X_N = \frac{1}{\sqrt{N}} W_N$$

$(W_N)_{ii}$ ,  $\sqrt{2}\Re e((W_N)_{ij})_{i<j}$ ,  $\sqrt{2}\Im m((W_N)_{ij})_{i<j}$  are i.i.d, with distribution  $\mu$  with variance  $\sigma^2$  and mean zero.

If  $\mu = \mathcal{N}(0, \sigma^2)$ ,  $W_N =: W_N^G$  is a *G.U.E*-matrix.

- **Wishart matrices**

$$X_N = \frac{1}{p} B_N B_N^*$$

$B_N$  is a  $N \times p(N)$  matrix,  $(B_N)_{u,v} = Z_{u,v} + iY_{u,v}$ ,  $Z_{u,v}$ ,  $Y_{u,v}$ ,  $u = 1, \dots, N$ ,  $v = 1, \dots, p(N)$  are i.i.d, with distribution  $\mu$  with variance  $\frac{1}{2}$  and mean zero.

If  $\mu = \mathcal{N}(0, \frac{1}{2})$ ,  $X_N$  is a *L.U.E* matrix.

# Convergence of the spectral measure

## Theorem (Wigner (50'))

$$\mu_{\frac{W_N}{\sqrt{N}}} := \frac{1}{N} \sum_{i=1}^N \delta_{\lambda_i(\frac{W_N}{\sqrt{N}})} \rightarrow \mu_{sc} \text{ a.s. when } N \rightarrow +\infty$$

$$\frac{d\mu_{sc}}{dx}(x) = \frac{1}{2\pi\sigma^2} \sqrt{4\sigma^2 - x^2} \mathbf{1}_{[-2\sigma, 2\sigma]}(x)$$

## Theorem (Marchenko-Pastur (1967))

If  $c_N := \frac{N}{p} \rightarrow c > 0$  when  $N \rightarrow \infty$ ,

$$\mu_{\frac{B_N B_N^*}{p}} \rightarrow \mu_{MP} \text{ a.s. when } N \rightarrow +\infty$$

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

$a = (1 - \sqrt{c})^2$ ,  $b = (1 + \sqrt{c})^2$ . and  $\mu_c(0) = 1 - \frac{1}{c}$  if  $c > 1$ .

## No outlier

## Theorem (Bai-Yin 1988)

If  $\int x^4 d\mu(x) < +\infty$ , then

$$\lambda_1\left(\frac{W_N}{\sqrt{N}}\right) \rightarrow 2\sigma \text{ and } \lambda_N\left(\frac{W_N}{\sqrt{N}}\right) \rightarrow -2\sigma \text{ a.s when } N \rightarrow +\infty.$$

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

If  $\int x^4 d\mu(x) < +\infty$ ,

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

$$\lambda_{\min(N,p)}\left(\frac{B_N B_N^*}{p}\right) \rightarrow (1 - \sqrt{c})^2 \text{ a.s when } N \rightarrow +\infty.$$

## Deformed models

$A_N$  is a deterministic matrix such that  $\sup_N \|A_N\| < \infty$ .

- **Deformed Wigner matrices**  $W_N$  is a Wigner matrix and  $A_N$  is an Hermitian matrix such that  $\mu_{A_N} \rightarrow_{N \rightarrow +\infty} \nu$  weakly.

$$M_N = \frac{W_N}{\sqrt{N}} + A_N$$

- **Sample covariance matrices**  $A_N$  is a non negative definite matrix such that  $\mu_{A_N} \rightarrow_{N \rightarrow +\infty} \nu$  weakly.

$$M_N = A_N^{\frac{1}{2}} \frac{B_N B_N^*}{p} A_N^{\frac{1}{2}}.$$

- **Information-Plus-Noise type matrices**,  $N \leq p(N)$ ,  $A_N$  is such that  $\mu_{A_N A_N^*} \rightarrow_{N \rightarrow +\infty} \nu$  weakly.

$$M_N = \left( \frac{B_N}{\sqrt{p}} + A_N \right) \left( \frac{B_N}{\sqrt{p}} + A_N \right)^*.$$

# Convergence of spectral measures

- **Deformed Wigner matrices**  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{dW}$  weakly.  
Pastur (72), Anderson&Guionnet&Zeitouni (2010)
- **Sample covariance matrices**  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{Scm}$  weakly.  
Marchenko&Pastur (67) Grenander&Silverstein(77), Wachter (78), Krishnaiah&Y.Q.Yin (83), Y.Q.Yin (86), Bai&Silverstein (95), Silverstein (95).
- **Information-Plus-Noise type matrices**  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{Ipn}$  weakly.  
Dozier&Silverstein (2007), Hachem&Loubaton&Najim (2007), Xie (2012)

# Convergence of spectral measures

- Deformed Wigner matrices  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{dW}$  weakly.

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{dW}}(z) = \int \frac{1}{z - \sigma^2 g_{\mu_{dW}}(z) - t} d\nu(t).$$

- Sample covariance matrices  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{Scm}$  weakly.

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{Scm}}(z) = \int \frac{1}{z - t(1 - c + czg_{\mu_{Scm}}(z))} d\nu(t).$$

- Information-Plus-Noise type matrices  $\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu_{Ipn}$  weakly.

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{Ipn}}(z) = \int \frac{1}{(1 - cg_{\mu_{Ipn}}(z))z - \frac{t}{1 - cg_{\mu_{Ipn}}(z)} - (1 - c)} d\nu(t).$$

$\mu_{dW}, \mu_{Scm}, \mu_{Ipn}$  are **deterministic**, in general non explicit. They are **universal** (do not depend on the distribution of the entries of  $W_N$  or  $B_N$ ) and only depend on  $A_N$  through the limiting spectral measure  $\nu$ .

# Free probabilistic interpretation

- Deformed Wigner matrices

$$\mu_{M_N} \xrightarrow{N \rightarrow +\infty} \mu_{dW} \text{ weakly, } \mu_{dW} = \mu_{sc} \boxplus \nu$$

- Sample covariance matrices

$$\mu_{M_N} \xrightarrow{N \rightarrow +\infty} \mu_{Scm} \text{ weakly, } \mu_{Scm} = \mu_{MP} \boxtimes \nu$$

- Information-Plus-Noise type matrices

$$\mu_{M_N} \xrightarrow{N \rightarrow +\infty} \mu_{Ipn} \text{ weakly, } \mu_{Ipn} = (\sqrt{\mu_{MP}} \boxplus_c \sqrt{\nu})^2.$$

# Equations satisfied by the limiting Stieltjes transforms

## ⇔ Free Subordination properties

- Deformed Wigner matrices

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{sc} \boxplus \nu}(z) = \int \frac{1}{z - \sigma^2 g_{\mu_{sc} \boxplus \nu}(z) - t} d\nu(t) = g_{\nu}(\omega_{\mu_{sc}, \nu}(z)).$$

$$\omega_{\mu_{sc}, \nu}(z) = z - \sigma^2 g_{\mu_{sc} \boxplus \nu}(z).$$

- Sample covariance matrices

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{MP} \boxtimes \nu}(z) = \int \frac{1}{z - t(1 - c + czg_{\mu_{MP} \boxtimes \nu}(z))} d\nu(t).$$

$$\rightarrow \quad \Psi_{\mu_{MP} \boxtimes \nu} \left( \frac{1}{z} \right) = \Psi_{\nu} \left( F_{\mu_{MP}, \nu} \left( \frac{1}{z} \right) \right)$$

$$\Psi_{\tau}(z) = \int \frac{tz}{1 - tz} d\tau(t) = \frac{1}{z} g_{\tau} \left( \frac{1}{z} \right) - 1,$$

$$F_{\mu_{MP}, \nu}(z) = z - cz + cg_{\mu_{MP} \boxtimes \nu} \left( \frac{1}{z} \right).$$

# Equations satisfied by the limiting Stieltjes transforms

## $\iff$ Free Subordination properties

- Information-Plus-Noise type matrices

$$\mu_{Ipn} = (\sqrt{\mu_{MP}} \boxtimes_c \sqrt{\nu})^2$$

$$\forall z \in \mathbb{C}^+, \quad g_{\mu_{Ipn}}(z) = \int \frac{1}{(1 - cg_{\mu_{Ipn}}(z))z - \frac{t}{1 - cg_{\mu_{Ipn}}(z)} - (1 - c)} d\nu(t).$$

$$\rightarrow H_{\sqrt{\mu_{Ipn}}}^{(c)}\left(\frac{1}{z}\right) = H_{\sqrt{\nu}}^{(c)}\left(\Omega_{\mu_{MP}, \nu}\left(\frac{1}{z}\right)\right)$$

$$H_{\sqrt{\tau}}^{(c)}(z) = \frac{c}{z} g_{\tau}\left(\frac{1}{z}\right)^2 + (1 - c)g_{\tau}\left(\frac{1}{z}\right),$$

$$\Omega_{\mu_{MP}, \nu}(z) = \frac{1}{\frac{1}{z}(1 - cg_{\mu_{Ipn}}\left(\frac{1}{z}\right))^2 - (1 - c)(1 - cg_{\mu_{Ipn}}\left(\frac{1}{z}\right))}$$

# Deep Studies of the limiting spectral measures

Support, density, behaviour of the density near its zeroes....

- **Deformed Wigner matrices**  $\mu_{dW} = \mu_{sc} \boxplus \nu$   
P. Biane (1997)
- **Sample covariance matrices**  $\mu_{Scm} = \mu_{MP} \boxtimes \nu$   
Choi&Silverstein (1995)
- **Information-Plus-Noise type matrices**  $\mu_{Ipn} = (\sqrt{\mu_{MP}} \boxplus_c \sqrt{\nu})^2$   
Dozier&Silverstein (2007)

# Characterization of the complement of the supports

(P.Biane 1997):

$$\mathcal{O} := \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) < \frac{1}{\sigma^2} \right\}$$

$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu = h_{\mu_{sc}, \nu}(\mathcal{O}).$$

$$h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_{\nu}(z).$$

$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu \begin{array}{c} \xrightarrow{\omega_{\mu_{sc}, \nu}} \\ \xleftarrow{h_{\mu_{sc}, \nu}} \end{array} \mathcal{O},$$

The additive subordination map  $\omega_{\mu_{sc}, \nu}(z) = z - \sigma^2 g_{\mu_{sc} \boxplus \nu}(z)$

$h_{\mu_{sc}, \nu}$  globally strictly increasing on  $\mathcal{O}$ .

- Deformed Wigner

$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu \begin{array}{c} \xrightarrow{\omega_{\mu_{sc}\nu}} \\ \longleftarrow \\ \phi_1 \end{array} \mathcal{O}_1 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_1, \quad \phi_1(\theta) = \theta + \sigma^2 g_\nu(\theta).$$

- Sample covariance matrices

$$\mathbb{R} \setminus \{\text{support } \mu_{MP} \boxtimes \nu \cup \{0\}\} \begin{array}{c} x \mapsto \frac{1}{F_{\mu_{MP}, \nu}(1/x)} \\ \xrightarrow{\quad} \\ \longleftarrow \\ \phi_2 \end{array} \mathcal{O}_2 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_2 \quad \phi_2(\theta) = \theta + c\theta \int \frac{t}{\theta-t} d\nu(t).$$

- Information-Plus-Noise type model

$$\mathbb{R} \setminus (\sqrt{\mu_{MP}} \boxtimes_c \sqrt{\nu})^2 \begin{array}{c} x \mapsto \frac{1}{\Omega_{\mu_{MP}, \nu}(1/x)} \\ \xrightarrow{\quad} \\ \longleftarrow \\ \phi_3 \end{array} \mathcal{O}_3 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_3, \quad \phi_3(\theta) = \theta(1 + c g_\nu(\theta))^2 + (1 - c)(1 + c g_\nu(\theta))$$







## Theorem (Péché 2006)

- If  $\theta_1 \leq \sigma$ ,  $\lambda_1(M_N) \rightarrow 2\sigma$
- If  $\theta_1 > \sigma$ ,  $\lambda_1(M_N) \rightarrow \rho_{\theta_1}$  with  $\rho_{\theta_1} := \theta_1 + \frac{\sigma^2}{\theta_1}$ .



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- If  $\theta_1 \leq \sigma$ ,  $\lambda_1(M_N) \rightarrow 2\sigma$
- If  $\theta_1 > \sigma$ ,  $\lambda_1(M_N) \rightarrow \rho_{\theta_1}$  with  $\rho_{\theta_1} := \theta_1 + \frac{\sigma^2}{\theta_1}$ .



Actually if for some  $i$ ,  $|\theta_i| > \sigma$  then exactly  $k_i$  eigenvalues of  $M_N$  converge towards  $\rho_{\theta_i} := \theta_i + \frac{\sigma^2}{\theta_i} \in ]-\infty; -2\sigma[ \cup ]2\sigma; +\infty[$ .

When  $A_N$  has finite rank, analog **B.B.P phase transition phenomena** proved for

$A_N + B_N$	$(I_N + A_N)^{1/2} B_N (I_N + A_N)^{1/2}$ $I_N + A_N > 0$	$(A_N + B_N)(A_N + B_N)^*$
$B_N = GUE$ Péché (2006)  $B_N = \text{Wigner}$ Féral&Péché (2007) C.&Donati-Martin&Féral (2009) Pizzo&Renfrew&Soshnikov (2013), Knowles&Yin (2014)	$B_N = \text{L.U.E}$ Baik&Ben Arous&Péché (2005)  $B_N = \text{Wishart}$ Baik&Silverstein (2006)	$B_N$ Ginibre matrix Loubaton&Vallet (2011)

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$B_N$ unitarily invariant $\mu_{B_N} \rightarrow_{N \rightarrow +\infty} \mu$ without outlier Benaych-Georges&Rao(2010)	$B_N$ unitarily invariant $\mu_{B_N} \rightarrow_{N \rightarrow +\infty} \mu$ $B_N \geq 0$ without outlier Benaych-Georges&Rao(2010)	$B_N$ biunitarily invariant $\mu_{B_N B_N^*} \rightarrow_{N \rightarrow +\infty} \mu$ without outlier Benaych-Georges&Rao(2010)

Free subordination properties shed light on these phenomena and provide a UNIFIED UNDERSTANDING, allowing to extend them to non-finite rank deformations.

$A_N$  Hermitian deterministic.  $\mu_{A_N} \rightarrow_{N \rightarrow +\infty} \nu$  compactly supported.  
The eigenvalues of  $A_N$ :

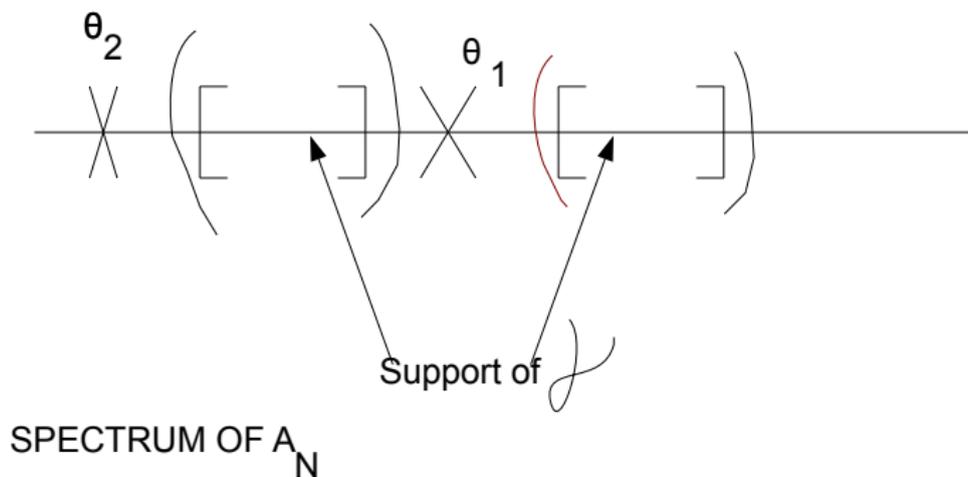
- $N - r$  ( $r$  fixed) eigenvalues  $\beta_i(N)$  such that

$$\max_{i=1}^{N-r} \text{dist}(\beta_i(N), \text{supp}(\nu)) \rightarrow_{N \rightarrow \infty} 0$$

- a **finite** number  $J$  of **fixed** (independent of  $N$ ) eigenvalues **(SPIKES)**  $\theta_1 > \dots > \theta_J, \forall i = 1, \dots, J, \theta_i \notin \text{supp}(\nu)$ , each  $\theta_j$  having a fixed multiplicity  $k_j, \sum_j k_j = r$ .

For large  $N$ , the  $\beta_i(N)$  are inside

an  $\varepsilon$  neighborhood of the support of  $\mathcal{J}$



# Naive intuition for general additive deformed models:

$$\mathfrak{g}_{\mu \boxplus \nu}(z) = \mathfrak{g}_{\nu}(\omega_{\mu, \nu}(z))$$

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$$g_{\mu_{M_N}}(z) \approx g_{\mu_{A_N}}(\omega_{\mu, \nu}(z))$$

If  $\rho \notin \text{support } \mu \boxplus \nu$  is a solution of  $\omega_{\mu, \nu}(\rho) = \theta_i$  for some  $i \in \{1, \dots, J\}$ ,

$\rho \notin \text{support } \mu \boxplus \nu$  BUT  $g_{\mu_{M_N}}(\rho) \approx g_{\mu_{A_N}}(\omega_{\mu, \nu}(\rho))$  explodes!

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Conjecture:

$\implies$  The spikes  $\theta_i$ 's of the perturbation  $A_N$  that may generate outliers in the spectrum of  $M_N$  belong to  $\omega_{\mu, \nu}(\mathbb{R} \setminus \text{support } \mu \boxplus \nu)$

# Naive intuition for general additive deformed models:

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$$g_{\mu_{M_N}}(z) \approx g_{\mu_{A_N}}(\omega_{\mu, \nu}(z))$$

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$\implies$  for large  $N$ , the  $\theta_i$ 's such that the equation

$$\omega_{\mu, \nu}(\rho) = \theta_i$$

has solutions  $\rho$  outside support  $\mu \boxplus \nu$  generate eigenvalues of  $M_N$  in a neighborhood of each of these  $\rho$ ...

# The particular case of spiked Deformed Wigner model

P.Biane 1997:

$$\omega_{\mu_{sc}, \nu}(\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu) = \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) < \frac{1}{\sigma^2} \right\}$$

$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu \begin{array}{c} \xrightarrow{\omega_{\mu_{sc}, \nu}} \\ \xleftarrow{h_{\omega_{\mu_{sc}, \nu}}} \end{array} \mathcal{O}, \quad h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_{\nu}(z).$$

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Previous conjecture becomes:

$\implies$  If  $\int \frac{1}{(\theta_i - x)^2} d\nu(x) < \frac{1}{\sigma^2}$ ,  $\theta_i$  generates outliers in a neighborhood of  $\rho = \theta_i + \sigma^2 g_{\nu}(\theta_i) \in \mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu$ .

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## Remark

When  $A_N$  has finite rank,  $\nu = \delta_0$ , this condition corresponds to  $|\theta_i| > \sigma$  and then  $\rho = \theta_i + \frac{\sigma^2}{\theta_i}$ .

# General spiked deformed models

Solving the problem of outliers consists in solving an equation involving the free subordination function and the spikes of the perturbation

General spiked deformed models		
$M_N = A_N + B_N$ $\mu_{A_N} \rightarrow_{N \rightarrow +\infty} \nu$ $\mu_{B_N} \rightarrow_{N \rightarrow +\infty} \mu$ $\theta \in \text{Spect}(A_N)$ $\theta$ multiplicity $k_i$ $\theta \notin \text{supp}(\nu)$	$M_N = A_N^{1/2} B_N A_N^{1/2}$ $\mu_{A_N A_N^*} \rightarrow_{N \rightarrow +\infty} \nu$ $\mu_{B_N B_N^*} \rightarrow_{N \rightarrow +\infty} \mu$ $\theta \in \text{Spect}(A_N)$ $\theta$ multiplicity $k_i$ $\theta > 0, \theta \notin \text{supp}(\nu)$	$M_N = (A_N + B_N)(A_N + B_N)^*$ $\mu_{A_N A_N^*} \rightarrow_{N \rightarrow +\infty} \nu$ $\mu_{B_N B_N^*} \rightarrow_{N \rightarrow +\infty} \mu$ $\theta \in \text{Spect}(A_N A_N^*)$ $\theta$ multiplicity $k_i$ $\theta > 0, \theta \notin \text{supp}(\nu)$
$\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu \boxplus \nu$	$\mu_{M_N} \rightarrow_{N \rightarrow +\infty} \mu \boxtimes \nu$	$\mu_{M_N} \rightarrow_{N \rightarrow +\infty} (\sqrt{\mu} \boxplus_c \sqrt{\nu})^2$
$g_\tau(z) = \int_{\mathbb{R}} \frac{d\tau(x)}{z-x}$	$\Psi_\tau(z) = \frac{1}{z} g_\tau\left(\frac{1}{z}\right) - 1$	$H_{\sqrt{\tau}}^{(c)} = \frac{c}{z} g_\tau\left(\frac{1}{z}\right)^2 + (1-c) g_\tau\left(\frac{1}{z}\right)$
$g_{\mu \boxplus \nu}(z) = g_\nu(\omega_{\mu, \nu}(z))$	$\Psi_{\mu \boxtimes \nu}(z) = \Psi_\nu(F_{\mu, \nu}(z))$	$H_{\sqrt{\mu \boxplus_c \nu}}^{(c)}(z) = H_{\sqrt{\nu}}^{(c)}(\Omega_{\mu, \nu}(z))$
$k_i$ outliers of $M_N$ in the neighborhood of each $\rho$ s.t $\omega_{\mu, \nu}(\rho) = \theta$	$k_i$ outliers of $M_N$ in the neighborhood of each $\rho$ s.t $\frac{1}{F_{\mu, \nu}(1/\rho)} = \theta$	$k_i$ outliers of $M_N$ in the neighborhood of each $\rho$ s.t $\frac{1}{\Omega_{\mu, \nu}(1/\rho)} = \theta$

When  $A_N$  has full rank, such results are proved for

$A_N + B_N$	$(A_N)^{1/2} B_N (A_N)^{1/2}$ $A_N > 0$	$(A_N + B_N)(A_N + B_N)$
$B_N = \text{Wigner}$ C.&D-M.&F.&F. (2011)	$B_N = \text{Wishart}$ Rao&Silverstein (2010) Bai&Yao(2012)	$B_N$ i.i.d matrix $A_N$ diagonal C. (2013)
$B_N = U_N D_N U_N^*$ $U_N$ Haar, $D_N$ deterministic $\mu_{D_N} \rightarrow_{N \rightarrow +\infty} \mu$ B.&B.&C.&F. (2012)	$B_N = U_N D_N U_N^*$ $U$ Haar, $D_N \geq 0$ deterministic $\mu_{D_N} \rightarrow_{N \rightarrow +\infty} \mu$ B.&B.&C.&F. (2014)	

(C.&D-M.&F.&F.=C.&Donati-Martin&Féral&Février)

(B.&B.&C.&F.= Belinschi&Bercovici&C.&Février)

FOR ALL DEFORMED MODELS IN THE PREVIOUS ARRAY, if  $\theta_i$  has multiplicity  $k_i$  in the spectrum of the deformation, then for each  $\rho$  which is a solution of the corresponding subordination equation (for instance in the additive case  $\omega_{\mu,\nu}(\rho) = \theta_i$ ), almost surely, for all large  $N$ , there are exactly  $k_i$  eigenvalues of  $M_N$  in a neighborhood of  $\rho$ .

For matricial models in the first row of the previous array, given one spike  $\theta$  there is at most one solution  $\rho$  for the corresponding equation and everything is explicit:

- Deformed Wigner

$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu \xrightarrow[\phi_1]{\omega_{\mu_{sc}\nu}} \mathcal{O}_1 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_1, \quad \rho = \phi_1(\theta) = \theta + \sigma^2 g_\nu(\theta).$$

- Sample covariance matrices

$$\mathbb{R} \setminus \{\text{support } \mu_{MP} \boxtimes \nu \cup \{0\}\} \xrightarrow[\phi_2]{x \mapsto \frac{1}{F_{\mu_{MP}, \nu}(1/x)}} \mathcal{O}_2 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_2, \quad \rho = \phi_2(\theta) = \theta + c\theta \int \frac{t}{\theta-t} d\nu(t).$$

- Information-Plus-Noise type model

$$\mathbb{R} \setminus (\sqrt{\mu_{MP}} \boxtimes_c \sqrt{\nu})^2 \xrightarrow[\phi_3]{x \mapsto \frac{1}{\Omega_{\mu_{MP}, \nu}(1/x)}} \mathcal{O}_3 \subset \mathbb{R} \setminus \text{support } \nu,$$

$$\theta \in \mathcal{O}_3, \quad \rho = \phi_3(\theta) = \theta(1 + c g_\nu(\theta))^2 + (1 - c)(1 + c g_\nu(\theta))$$

BUT

CONCERNING SOME MODELS OF THE LAST ROW OF THE PREVIOUS ARRAY (deformations of unitarily invariant matrices), the restriction to the real line of **some subordination maps may be many-to-one** so that **for one  $\theta_i$ , there may exist several distinct  $\rho$  solving the corresponding subordination equation.**

$\implies$  **For such models, a single spiked eigenvalue of  $A_N$  may generate several outliers of  $M_N$ .**

# Example: Deformed GUE

$$M_N = GUE\left(N, \frac{1}{4N}\right) + \text{diag}\left(\underbrace{-1, \dots, -1}_{\frac{N-1}{2}}, \underbrace{1, \dots, 1}_{\frac{N}{2}}, 10\right)$$

$$\nu = \frac{1}{2}\delta_1 + \frac{1}{2}\delta_{-1}, \quad \sigma^2 = \frac{1}{4} \quad \text{and} \quad \theta = 10.$$

$$\int \frac{1}{(10-x)^2} d\nu(x) < 4,$$

$$g_{\mu_{sc} \boxplus \nu}(z) = g_\nu(\omega_{\mu_{sc}, \nu}^{(1)}(z))$$

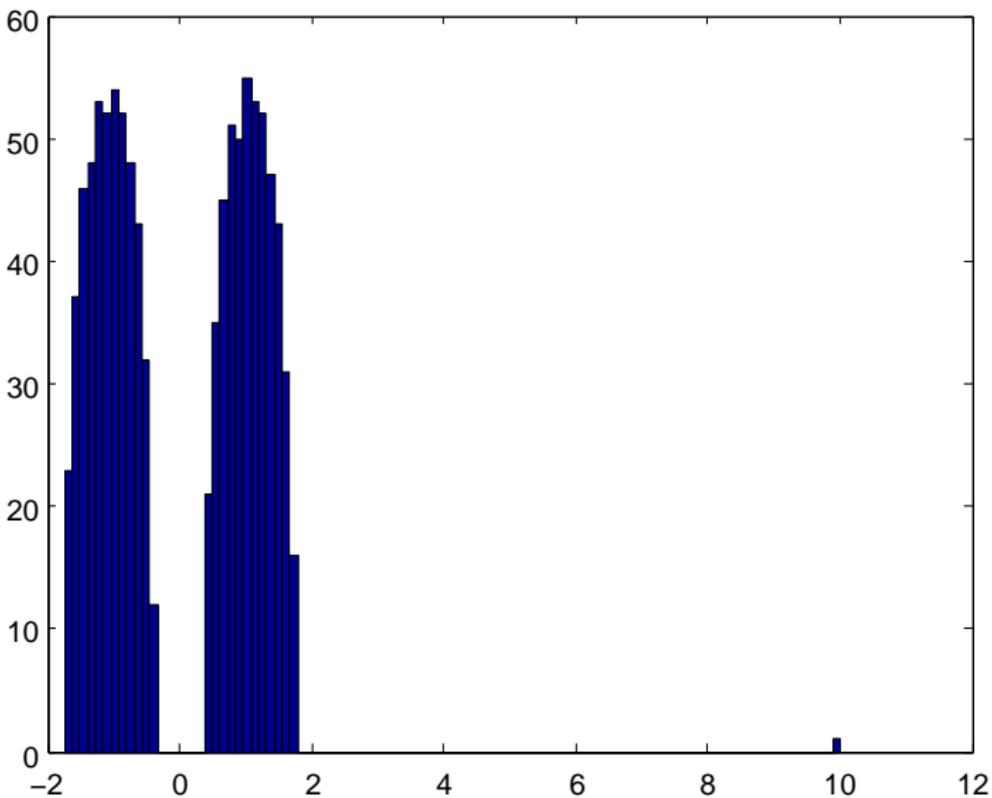
$\omega_{\mu_{sc}, \nu}^{(1)}$  is injective on  $\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu$

$\omega_{\mu_{sc}, \nu}^{(1)}(\rho) = 10$  has 1 solution

$$\rho = 10 + \frac{1}{4} \left( \frac{1}{2} \frac{1}{10-1} + \frac{1}{2} \frac{1}{10-1} \right) \approx 10,05.$$

## General spiked deformed models

N=1000



# Example

$W^G := GUE(N-1, \frac{1}{4(N-1)})$ ,  $U_N$  Haar matrix independent from  $W^G$

$$M_N = \begin{pmatrix} W^G & (0) \\ (0) & 10 \end{pmatrix} + U_N \text{diag}(\underbrace{-1, \dots, -1}_{\frac{N}{2}}, \underbrace{1, \dots, 1}_{\frac{N}{2}}) U_N^*$$

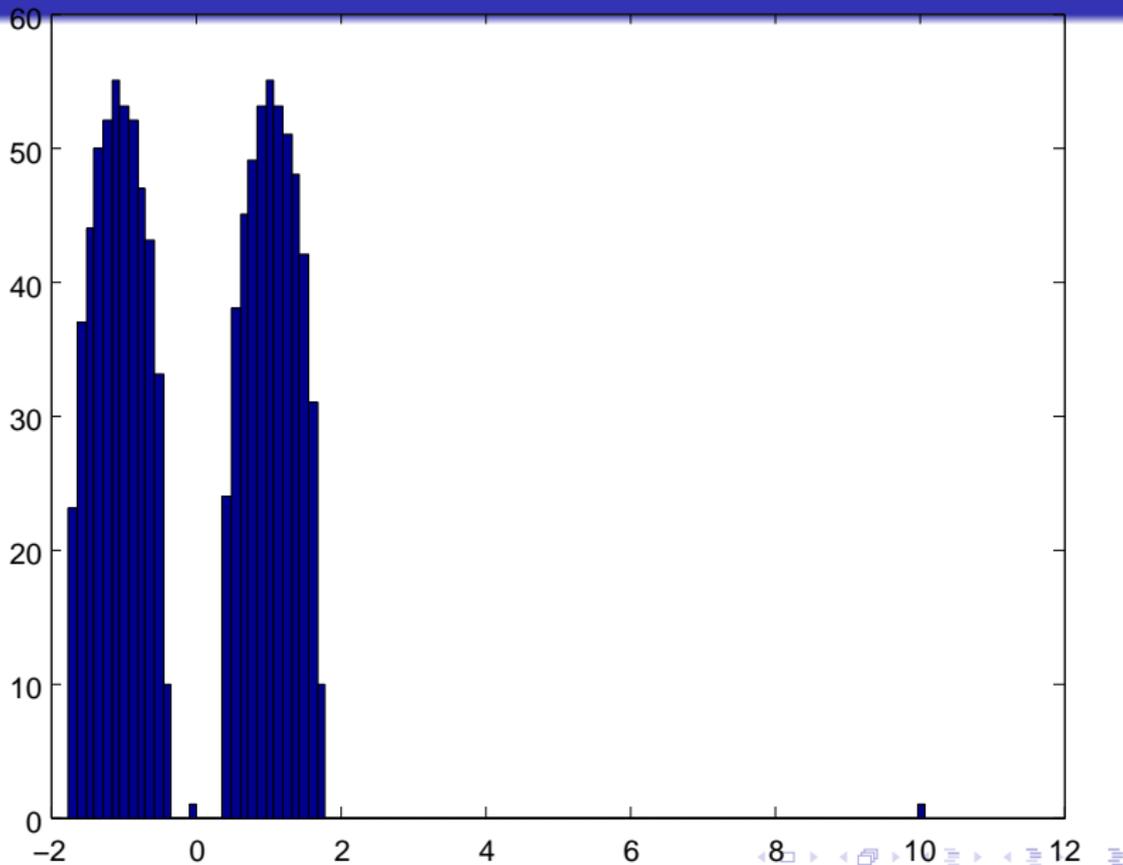
This is not a spiked deformed GUE model and now, the spike  $\theta = 10$  is associated to the matrix approximating the semicircular distribution!!!!!!!

$$g_{\mu_{sc} \boxplus \nu}(z) = g_{\nu}(\omega_{\mu_{sc}, \nu}^{(1)}(z)) = g_{\mu_{sc}}(\omega_{\nu, \mu_{sc}}^{(2)}(z))$$

$\omega_{\mu_{sc}, \nu}^{(1)}$  is injective on  $\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu$  but  $\omega_{\nu, \mu_{sc}}^{(2)}$  may be many to one.  $\omega_{\nu, \mu_{sc}}^{(2)}(\rho) = 10$  has 2 solutions  $\rho_1$  and  $\rho_2$ .

## General spiked deformed models

N=1000



More funny...(Belinschi&Bercovici&C.&Février (2014))

$$M_N = U_N B_N U_N^* + A_N, \quad U_N \text{ Haar unitary, } \quad A_N, B_N \text{ deterministic diagonal}$$

$$\mu_{B_N} \rightarrow \mu, \quad \mu_{A_N} \rightarrow \nu$$

$\theta \notin \text{supp}(\nu)$ , with multiplicity  $k$  in the spectrum of  $A_N$

$\alpha \notin \text{supp}(\mu)$ , with multiplicity  $l$  in the spectrum of  $B_N$

whereas the other eigenvalues are uniformly close to the limiting supports.

$$g_{\mu \boxplus \nu}(z) = g_{\nu}(\omega_{\mu, \nu}^{(1)}(z)) = g_{\mu}(\omega_{\nu, \mu}^{(2)}(z)).$$

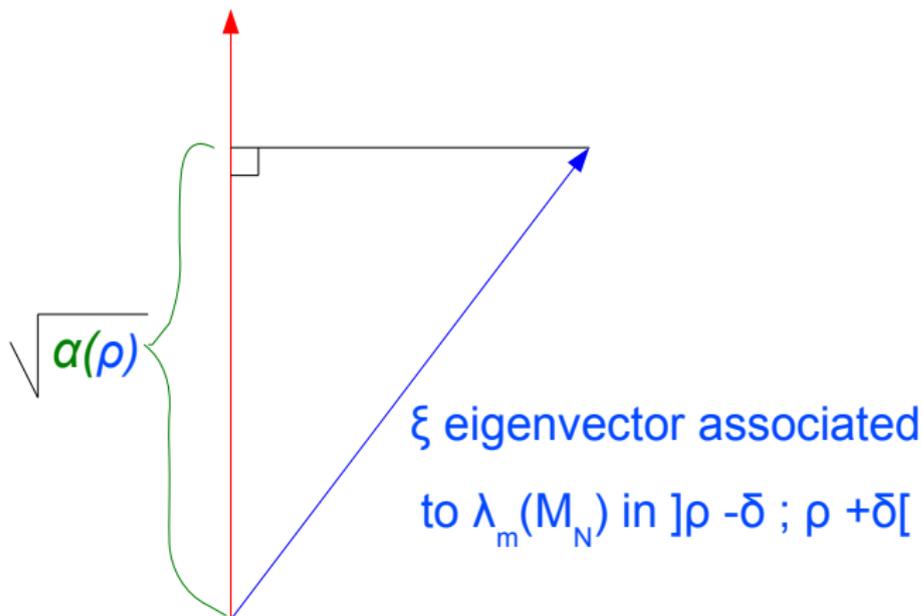
If there exists  $\rho \in \mathbb{R} \setminus \text{supp}(\mu \boxplus \nu)$  such that

$$\begin{cases} \omega_{\nu, \mu}^{(2)}(\rho) = \alpha \\ \omega_{\mu, \nu}^{(1)}(\rho) = \theta \end{cases}$$

then for all large  $N$ , there are  $k + l$  outliers of  $M_N$  in a neighborhood of  $\rho$ .

$k_j=1$  When  $N$  goes to infinity,

$V$  eigenvector associated to  $\theta_j = \lambda_p(A_N)$



$$\text{where } \alpha(\rho) = \begin{cases} \frac{1}{\omega'_{\mu,\nu}(\rho)} & \text{if } M_N = X_N + A_N \\ \frac{\rho F_{\mu,\nu}(1/\rho)}{F'_{\mu,\nu}(1/\rho)} & \text{if } M_N = A_N^{1/2} X_N A_N^{1/2} \end{cases}$$

This result is proved

- for  $X_N + A_N$  when  $X_N$  is a **Wigner** matrix [C. 2011] and when the distribution of  $X_N$  is **unitarily invariant** [Benaych-Georges&Rao (2010) Belinschi&Bercovici& C.& Février (2014)]
- for  $A_N^{1/2} X_N A_N^{1/2}$  when  $X_N$  is a **Wishart** matrix [C. 2011] and when the distribution of  $X_N$  is **unitarily invariant** [Benaych-Georges&Rao (2010) Belinschi&Bercovici& C.& Février (2014)]

(for information-plus-noise type models, results of Benaych-Georges-Rao (2012) dealing with finite rank perturbations)

# “Deterministic fundamental measure”

## Deformed Wigner matrices

$W_N$ : a Wigner matrix ,  $A_N$ : Hermitian deterministic.

$$M_N = \frac{W_N}{\sqrt{N}} + A_N$$

## The deterministic measure

$$\mu_{A_N} \boxplus \mu_{sc}$$

plays a central role in the study of the spectrum.

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plays a central role in the study of the spectrum.

- “No eigenvalue outside the support of this measure”
- “Exact separation phenomenon” involving this measure
- Universality of the fluctuations at some edges of the support of this measure

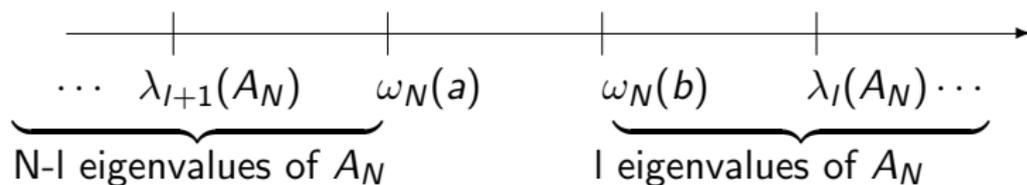
## Exact separation phenomenon for deformed Wigner model

$\omega_N(z) = z - \sigma^2 g_{\mu_{sc} \boxplus \mu_{A_N}}(z)$  (the subordination map of  $\mu_{sc} \boxplus \mu_{A_N}$  w.r.t  $\mu_{A_N}$ )

Then, almost surely, for large  $N$ ,

$$[a, b] \subset \mathbb{R} \setminus \text{support}(\mu_{sc} \boxplus \mu_{A_N}) \iff [\omega_N(a), \omega_N(b)]$$

$$\text{gap in Spect}(M_N) \iff \text{gap in Spect}(A_N)$$



# Exact separation phenomena

involving the additive, multiplicative, rectangular subordination maps

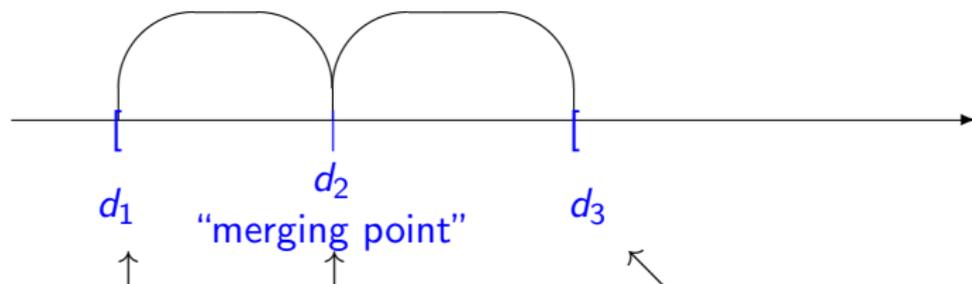
- [Deformed Wigner matrices](#)  
C.&Donati-Martin&Féral&Février (2011)
- [Sample Covariance matrices](#)  
Bai&Silverstein (1999)
- [Information-Plus-Noise type models](#)  
Loubaton&Vallet (2011) C. (2014)

# FLUCTUATIONS AT EDGES



Assumption:  $\forall u \in \text{support}(\nu), \int \frac{d\nu(x)}{(u-x)^2} > \frac{1}{\sigma^2}$ .

Example,  $p$  density of  $\mu_{sc} \boxplus \nu$



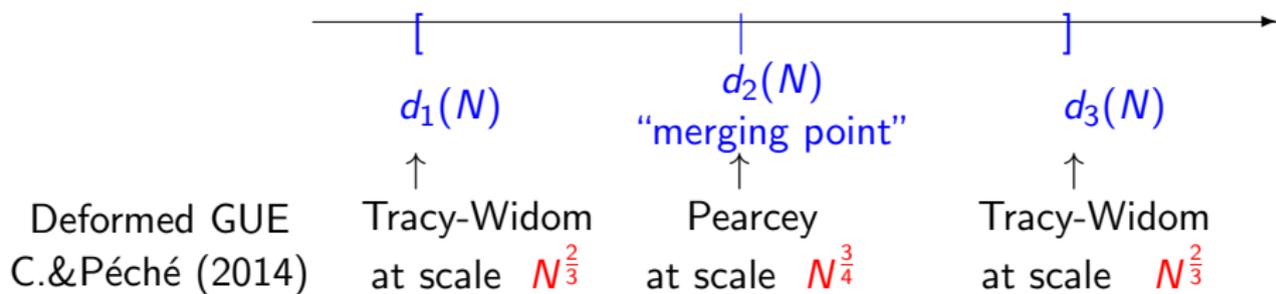
$$p(x) \sim C|d_1 - x|^{\frac{1}{2}} \quad p(x) \sim C|d_2 - x|^{\frac{1}{3}} \quad p(x) \sim C|d_3 - x|^{\frac{1}{2}} \quad (\text{Biane 97})$$

For  $\epsilon$  small enough, for all large  $N$ ,

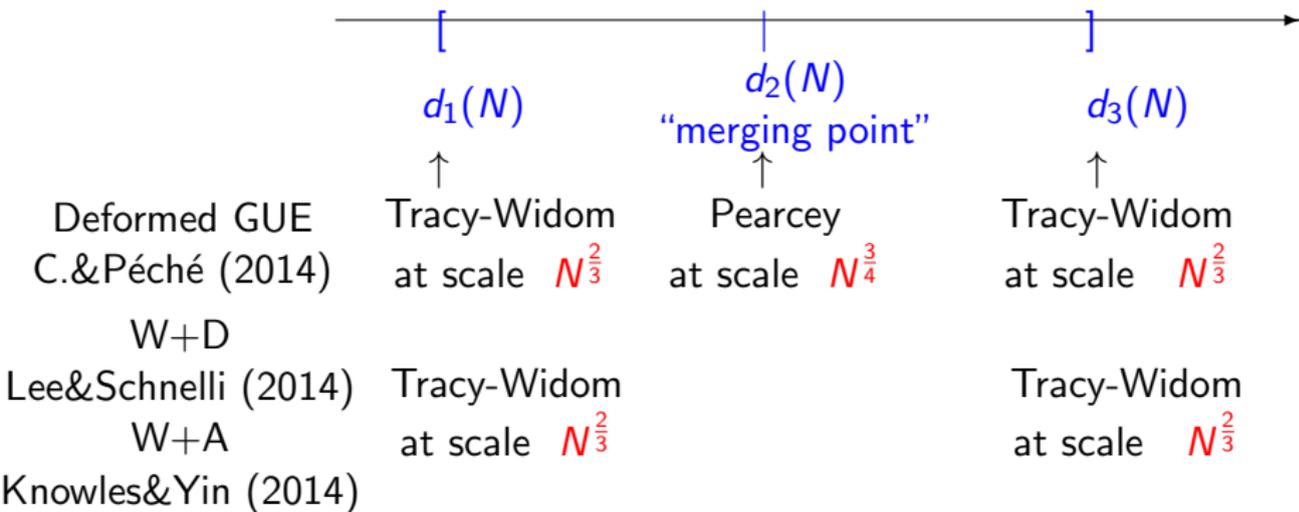
$\exists! d_1(N)$  left edge of  $\mu_{A_N} \boxplus \mu_{sc}$  in  $]d_1 - \epsilon; d_1 + \epsilon[$

$\exists! d_2(N)$  "merging point" of  $\mu_{A_N} \boxplus \mu_{sc}$  in  $]d_2 - \epsilon; d_2 + \epsilon[$

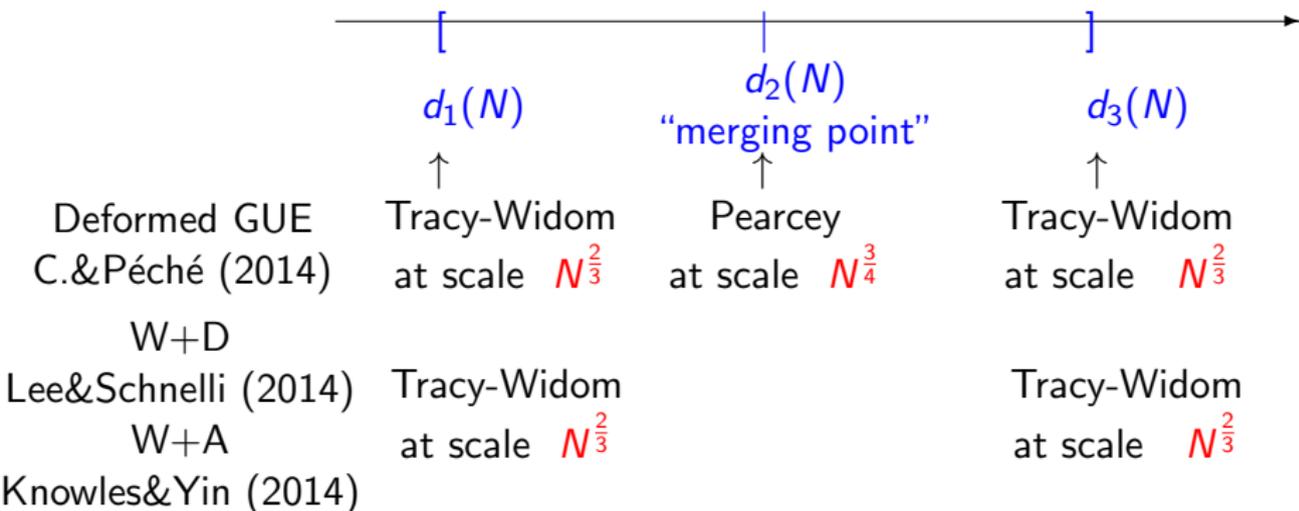
$\exists! d_3(N)$  right edge of  $\mu_{A_N} \boxplus \mu_{sc}$  in  $]d_3 - \epsilon; d_3 + \epsilon[$



## Fluctuations at edges



## Fluctuations at edges



$\implies$  Universality of the fluctuations around the edges  $d_i(N)$  of  $\mu_{A_N} \boxplus \mu_{sc}$

Considering fluctuations around  $d_i$  (instead of  $d_i(N)$ ) may imply making assumption on the rate of convergence of  $g_{\mu_{A_N}}$  towards  $g_{\nu}$ .  
Scherbina (2011)

## Remark

*Previous works of Brezin&Hikami (1998), Aptekarev&Bleher&Kuijilars (2004), (2005), Adler&Cafasso&Van Moerbeke (2007), (2011) when  $\mu_{A_N} = \nu$  is a finite combination of Dirac Delta masses.*

# Fluctuation of outliers

$$M_N = GUE(N, \frac{\sigma^2}{N}) + A_N, \quad A_N = \text{diag}(\beta_1, \dots, \beta_{N-r}, \theta_1, \dots, \theta_J)$$

$\mu_{A_N} \rightarrow \nu$ ,  $\nu$  compactly supported.

- $\max_{i=1}^{N-r} \text{dist}(\beta_i(N), \text{supp}(\nu)) \xrightarrow{N \rightarrow \infty} 0$
- a **finite** number  $J$  of **fixed** (independent of  $N$ ) eigenvalues **(SPIKES)**  $\theta_1 > \dots > \theta_J$ ,  $\forall i = 1, \dots, J$ ,  $\theta_i \notin \text{supp}(\nu)$ , each  $\theta_j$  having a fixed multiplicity  $k_j$ .

Let  $\theta_i$  be such that  $\int \frac{d\nu(x)}{(\theta_i - x)^2} < 1$  and  $\rho_{\theta_i} = h_{\mu_{sc}, \nu}(\theta_i)$ . Then, for  $\epsilon > 0$  small enough, for all large  $N$ ,  $\text{supp}(\mu_{sc} \boxplus \mu_{A_N})$  has a unique connected component  $[L_i(N); D_i(N)]$  inside  $] \rho_{\theta_i} - \epsilon; \rho_{\theta_i} + \epsilon [$ .

Moreover, the  $k_i$  outliers of  $M_N$  close to  $\rho_{\theta_i}$  fluctuate at rate  $\sqrt{N}$  around  $\frac{L_i(N) + D_i(N)}{2}$  as the eigenvalues of a  $k_i \times k_i$  GUE.

# Some remarks

## Remark

*Analog results at soft edges for Sample covariance matrices by Hachem&Hardy&Najim (2014), Lee&Schnelli (2014), Bao&Pan&Zhou (2014) and for outliers of Sample covariance matrices by Bai&Yao (2012)*

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## Remark

*According to previous studies dealing with finite rank perturbations, universality of fluctuations of outliers of deformed Wigner models is not expected in full generality.*

## Example

support  $\nu$ 

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$$\mathbb{R} \setminus \text{support } \mu_{sc} \boxplus \nu = h_{\mu_{sc}, \nu}(\mathcal{O}), \quad h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_{\nu}(z)$$

$$\mathcal{O} := \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) < \frac{1}{\sigma^2} \right\}$$

$$\begin{aligned} {}^c\mathcal{O} &:= \overline{\text{support } \nu \cup \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) \geq \frac{1}{\sigma^2} \right\}} \\ &= \left\{ u \in \mathbb{R}, \int \frac{1}{(u-x)^2} d\nu(x) > \frac{1}{\sigma^2} \right\} \end{aligned}$$

Each connected component of  ${}^c\mathcal{O}$  contains at least one connected component of support  $\nu$

support  $\nu$



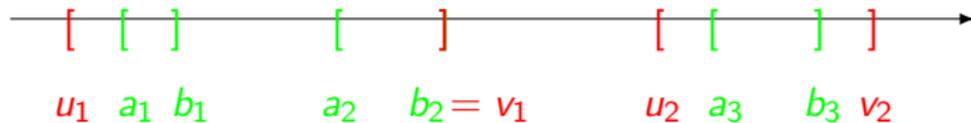
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${}^c\mathcal{O}$  support  $\nu$

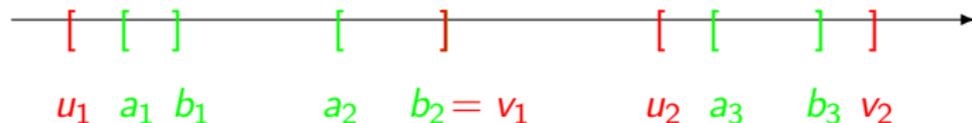


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${}^c\mathcal{O}$  support  $\nu$



$$h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_\nu(z).$$

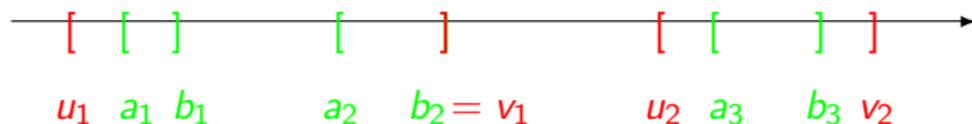
support  $\mu_{sc} \boxplus \nu$



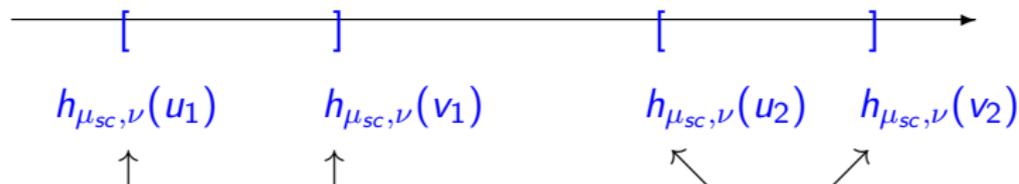
## Example

$\mu_{sc} \boxplus \nu$  is absolutely continuous.  $p$ : density of  $\mu_{sc} \boxplus \nu$

$^c\mathcal{O}$  support  $\nu$



support  $\mu_{sc} \boxplus \nu$



$$p(x) \sim C|d_i - x|^{\frac{1}{2}}$$

the singularity of  $p$   
may change!

$$p(x) \sim C|d_i - x|^{\frac{1}{2}}$$

**Example investigated by Lee & Schnelli (2013) :**

$$d\nu(x) := Z^{-1}(1+x)^a(1-x)^b f(x) 1_{[-1,1]}(x) dx$$

where  $a < 1, b > 1$ ,  $f$  is a strictly positive  $C^1$ -function and  $Z$  is a normalization constant.

$$\int \frac{1}{(1-x)^2} d\nu(x) = \frac{1}{\sigma_0^2}$$

$${}^c\mathcal{O} = \text{support } \nu \cup \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) \geq \frac{1}{\sigma^2} \right\}$$

$$h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_\nu(z)$$

$\forall \sigma > \sigma_0, {}^c\mathcal{O} = [u_\sigma; v_\sigma]$  with

$$u_\sigma < -1 < 1 < v_\sigma, \text{ support } \mu \boxplus \nu = [h_{\mu_{sc}, \nu}(u_\sigma); h_{\mu_{sc}, \nu}(v_\sigma)]$$

$$\longrightarrow p(x) \sim C(h_{\mu_{sc}, \nu}(v_\sigma) - x)^{\frac{1}{2}}$$

### Example investigated by Lee & Schnelli (2013) :

$$d\nu(x) := Z^{-1}(1+x)^a(1-x)^b f(x) 1_{[-1,1]}(x) dx$$

where  $a < 1, b > 1$ ,  $f$  is a strictly positive  $C^1$ -function and  $Z$  is a normalization constant.

$$\int \frac{1}{(1-x)^2} d\nu(x) = \frac{1}{\sigma_0^2}$$

$${}^c\mathcal{O} = \text{support } \nu \cup \left\{ u \in \mathbb{R} \setminus \text{support } \nu, \int \frac{1}{(u-x)^2} d\nu(x) \geq \frac{1}{\sigma^2} \right\}$$

$$h_{\mu_{sc}, \nu} : z \mapsto z + \sigma^2 g_\nu(z)$$

$\forall \sigma > \sigma_0, {}^c\mathcal{O} = [u_\sigma; v_\sigma]$  with

$$u_\sigma < -1 < 1 < v_\sigma, \text{ support } \mu \boxplus \nu = [h_{\mu_{sc}, \nu}(u_\sigma); h_{\mu_{sc}, \nu}(v_\sigma)]$$

$$\rightarrow p(x) \sim C(h_{\mu_{sc}, \nu}(v_\sigma) - x)^{\frac{1}{2}}$$

$\forall \sigma \leq \sigma_0, {}^c\mathcal{O} = [u_\sigma; 1], u_\sigma < -1, \text{ support } \mu \boxplus \nu = [h_{\mu_{sc}, \nu}(u_\sigma); h_{\mu_{sc}, \nu}(1)]$

$$\rightarrow p(x) \sim C(h_{\mu_{sc}, \nu}(1) - x)^b$$

Letting the perturbation  $A_N$  be random ...

Lee &amp; Schnelli (2014)

$$\frac{W_N}{\sqrt{N}} + \text{diag}(v_1, \dots, v_N), \quad v_i \text{ i.i.d. } \sim d\nu(x) = Z^{-1}(1+x)^a(1-x)^b f(x) 1_{[-1,1]}(x) dx$$

 $a < 1, b > 1, f > 0$   $C^1$ -function.

$$\sigma_0 \text{ defined by } \int \frac{1}{(1-x)^2} d\nu(x) = \frac{1}{\sigma_0^2}, \quad \text{support } \mu_{sc} \boxplus \nu = [d_\sigma^-; d_\sigma^+]$$

$$\bullet \forall \sigma > \sigma_0, p(x) \sim C(d_\sigma^+ - x)^{\frac{1}{2}}.$$

 $d_\sigma^+(N)$ : upper right edge of support  $\mu_{sc} \boxplus \mu_{A_N}$ ,

$$N^{2/3}(\lambda_1(M_N) - d_\sigma^+(N)) \xrightarrow{\mathcal{D}} TW,$$

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$$N^{\frac{1}{b+1}}(\lambda_1(M_N) - d_\sigma^+) \xrightarrow{\mathcal{D}} G_{b+1}(s)$$

as  $N$  goes to infinity, where  $G_{b+1}(s) = (1 - \exp(-(\frac{s}{c})^{b+1})) \mathbf{1}_{[0; +\infty[}(s)$   
 (Weibull distribution with parameters  $b+1$  and  $c = c(\nu, \sigma)$ ).

THANK YOU FOR YOUR  
ATTENTION!