

# Limit Theorems for Linear Eigenvalue Statistics of Random Matrices with Independent Entries

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- We are interested in the limiting laws of  $\mathcal{N}_n[\varphi]$  as  $n \rightarrow \infty$ .  
possibly after putting a normalization factor in front (LLN and CLT type)



- LT's is an active field of the RMT:  
*Marchenko, P 67; P 72; Girko 70-80; Bai-Silverstein 80-90, Costin-Lebowitz 95; Khorunzhy-Khoruzhenko-P. 96; Spohn 97; Johansson 98; Sinai-Soshnikov 98; Soshnikov 98, 00; Keating-Snaith 00; Cabanal-Duvillard 01; Diaconis-Evans 01; Guionnet 02; Bai-Silverstein 04; Anderson-Zeitouni 05; P. 06; Lytova-P. 09*

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- Noblesse oblige (L.P.): *Lyapunov* (first modern proof of CLT), *S. Bernstein* (first CLT for dependent r.v.'s), both from Kharkov

# Gaussian Ensembles

## Generalities

Definition:  $M_n = n^{-1/2} W_n$ ,  $W_n = \{W_{jk}\}_{j,k=1}^n$

$$P(dW) = Z_n^{-1} e^{-\text{Tr}W^2/4w^2} \prod_{1 \leq j \leq k \leq n} dW_{jk}.$$

Since

$$\text{Tr}W_n^2 = \sum_{1 \leq j \leq n} W_{jj}^2 + 2 \sum_{1 \leq j < k \leq n} W_{jk}^2,$$

the above implies that  $\{W_{jk}\}_{1 \leq j \leq k \leq n}$  are independent Gaussian random variables such that

$$\mathbf{E}\{W_{jk}\} = 0, \quad \mathbf{E}\{W_{jk}^2\} = w^2(1 + \delta_{jk}).$$

Gaussian Orthogonal Ensemble (GOE)

# Gaussian Ensembles

## Law of Large Numbers (LLN)

### Theorem

Let  $M_n$  be the GOE) and  $\mathcal{N}_n[\varphi]$  be a linear eigenvalue statistics of its eigenvalues. Then we have for any bounded and continuous  $\varphi : \mathbb{R} \rightarrow \mathbb{C}$  with probability 1:

$$\lim_{n \rightarrow \infty} \frac{1}{n} \sum_{l=1}^n \varphi(\lambda_l^{(n)}) = \int \varphi(\lambda) N_{sc}(d\lambda),$$

where the measure

$$N_{sc}(d\lambda) = \rho_{sc}(\lambda) d\lambda, \quad \rho_{sc}(\lambda) = (2\pi w^2)^{-1} \sqrt{4w^2 - \lambda^2} \mathbf{1}_{|\lambda| \leq 2w}$$

is known as the Wigner or the semicircle law.

Wigner 52 and many others.

# Gaussian Ensembles

## Law of Large Numbers (proof)

It suffices to consider the Normalized Counting Measure of eigenvalues (NCM)

$$N_n(\Delta) = \#\{\lambda_l^{(n)} \in \Delta\} / n, \quad \forall \Delta \subset \mathbb{R}$$

and its Stieltjes transform

$$g_n(z) = \int \frac{N_n(d\lambda)}{\lambda - z}, \quad \text{Im } z \neq 0,$$

determining  $N_n$ . Use now

(i) **Gaussian differentiation formula:**

$$\mathbf{E}\{\xi_l \Phi(\xi)\} = \mathbf{E}\{\xi_l^2\} \mathbf{E}\{\Phi'_l(\xi)\}, \quad l = 1, \dots, p;$$

(ii) **Poincaré-Nash-Chernoff inequality:**

$$\mathbf{Var}\{\Phi\} \leq \sum_{l=1}^p \mathbf{E}\{\xi_l^2\} \mathbf{E}\{|\Phi'_l|^2\}.$$

# Gaussian Ensembles

## Law of Large Numbers (proof)

By spectral theorem  $g_n(z) = n^{-1} \text{Tr}(M_n - z)^{-1}$ , by resolvent identity for  $f_n(z) = \mathbf{E}\{g_n(z)\}$

$$f_n(z) = z^{-1} + (zn)^{-1} \sum_{j,k=1}^n \mathbf{E}\{M_{jk} G_{kj}(z)\},$$

by (i)  $f_n(z) = z^{-1} + z^{-1} \mathbf{E}\{g_n^2(z)\}$ , and by (ii) (*Bose-Chatterjee 04; P. 05*)

$$\mathbf{Var}\{g_n(z)\} \leq 2w^2/n^2 |\text{Im } z|^4$$

while  $\mathbf{Var}\{g_n(z)\} \leq w^2/n |\text{Im } z|^4$  for random Schrodinger.

This leads to

$$f_{sc}(z) = z^{-1} + z^{-1} f_{sc}^2(z)$$

for  $\lim_{n \rightarrow \infty} f_n = f_{sc}$  uniformly on any compact set of  $\mathbb{C} \setminus \mathbb{R}$ , thus

$f_{sc}(z) = (\sqrt{z^2 - 4w^2} - z)/2w^2$  ( $\text{Im } f(z) \text{ Im } z > 0$ ). Convergence of  $g_n$  to  $f_{sc}$ , hence  $N_n$  to  $N_{sc}$  by Borel-Cantelli.



### Theorem

Let  $M_n$  be the GOE matrix,  $\varphi : \mathbb{R} \rightarrow \mathbb{R}$  be a differentiable function with a polynomially bounded derivative. Then  $\mathcal{N}_n^\circ[\varphi] = \mathcal{N}_n[\varphi] - \mathbf{E}\{\mathcal{N}_n[\varphi]\}$  converges in distribution to the Gaussian random variable with zero mean and the variance

$$V_{GOE}[\varphi] = \frac{1}{2\pi^2} \int_{-2w}^{2w} \int_{-2w}^{2w} \left( \frac{\varphi(\lambda_1) - \varphi(\lambda_2)}{\lambda_1 - \lambda_2} \right)^2 \times \frac{4w^2 - \lambda_1\lambda_2}{\sqrt{4w^2 - \lambda_1^2} \sqrt{4w^2 - \lambda_2^2}} d\lambda_1 d\lambda_2.$$

**A MIRACLE!**

# Gaussian Ensembles

## Central Limit Theorem (proof)

Proof is again based on the Gaussian differentiation formula and the bound

$$\mathbf{Var}\{\mathcal{N}_n[\varphi]\} \leq \frac{2w^2}{n} \mathbf{E}\{\mathrm{Tr}\varphi'(M_n)(\varphi'(M_n)^*)\} \leq 2w^2 \left(\sup_{\lambda \in \mathbb{R}} |\varphi'(\lambda)|\right)^2$$

for  $\mathcal{N}_n[\varphi] = \mathrm{Tr}\varphi(M)$  by Poincaré. We have to prove that

$$\lim_{n \rightarrow \infty} Z_n(x) = \exp\{-x^2 V_{GOE}[\varphi]/2\}, \quad Z_n(x) = \mathbf{E}\left\{e^{ix\overset{\circ}{\mathcal{N}}_n[\varphi]}\right\}$$

uniformly in  $x$ , varying on a finite interval of  $\mathbb{R}$ . Assume first that  $\varphi$  admits the Fourier transform  $\hat{\varphi}$  and  $(1 + |t|)|\hat{\varphi}(t)| \in L^1(\mathbb{R})$ . Then

$$Z_n(x) = 1 + \int_0^x Z'_n(y) dy, \quad Z'_n(x) = i \int \hat{\varphi}(t) Y_n(x, t) dt,$$

where

$$Y_n(x, t) = \mathbf{E}\left\{\overset{\circ}{u}_n(t) e_n(x)\right\}, \quad e_n(x) = e^{ix\overset{\circ}{\mathcal{N}}_n[\varphi]}, \quad u_n(t) = \mathrm{Tre}^{itM}.$$

# Gaussian Ensembles

## Central Limit Theorem (proof)

Use  $U_n(t) = e^{itM_n}$  (instead of  $G_n(z) = (M_n - z)^{-1}$ ) and the Duhamel formula

$$u_n(t) = n + i \int_0^t \sum_{j,k=1}^n M_{jk} U_{jk}(t_1) dt_1,$$

the differentiation formula, the Poincaré, and the Schwarz to obtain

$$Y_n(x, t) + 2w^2 \int_0^t dt_1 \int_0^{t_1} dt_2 \bar{v}_n(t_1 - t_2) Y_n(x, t_2) = xZ_n(x)A_n(t) + r_n(x, t),$$

$$A_n(t) = -2w^2 \int_0^t dt_1 \int e^{it_1\lambda} \varphi'(\lambda) \mathbf{E}\{N_n(d\lambda)\}, \quad \bar{v}_n(t) = \mathbf{E}\{n^{-1}\text{Tr}U(t)\}$$

$$|Y_n| \leq \sqrt{2}w|t| \sup_{t \in \mathbb{R}} |\varphi'(\lambda)|,$$

$$|(Y_n)'_t| \leq \sqrt{2}w(1 + w^2 t^2)^{1/2}, \quad |(Y_n)'_x| \leq 2w^2 t \sup_{\lambda \in \mathbb{R}} |\varphi'|.$$

Hence, there exists  $\{Y_{n_j}\}$  converging uniformly on any compact set of  $\mathbb{R}^2$  to  $Y$ , satisfying

# Gaussian Ensembles

## Central Limit Theorem (proof)

$$Y(x, t) + 2w^2 \int_0^t dt_1 \int_0^{t_1} dt_2 v(t_1 - t_2) Y(x, t_2) = xZ(x)A(t),$$

$$A(t) = -2w^2 \int_0^t dt_1 \int e^{it_1\lambda} \varphi'(\lambda) N_{sc}(d\lambda), \quad v(t) = \int e^{i\lambda t} N_{sc}(d\lambda).$$

This leads (by the Laplace transformation) to

$$Z(x) = 1 - V_{GOE} \int_0^x yZ(y)dy.$$

The equation is uniquely soluble and yield the result for  $(1 + |t|)|\hat{\varphi}(t) \in L^1(\mathbb{R})$ . General case of  $C^1$  (even *Lip* 1) test functions is obtained by Poincaré and approximations.

The scheme dates back to *Khorunzhy-Khoruzhenko-P. 96*, where the Stieljtes transform (the resolvent) was used, thus real analytic test functions. Here we use the Fourier transform and obtain  $C^1$  test functions.

# Wigner Ensembles

## Generalities

$$M_n = n^{-1/2} W_n, \quad W_n = \{W_{jk}^{(n)}\}_{j,k=1}^n$$

with  $W_{jk}^{(n)} = W_{kj}^{(n)} \in \mathbb{R}$ ,  $1 \leq j \leq k \leq n$  independent and

$$\mathbf{E}\{W_{jk}^{(n)}\} = 0, \quad \mathbf{E}\{(W_{jk}^{(n)})^2\} = (1 + \delta_{jk})w^2,$$

i.e. the two first moments of the entries coincide with those of the GOE or

$$\mathbf{P}(dW_n) = \prod_{1 \leq j \leq k \leq n} F_{jk}^{(n)}(dW_{jk}),$$

where  $F_{jk}^{(n)}$  has above moments. The GOE corresponds to

$$F_{jk}^{(n)}(dW) = \frac{1}{(2\pi\sigma_{jk}^2)^{1/2}} e^{-W^2/2\sigma_{jk}^2} dW, \quad \sigma_{jk}^2 = (1 + \delta_{jk})w^2.$$

# Wigner Ensembles

## Law of Large Numbers (semicircle law)

### Theorem

Let  $M_n = n^{-1/2}W_n$  be the Wigner matrix, satisfying the L2 (à la Lindeberg)

$$\lim_{n \rightarrow \infty} n^{-2} \sum_{1 \leq j \leq k \leq n} \int_{|w| \geq \tau \sqrt{n}} W^2 F_{jk}^{(n)}(dW), \quad \forall \tau > 0.$$

and  $N_n$  be the Normalized Counting Measure of its eigenvalues. Then with *p.1*:  $\lim_{n \rightarrow \infty} N_n(\Delta) = N_{sc}(\Delta)$ ,  $\forall \Delta \subset \mathbb{R}$  (macroscopic universality).

*P. 72; Girko 75.* No Poincaré but the martingale-type bounds  $\mathbf{E}\{|N_n^\circ(\Delta)|^4\} = O(n^{-2})$ . Thus, it suffices to prove that if  $M_n$  is the Wigner matrix and  $\widehat{M}_n$  is the corresponding GOE, then

$$R_n(z) := \mathbf{E}\{n^{-1}\text{Tr}(M_n - z)^{-1}\} - \mathbf{E}\{n^{-1}\text{Tr}(\widehat{M}_n - z)^{-1}\} \rightarrow 0, \quad n \rightarrow \infty$$

uniformly on a compact set of  $\mathbb{C} \setminus \mathbb{R}$ , cf recent results by *Erdos et al 09*

for  $\text{Im } z = O(n^{-1})$

# Wigner Ensembles

## Law of Large Numbers (proof)

Proof is based on

- **General differentiation formula** (*Khorunzhy-Khoruzhenko-P. 95*):  
If  $\mathbf{E}\{|\zeta|^{p+2}\} < \infty$ ,  $p \in \mathbb{N}$ ,  $\Phi : \mathbb{R} \rightarrow \mathbb{C}$  of  $C^{p+1}$  with bounded derivatives, then

$$\mathbf{E}\{\zeta \Phi(\zeta)\} = \sum_{j=0}^p \frac{\kappa_{j+1}}{j!} \mathbf{E}\{\Phi^{(j)}(\zeta)\} + \varepsilon_p,$$
$$|\varepsilon_p| \leq C_p \mathbf{E}\{|\zeta|^{p+2}\} \sup_{t \in \mathbb{R}} |\Phi^{(p+1)}(t)|,$$

where  $\{\kappa_l\}_{l=1}^{\infty}$  are the cumulants of  $W_{12}$ . Note that the  $l = 1$  term is "Gaussian".

- **"Interpolation trick"** (*P. 2000*): use the product space of the Wigner  $M_n$  and the GOE  $\widehat{M}_n$  with the same first and second moments and set

$$M_n(s) = s^{1/2} M_n + (1-s)^{1/2} \widehat{M}_n, \quad 0 \leq s \leq 1,$$

# Wigner Ensembles

## Law of Large Numbers (proof)

Assume first  $w_3 := \sup_n \max_{1 \leq j \leq k \leq n} \mathbf{E} \left\{ |W_{jk}^{(n)}|^3 \right\} < \infty$  and write

$$R_n(z) = \int_0^1 \frac{d}{ds} \mathbf{E} \left\{ n^{-1} \text{Tr} (M_n(s) - z)^{-1} \right\} ds = \frac{1}{2} \int_0^1 (T_1 - T_2) ds$$

$$T_1 = (n^3 s)^{-1/2} \sum_{1 \leq j, k \leq n} \mathbf{E} \{ W_{jk}^{(n)} (G^2)_{jk} \},$$

$$T_2 = (n^3 (1-s))^{-1/2} \sum_{1 \leq j, k \leq n} \mathbf{E} \{ \widehat{W}_{jk} (G^2)_{jk} \}.$$

Apply to  $T_1$  the general differentiation formula with  $p = 1$  and  $\Phi = (G_n^2)_{jk}$  and to  $T_2$  the Gaussian differentiation formula. We have the **cancelation**, resulting only in  $\varepsilon_1$ :

$$|\varepsilon_1| \leq \frac{C_1 w_3}{n^{5/2}} \sum_{1 \leq j \leq k \leq n} \sup_{M \in \mathcal{S}_n} |D_{jk} (G_n^2)_{jk}| \leq \frac{C'_1 w_3}{n^{1/2} |\Im z|^4}, \quad D_{jk} = \frac{\partial}{\partial M_{jk}}.$$

$\mathcal{S}_n$  is the set of  $n \times n$  real symmetric matrices.



# Wigner Ensembles

Central Limit Theorem (zero excess)

## Theorem

Let  $M_n = n^{-1/2}W_n$ ,  $W_n = \{W_{jk}^{(n)}\}_{j,k=1}^n$  be the real symmetric Wigner random matrix. Assume that  $\mu_4 = \mathbf{E}\{(W_{jk}^{(n)})^4\}$  does not depend on  $j, k$  and  $n$ ,  $\kappa_4 = \mu_4 - 3w^4 = 0$ , and the L4:

$$\lim_{n \rightarrow \infty} n^{-2} \sum_{j,k=1}^n \int_{|W| \geq \tau \sqrt{n}} W^4 F_{jk}^{(n)}(dW) = 0, \quad \forall \tau > 0,$$

If  $\varphi$  possesses the Fourier transform  $\hat{\varphi}$  and  $(1 + |t|^5)|\hat{\varphi}(t)| \in L^1(\mathbb{R})$ , then  $\mathcal{N}_n[\varphi]$  converges in distribution to the Gaussian random variable with zero mean the GOE variance (again the macroscopic universality, even a bit more).

Proof by the "interpolation" trick from the GOE. For "Lindeberg-4" see *KKP, 95*.

# Wigner Ensembles

Central Limit Theorem (general case)

## Theorem

Let  $M_n = n^{-1/2}W_n$  be the real symmetric Wigner random matrix,  $\mu_4 = \mathbf{E}\{(W_{jk}^{(n)})^4\}$  do not depend on  $j, k$  and  $n$  and

$$w_6 := \sup_n \max_{1 \leq j \leq k \leq n} \mathbf{E}\{(W_{jk}^{(n)})^6\} < \infty.$$

If  $(1 + |t|^5)|\hat{\varphi}(t)| \in L^1(\mathbb{R})$ , then  $\hat{\mathcal{N}}_n[\varphi] = \mathcal{N}_n[\varphi] - \mathbf{E}\{\mathcal{N}_n[\varphi]\}$  converges in distribution to the Gaussian random variable of zero mean and of variance

$$\begin{aligned} V_{Wig}[\varphi] &= V_{GOE}[\varphi] + \frac{\kappa_4}{2\pi^2 w^8} I_{Wig}^2, \\ I_{Wig} &= \int_{-2w}^{2w} \varphi(\mu) \frac{2w^2 - \mu^2}{\sqrt{4w^2 - \mu^2}} d\mu, \end{aligned}$$

Assume that  $\kappa_4 \neq 0$ , then:

$I_{Wig} = 0$ : the GOE CLT, e.g. for an ODD  $\varphi$ .

$I_{Wig} \neq 0$ : a modified CLT, generically and, in particular, for an EVEN  $\varphi$  such that

$$\int_0^{2w} \varphi(\mu) \frac{2w^2 - \mu^2}{\sqrt{4w^2 - \mu^2}} d\mu \neq 0.$$

# Wigner Ensembles

CLT ( $O(1)$  bound for the variance of linear statistics)

Proof: by combining the schemes of proof of the CLT for the GOE and the "interpolation" trick, in particular, by proving and using

## Theorem

Let  $M_n = n^{-1/2} W_n$  be the real symmetric Wigner random matrix and  $\mathcal{N}_n[\varphi]$  be the linear eigenvalue statistic of its eigenvalues. Assume that

$$w_6 := \sup_n \max_{1 \leq j, k \leq n} \mathbf{E} \left\{ \left| W_{jk}^{(n)} \right|^6 \right\} < \infty.$$

Then

$$\mathbf{Var}\{\mathcal{N}_n[\varphi]\} \leq C(w_6) \left( \int (1 + |t|^{5/2}) |\widehat{\varphi}(t)| dt \right)^{1/2},$$

where  $C(w_6)$  depends only on  $w_6$ .

The bound replaces the Poincaré one in the case of Wigner ensembles.

# Wigner Ensembles

CLT (origin of new term in the variance)

$$Y_n(x, t) = \sum_{a=1}^3 T_a + \varepsilon_3,$$

where now

$$T_a = \frac{i}{a!n^{(a+1)/2}} \int_0^t \sum_{j,k=1}^n \kappa_{a+1,jk} \mathbf{E} \{ D_{jk}^a(U_{jk}(s)e_n^\circ(x)) \} ds, \quad D_{jk} = \frac{\partial}{\partial M_{jk}}$$

and

$$|\varepsilon_3| \leq C(x) w_6^{5/6} (1 + |t|^4) / n^{1/2}.$$

The term  $T_3$  contains  $U_{jj}(t_1)U_{jj}(t_2)U_{kk}(t_3)U_{kk}(t_4)$  Because of

$$D_{jk} U_{ab}(t) = i\beta_{jk} \int_0^t ds [U_{aj}(t-s)U_{bk}(s) + U_{bj}(t-s)U_{ak}(s)].$$

These are only combinations of  $U$ 's that contribute.

# Wigner Ensembles

Universality Classes w.r.t. CLT

*Universality class w.r.t. to the CLT: the set of random matrices, having the same CLT (variance) for linear eigenvalue statistics.*

Universality classes of the Wigner matrices w.r.t. the CLT are indexed by the first two even moments of their off-diagonal entries:

$$w^2 = \mathbf{E}\{(W_{jk}^{(n)})^2\}, \mu_4 = \mathbf{E}\{(W_{jk}^{(n)})^4\}, 1 \leq j < k \leq n$$

(two dimensional moduli space).

An example of "collective theorem", *Linnik 70's*.

The Gaussian universality classes:  $\kappa_4 := \mu_4 - 3w^4 = 0$ .

In the conventional probability setting for the CLT of independent random variables  $\{\xi_l^{(n)}\}_{l=1}^n$  the universality classes w.r.t. the CLT of linear statistics are indexed by a single parameter, the variance  $\sigma^2 = \mathbf{E}\{(\xi_l^{(n)})^2\}$ . All classes are Gaussian.

# Sample Covariance Matrices

## Generalities

$M_{m,n}$  is a  $n \times n$  real symmetric matrix of the form (matrix  $\chi^2$ )

$$M_{m,n} = n^{-1} A_{m,n}^T A_{m,n},$$

with  $A_{m,n} = \{A_{\alpha j}\}_{\alpha,j=1}^{m,n}$  having i.i.d. entries ( $m$  observation on  $n$  parameters)

$$\mathbf{P}(dA_{m,n}) = \prod_{\alpha=1}^m \prod_{j=1}^n F_{\alpha j}^{(n)}(dA_{\alpha j})$$

such that

$$\mathbf{E}\{A_{\alpha j}\} = 0, \quad \mathbf{E}\{A_{\alpha j}\}^2 = a^2.$$

The case of i.i.d. Gaussian  $\{A_{\alpha j}\}_{\alpha,j=1}^{m,n}$  is known since the early 30's as the (white or null) Wishart Ensemble.

# Sample Covariance Matrices

## Law of Large Numbers

### Theorem

Let  $M_{m,n}$  be the sample covariance matrix such that  $\tau > 0$

$$\lim_{n \rightarrow \infty, m \rightarrow \infty, m/n \rightarrow c \in (0, \infty)} \frac{1}{mn} \sum_{\alpha=1}^m \sum_{j=1}^n \int_{|y| > \tau \sqrt{n}} y^2 F_{\alpha j}^{(n)}(dy) \rightarrow 0, \quad \forall \tau > 0$$

Then its Normalized Counting Measures  $N_n$  converges with probability 1 to the non-random measure:  $N_W(d\lambda) = \rho_W(\lambda) d\lambda$

$$\rho_W(\lambda) = (1 - c)_+ \delta_0 + \sqrt{((\lambda - a_-)(a_+ - \lambda))_+} / 2\pi a^2 \lambda,$$

where  $a_{\pm} = a^2(1 \pm \sqrt{c})^2$  (macroscopic universality again)

Marchenko, P. 67; Girko 70's.

Proof: Wishart by the resolvent identity, Gaussian differentiation formula, and Poincaré. General case as for the Wigner (i.e. the interpolation again).



# Sample Covariance Matrices

CLT (Wishart)

## Theorem

Let  $M_{m,n}$  be the Wishart random matrix. If  $\varphi$  is  $C^1$ , then  $\hat{\mathcal{N}}_n[\varphi]$  converges in distribution as  $m, n \rightarrow \infty$ ,  $m/n \rightarrow c > 0$  to the Gaussian random variable with zero mean and the variance

$$V_{Wish}[\varphi] = \frac{1}{2\pi^2} \int_{a_-}^{a_+} \int_{a_-}^{a_+} \left( \frac{\varphi(\lambda_1) - \varphi(\lambda_2)}{\lambda_1 - \lambda_2} \right)^2 \times \frac{4a^4 c - (\lambda_1 - \bar{a})(\lambda_2 - \bar{a})}{\sqrt{4a^4 c - (\lambda_1 - \bar{a})^2} \sqrt{4a^4 c - (\lambda_2 - \bar{a})^2}} d\lambda_1 d\lambda_2,$$

where  $\bar{a} = 1/2(a_- + a_+) = a^2(c + 1)$ .

Proof: By mimicking the proof for the GOE, i.e. by the Gaussian differentiation formula and Poincaré.

# Sample Covariance Matrices

CLT (4th cumulant is zero)

## Theorem

Let  $M_{m,n}$  be the sample covariance matrix such that:

- (i)  $w_5 := \sup_{m,n} \max_{1 \leq \alpha \leq m, 1 \leq j \leq n} \mathbf{E} \left\{ |A_{\alpha j}|^5 \right\} < \infty$
- (ii)  $\mu_4 = \mathbf{E} \left\{ |A_{\alpha j}|^4 \right\}$  do not depend on  $\alpha, j, m,$  and  $n,$  and

$$\kappa_4 := \mu_4 - 3a^4 = 0.$$

If  $(1 + |t|^5)|\hat{\varphi}(t)| \in L^1(\mathbb{R})$ , then  $\hat{\mathcal{N}}_n[\varphi]$  converges in distribution as  $m, n \rightarrow \infty, m/n \rightarrow c > 0$  to the Gaussian random variable with zero mean and the variance  $V_{Wish}[\varphi]$ .

Proof: by interpolation from Wishart.

*Bai, Silverstein, 04:* Stieltjes transform, real analytic test functions, direct and rather long proof.

# Sample Covariance Matrices

CLT (general case)

## Theorem

Let  $M_{m,n}$  be the sample covariance matrix such that:

(i)  $w_6 := \sup_{m,n} \max_{1 \leq \alpha \leq m, 1 \leq j \leq n} \mathbf{E} \left\{ |A_{\alpha j}|^6 \right\} < \infty$

(ii)  $\mu_4 = \mathbf{E} \left\{ |A_{\alpha j}|^4 \right\}$  do not depend on  $\alpha, j, m$ , and  $n$ .

If  $(1 + |t|^4) |\widehat{\varphi}(t)| \in L^1(\mathbb{R})$ , then  $\overset{\circ}{\mathcal{N}}_n[\varphi]$  converges in distribution as  $m, n \rightarrow \infty, m/n \rightarrow c > 0$  to the Gaussian random variable with zero mean and the variance

$$V_{Wish}[\varphi] + \frac{\kappa_4}{4c\pi^2 a^8} \left( \int_{a_-}^{a_+} \varphi(\mu) \frac{\mu - \bar{a}}{\sqrt{4a^4 c - (\mu - \bar{a})^2}} d\mu \right)^2.$$

Proof: by the same scheme as in the Wigner case, i.e., by combining the schemes of proof of the CLT for the Wishart case and the "interpolation" trick

$\varphi : \mathbb{R}^p \rightarrow \mathbb{R}$  symmetric and

$$\mathcal{U}_{p,n}[\varphi] = \sum_{0 \leq l_1 < \dots < l_p \leq n} \varphi(\lambda_{l_1}^{(n)}, \dots, \lambda_{l_p}^{(n)}),$$

$$\mathcal{N}_{p,n}[\varphi] = \sum_{l_1 = \dots = l_p = 1}^n \varphi(\lambda_{l_1}^{(n)}, \dots, \lambda_{l_p}^{(n)}).$$

We have:

1. with probability 1 (LLN):

$$\begin{aligned} \lim_{n \rightarrow \infty} n^{-p} \mathcal{U}_{p,n}[\varphi] &= \lim_{n \rightarrow \infty} n^{-p} \mathcal{N}_{p,n}[\varphi] \\ &= \int p \text{ times} \int \varphi(\lambda_1, \dots, \lambda_p) N_{scl}(d\lambda_1) \dots N_{scl}(d\lambda_p); \end{aligned}$$

## 2. in distribution

$$\begin{aligned}\lim_{n \rightarrow \infty} n^{-p+1} \mathcal{U}_{p,n}[\varphi] &= \lim_{n \rightarrow \infty} n^{-p+1} \mathcal{N}_{p,n}[\varphi] \\ &= \lim_{n \rightarrow \infty} \mathcal{N}_{1,n}[\varphi^*],\end{aligned}$$

where

$$\begin{aligned}\varphi^*(\lambda) &= \int (p-1) \text{ times } \int \varphi(\lambda, \lambda_2, \dots, \lambda_p) \\ &\quad \times N_{scl}(d\lambda_2) \dots N_{scl}(d\lambda_p),\end{aligned}$$

i.e., the CLT.

Both assertions are valid in the cases, where there are corresponding results for  $p = 1$ .

## Theorem

Let  $U_n$  be a  $n \times n$  unitary random matrix, whose probability law is the normalized Haar measure on  $U(n)$ , and  $A_n$  be a  $n \times n$  matrix such that

$$\lim_{n \rightarrow \infty} n^{-1} \text{Tr} A_n^* A_n = 1.$$

Then  $\text{Tr} U_n A_n$  converges in distribution to the standard complex Gaussian variable:  $\gamma = \gamma_1 + i\gamma_2$ ,  $\mathbf{E}\{\gamma_1\} = \mathbf{E}\{\gamma_2\} = 0$ ,  $\mathbf{E}\{\gamma_1^2\} = \mathbf{E}\{\gamma_2^2\} = 1/2$ .

*E. Borel 05* ( $A_n = \{\delta_{j1}\delta_{k1}\}_{j,k=1}^n$ ), *Diaconis et al 03*; *Snyady-Stolz 06*.

On the other hand, by using analogs of the differentiation formula and the Poincaré type inequality for  $U(n)$  and  $O(n)$  (*P.-Vasilchuk 06*) and the above scheme, a short and simple proof of the assertion can be obtained. Analogous assertions are valid for  $O(n)$  and  $Sp(n)$ .