

# *Exponential tail inequalities for eigenvalues of random matrices*

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## exponential tail inequalities

classical theme in probability and statistics

quantify the asymptotic statements

central limit theorems

large deviation principles

## classical exponential inequalities

sum of independent random variables

$$S_n = \frac{1}{\sqrt{n}} (X_1 + \cdots + X_n)$$

$0 \leq X_i \leq 1$  independent

$$\mathbb{P}(S_n \geq \mathbb{E}(S_n) + t) \leq e^{-t^2/2}, \quad t \geq 0$$

Hoeffding's inequality

same as for  $X_i$  standard Gaussian

central limit theorem

## measure concentration ideas

asymptotic geometric analysis

V. D. Milman (1970)

$$S_n = \frac{1}{\sqrt{n}} (X_1 + \cdots + X_n)$$

$F(X) = F(X_1, \dots, X_n), \quad F : \mathbb{R}^n \rightarrow \mathbb{R}$  Lipschitz

Gaussian sample

independent random variables (M. Talagrand 1995)

## concentration inequalities

$$S_n = \frac{1}{\sqrt{n}} (X_1 + \cdots + X_n)$$

$F(X) = F(X_1, \dots, X_n)$ ,  $F : \mathbb{R}^n \rightarrow \mathbb{R}$  1-Lipschitz

$X_1, \dots, X_n$  independently standard Gaussian

$$\mathbb{P}(F(X) \geq \mathbb{E}(F(X)) + t) \leq e^{-t^2/2}, \quad t \geq 0$$

$0 \leq X_i \leq 1$  independent,  $F$  1-Lipschitz and convex

$$\mathbb{P}(F(X) \geq \mathbb{E}(F(X)) + t) \leq 2e^{-t^2/4}, \quad t \geq 0$$

M. Talagrand (1995)

concentration inequalities

## numerous applications

- geometric functional analysis
- discrete and combinatorial probability
- empirical processes
- statistical mechanics
- random matrix theory

recent studies of

## random matrix and random growth models

new asymptotics

common, non central, rate  $(\text{mean})^{1/3}$

universal limiting **Tracy-Widom** distribution

random matrices, longest increasing subsequence,

random growth models, last passage percolation...

## random matrix models

Wigner matrix

$X^N = (X_{ij}^N)_{1 \leq i,j \leq N}$  symmetric  $N \times N$  matrix

$X_{ij}^N, i \leq j$ , independent identically distributed

$$\mathbb{E}(X_{ij}^N) = 0, \quad \mathbb{E}((X_{ij}^N)^2) = 1$$

eigenvalues  $\lambda_1^N \leq \dots \leq \lambda_N^N$  of  $X^N$

asymptotics of the eigenvalues as the size  $N \rightarrow \infty$

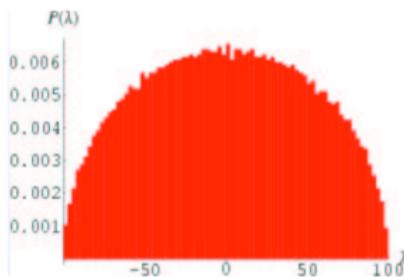
scaling  $\hat{X}^N = X^N / \sqrt{N}$

## Wigner's theorem (1955)

asymptotic behavior of the spectral measure  $(\hat{\lambda}_k^N = \lambda_k^N / \sqrt{N})$

$$\frac{1}{N} \sum_{k=1}^N \delta_{\hat{\lambda}_k^N} \rightarrow d\nu(x) = \frac{1}{2\pi} \sqrt{4 - x^2} dx \text{ on } (-2, +2)$$

## Wigner's semi-circle law



## Wigner's theorem

$$\frac{1}{N} \sum_{k=1}^N \delta_{\hat{\lambda}_k^N} \rightarrow d\nu(x) = \frac{1}{2\pi} \sqrt{4 - x^2} dx \quad \text{semi-circle law}$$

### global regime

large deviation asymptotics of the spectral measure

fluctuations of the spectral measure

$$\sum_{k=1}^N [f(\hat{\lambda}_k^N) - \int f d\nu] \rightarrow G \quad \text{Gaussian variable}$$

$f : \mathbb{R} \rightarrow \mathbb{R}$  smooth

## Wigner's theorem

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local regime

behavior of the individual eigenvalues

spacings (bulk behavior)

extremal eigenvalues (edge behavior)

## extremal eigenvalues

largest eigenvalue  $\lambda_N^N = \max_{1 \leq k \leq N} \lambda_k^N$

$$\hat{\lambda}_N^N = \lambda_N^N / \sqrt{N}$$

## Wigner's theorem

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$$\hat{\lambda}_N^N = \lambda_N^N / \sqrt{N} \rightarrow 2$$

fluctuations around 2

Gaussian Orthogonal or Unitary Ensemble (GOE, GUE)

$$N^{2/3} [\hat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}} \quad \text{C. Tracy, H. Widom (1994)}$$

$$\text{rate (mean)}^{1/3} : N^{1/6} [\lambda_N^N - 2\sqrt{N}]$$

GUE  $F_{\text{TW}}(s) = \exp \left( - \int_s^\infty (x-s) u(x)^2 dx \right), \quad s \in \mathbb{R}$

$$u'' = 2u^3 + xu \quad \text{Painlevé II equation} \quad (\text{similar for GOE})$$

## Gaussian Orthogonal or Unitary Ensemble (GOE, GUE)

independent entries  $X_{ij}^N$  real or complex Gaussian

$$P(dX) = \frac{1}{Z} \exp\left(-\text{Tr}(\beta X^2/4)\right) dX$$

$dX$  Lebesgue measure on  $N \times N$  matrices

orthogonal ( $\beta = 1$ ) or unitary conjugation ( $\beta = 2$ )

joint law of the eigenvalues  $(\lambda_1^N \leq \dots \leq \lambda_N^N)$  of  $X^N$

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$$P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{k=1}^N e^{-\beta x_k^2/4} dx_k$$

$$x = (x_1, \dots, x_N) \in \mathbb{R}^N$$

$$\Delta_N(x) = \prod_{k < \ell} (x_k - x_\ell) \quad \text{Vandermonde determinant}$$

$$\beta = 1 : \text{ real} \quad \beta = 2 : \text{ complex}$$

large deviation principles (Laplace methods)

fluctuations of the spectral measure

law of  $(\lambda_1^N, \dots, \lambda_N^N) :$   $P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{k=1}^N e^{-\beta x_k^2/4} dx_k$

fluctuations of extremal eigenvalues

$$N^{2/3} [\widehat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}} \quad \text{Tracy-Widom distribution}$$

orthogonal polynomial method  $(\beta = 2)$

marginals of  $(\lambda_1^N, \dots, \lambda_N^N)$  in terms of the kernel

$$K_N(x, y) = \sum_{\ell=0}^{N-1} P_\ell(x) P_\ell(y)$$

$P_\ell$  Hermite orthogonal polynomials for  $d\mu(x) = e^{-\beta x^2/4} \frac{dx}{Z}$

orthogonal polynomial asymptotics

## fluctuations of **extremal eigenvalues**

recent years : extensions to **(non Gaussian)** Wigner matrices

comparison with the **Gaussian case**

$N^{2/3}[\widehat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}}$  **Tracy-Widom** distribution

**A. Soshnikov (1999)** combinatorial moment method

$$\mathbb{E}\left(\sum_{k=1}^N (\lambda_k^N)^p\right) = \mathbb{E}\left(\text{Tr}((X^N)^p)\right), \quad p = O(N^{2/3})$$

**T. Tao, V. Vu (2009)** Lindeberg method

survey of recent approaches to  
**non asymptotic exponential inequalities**

quantify the limit theorems

spectral measure

extremal eigenvalues

**catch** the **new rate**  $(\text{mean})^{1/3}$

## two main questions and objectives

tail inequalities for the spectral measure

$$\mathbb{P}\left(\sum_{k=1}^N f(\hat{\lambda}_k^N) \geq t\right)$$

## Wigner's theorem

$$\frac{1}{N} \sum_{k=1}^N \delta_{\hat{\lambda}_k^N} \rightarrow d\nu(x) = \frac{1}{2\pi} \sqrt{4 - x^2} dx \quad \text{semi-circle law}$$

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$f : \mathbb{R} \rightarrow \mathbb{R}$  smooth

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tail inequalities for the extremal eigenvalues

$$\mathbb{P}(\hat{\lambda}_N^N \geq 2 + \varepsilon), \quad \mathbb{P}(\hat{\lambda}_N^N \leq 2 - \varepsilon)$$

## extremal eigenvalues

largest eigenvalue  $\lambda_N^N = \max_{1 \leq k \leq N} \lambda_k^N$

$$\widehat{\lambda}_N^N = \lambda_N^N / \sqrt{N} \rightarrow 2$$

fluctuations around 2

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Gaussian and more general Wigner matrices

measure concentration tool :  $F = F(X^N) = F(X_{ij}^N)$

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## tail inequalities for the spectral measure

A. Guionnet, O. Zeitouni (2000)

$f$  smooth measure concentration

$f : \mathbb{R} \rightarrow \mathbb{R}$  Lipschitz

$$F : X^N \rightarrow \text{Tr } f(\widehat{X}^N) = \sum_{k=1}^N f(\widehat{\lambda}_k^N) \quad \text{Lipschitz}$$

with respect to the Euclidean structure on  $N \times N$  matrices

convex if  $f$  is convex

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$$\mathbb{P}\left( \sum_{k=1}^N [f(\widehat{\lambda}_k^N) - \mathbb{E}(f(\widehat{\lambda}_i^N))] \geq t \right) \leq C e^{-t^2/C}, \quad t \geq 0$$

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## non Lipschitz functions $f$

typically  $f = \mathbf{1}_A$ ,  $A$  interval

$$\sum_{k=1}^N f(\hat{\lambda}_k^N) = \# \{ \hat{\lambda}_k^N \in A \} \quad \text{counting function}$$

suitable  $X^N = (X_{ij}^N)$ ,  $A$  interval in  $(-2, +2)$

$$\frac{1}{\sqrt{\log N}} \sum_{k=1}^N \left[ f(\hat{\lambda}_k^N) - \mathbb{E}(f(\hat{\lambda}_k^N)) \right] \rightarrow G \quad \text{Gaussian variable}$$

exponential tail inequalities (GUE)

$$\mathbb{P} \left( \sum_{k=1}^N \left[ f(\hat{\lambda}_k^N) - \mathbb{E}(f(\hat{\lambda}_k^N)) \right] \geq t \right) \leq C e^{-ct \log(1+t/\log N)}, \quad t \geq 0$$

non Gaussian Wigner matrices?

partial results (Stieltjes transform)

control of the Kolmogorov distance in Wigner's theorem

F. Götze, A. Tikhomirov

crucial tool in the recent solution (2009)

of the (bulk) spacings for Wigner matrices

L. Erdős, S. Péché, J. Ramirez, B. Schlein, H.-T. Yau

V. Vu, T. Tao

$$\mathbb{P}\left(\#\{\widehat{\lambda}_k^N \in A\} \geq t|A|N\right) \leq C e^{-t|A|N/C}, \quad t \geq C$$

Wigner's law at small scales  $|A| \sim \frac{1}{N}$

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tail inequalities for the extremal eigenvalues

$$\mathbb{P}(\hat{\lambda}_N^N \geq 2 + \varepsilon), \quad \mathbb{P}(\hat{\lambda}_N^N \leq 2 - \varepsilon)$$

Gaussian and more general Wigner matrices

measure concentration tool :  $F = F(X^N) = F(X_{ij}^N)$

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## tail inequalities for the extremal eigenvalues

fluctuations of the largest eigenvalue

$N^{2/3} [\hat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}}$  **Tracy-Widom** distribution

$\mathbb{P}(\hat{\lambda}_N^N \leq 2 + s N^{-2/3}) \rightarrow F_{\text{TW}}(s), \quad s \in \mathbb{R}$

**finite  $N$**  inequalities

at the (mean) $^{1/3}$  rate

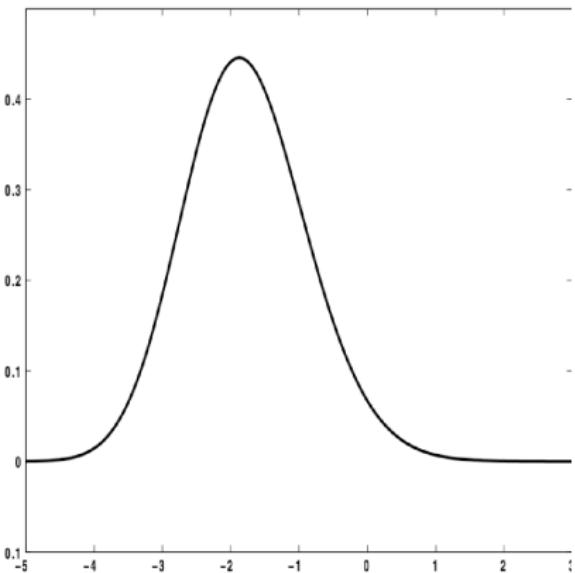
reflecting the tails of  $F_{\text{TW}}$

bounds on  $\text{Var}(\hat{\lambda}_N^N)$

$N^{2/3}[\widehat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}}$  **Tracy-Widom** distribution

GUE       $F_{\text{TW}}(s) = \exp \left( - \int_s^\infty (x-s) u(x)^2 dx \right), \quad s \in \mathbb{R}$

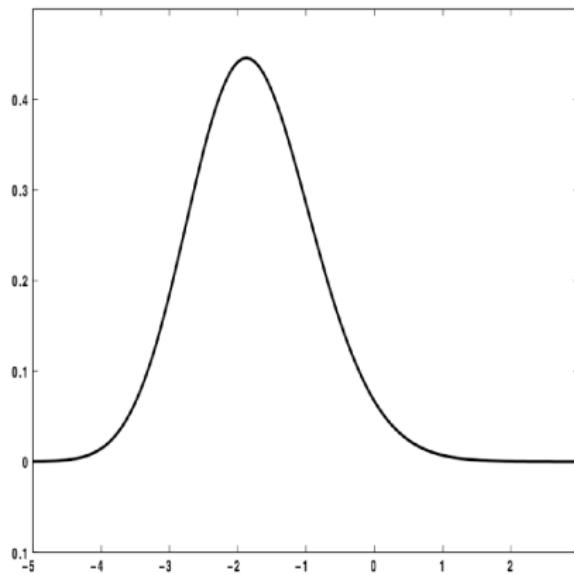
$u'' = 2u^3 + xu$       Painlevé II equation



mean  $\simeq -1.77$

$$F_{\text{TW}}(s) \sim e^{-s^3/12} \quad \text{as} \quad s \rightarrow -\infty$$

$$1 - F_{\text{TW}}(s) \sim e^{-4s^{3/2}/3} \quad \text{as} \quad s \rightarrow +\infty$$



density

(similar for GOE)

## non asymptotic small deviation inequalities

$$\mathbb{P}(\widehat{\lambda}_N^N \leq 2 + s N^{-2/3}) \rightarrow F_{\text{TW}}(s), \quad s \in \mathbb{R}$$

### expected bounds

$$\mathbb{P}(\widehat{\lambda}_N^N \geq 2 + \varepsilon)$$

## non asymptotic small deviation inequalities

$$\mathbb{P}(\widehat{\lambda}_N^N \leq 2 + s N^{-2/3}) \rightarrow F_{\text{TW}}(s), \quad s \in \mathbb{R}$$

$$1 - F_{\text{TW}}(s) \sim e^{-s^{3/2}/C} \quad (s \rightarrow +\infty)$$

## expected bounds

$$\mathbb{P}(\widehat{\lambda}_N^N \geq 2 + \varepsilon) \leq C e^{-N\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq 1$$

fit the **Tracy-Widom** asymptotics  $(\varepsilon = s N^{-2/3})$

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$$\mathbb{P}(\hat{\lambda}_N^N \leq 2 - \varepsilon) \leq C e^{-N^2\varepsilon^3/6}, \quad 0 < \varepsilon \leq 1$$

fit the **Tracy-Widom** asymptotics  $(\varepsilon = s N^{-2/3})$

## general concentration inequalities useless

Gaussian matrix

$$\lambda_N^N = \sup_{|v|=1} \langle X^N v, v \rangle$$

$\lambda_N^N$  Lipschitz of the Gaussian entries  $X_{ij}^N$

Gaussian concentration

$$\mathbb{P}(\hat{\lambda}_N^N \geq \mathbb{E}(\hat{\lambda}_N^N) + t) \leq e^{-Nt^2/C}, \quad t \geq 0$$

$$\mathbb{E}(\hat{\lambda}_N^N) \sim 2$$

**does not fit** the (small deviation) regime  $t = s N^{-2/3}$

correct large deviation bounds  $(t \geq 1)$

## several (specific) approaches

variety of techniques

to the non asymptotic inequalities

fragmentary results : geometric percolation model

orthogonal polynomial bounds, Riemann-Hilbert methods (GUE, GOE)

- moment recurrence equations (GUE, GOE, invariant ensembles)
- combinatorial moment method (Wigner, sample covariance matrices)
- tridiagonal matrices ( $\beta$ -ensembles)

## moment recurrence equations

Gaussian Unitary Ensemble (GUE)

Harer-Zagier (1986) (map enumeration)

$$a_p^N = \mathbb{E} \left( \text{Tr}((X^N)^{2p}) \right) = \mathbb{E} \left( \sum_{k=1}^N (\lambda_k^N)^{2p} \right), \quad p \in \mathbb{N}$$

three term recurrence equation

$$(p+1)a_p^N = (4p-2)Na_{p-1}^N + (p-1)(2p-1)(2p-3)a_{p-2}^N$$

$$a_p^N = E \left( \text{Tr}((X^N)^{2p}) \right) \leq C (4N)^p e^{Cp^3/N^2}, \quad p^3 \geq N^2$$

$$\mathbb{P}(\hat{\lambda}_N^N \geq 2 + \varepsilon) \leq C e^{-N\varepsilon^3/2/C}, \quad 0 < \varepsilon \leq 1$$

extends to various **unitary invariant models**

$$P(dX) = \frac{1}{Z} \exp\left(-\text{Tr}(v(X))\right) dX, \quad v : \mathbb{R} \rightarrow \mathbb{R}$$

$$d\mu = e^{-v} \frac{dx}{Z} \quad \text{classical orthogonal polynomials}$$

**continuous** : Hermite, Laguerre, Jacobi

**discrete** : Charlier, Meixner, Krawtchouk, Hahn

oriented percolation model

length of the longest increasing subsequence

**real** Gaussian Orthogonal Ensemble (GOE)

five term recurrence equation

## combinatorial moment method

Wigner matrices (proof of Wigner's theorem)

$$\mathbb{E} \left( \sum_{k=1}^N (\lambda_k^N)^p \right) = \mathbb{E} \left( \text{Tr}((X^N)^p) \right)$$

asymptotic results : **A. Soshnikov (1999)**

**O. Feldheim, S. Sodin (2009)**

non asymptotic moment inequalities

$$\mathbb{P}(\widehat{\lambda}_N^N \geq 2 + \varepsilon) \leq C e^{-N\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq 1$$

entries  $X_{ij}^N$  symmetric, subGaussian

extension to sample covariance matrices

## sample covariance matrices

multivariate statistical inference

principal component analysis

population  $(Y_1, \dots, Y_N)$

$Y_j$  vector in  $\mathbb{R}^M$  (characters)

sample covariance matrix  $Y^*Y$

$M \sim \kappa N, \quad N \rightarrow \infty$

(Gaussian) Wishart matrix models

## sample covariance matrices

$Y = Y^{M,N} = (Y_{ij})_{1 \leq i \leq M, 1 \leq j \leq N}$   $M \times N$  matrix ( $M \geq N$ )

$Y_{ij}$  independent identically distributed

$$\mathbb{E}(Y_{ij}) = 0, \quad \mathbb{E}((Y_{ij})^2) = 1$$

eigenvalues  $0 \leq \lambda_1^N \leq \dots \leq \lambda_N^N$  of  $Y^*Y$

asymptotic spectral measure  $M \sim \kappa N$ ,  $\kappa \geq 1$

$$\widehat{\lambda}_k^N = \lambda_k^N/N$$

$$\frac{1}{N} \sum_{k=1}^N \delta_{\widehat{\lambda}_k^N} \rightarrow \rho \quad \text{on} \quad ((\sqrt{\kappa} - 1)^2, (\sqrt{\kappa} + 1)^2)$$

$\rho$  **Marchenko-Pastur distribution**

## Tracy-Widom theorem for the largest eigenvalue

O. Feldheim, S. Sodin (2009)

non asymptotic moment inequalities

largest eigenvalue (symmetric, subGaussian entries)

$$\mathbb{P}(\hat{\lambda}_N^N \geq (\sqrt{\kappa} + 1)^2 + \varepsilon) \leq C e^{-N\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq 1$$

## sample covariance matrices

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$$\widehat{\lambda}_k^N = \lambda_k^N / N$$

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smallest eigenvalue

at the soft edge  $M \sim \kappa N, \quad \kappa > 1$

$$\mathbb{P}(\hat{\lambda}_1^N \leq (\sqrt{\kappa} - 1)^2 - \varepsilon) \leq C e^{-N\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq 1$$

## tridiagonal models ( $\beta$ -ensembles)

GOE, GUE, Wishart models

joint law of the eigenvalues  $(\lambda_1^N \leq \dots \leq \lambda_N^N)$  of  $X^N$

$$P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{k=1}^N e^{-\beta x_k^2/4} dx_k$$

$\beta = 1$  : GOE     $\beta = 2$  : GUE

## tridiagonal representation

$$\begin{pmatrix} g_1 & \chi_{N-1} & 0 & \cdots & \cdots & 0 \\ \chi_{N-1} & g_2 & \chi_{N-2} & 0 & \cdots & \vdots \\ 0 & \chi_{N-2} & g_3 & \chi_{N-3} & \ddots & \vdots \\ \vdots & 0 & \ddots & \ddots & \ddots & 0 \\ \vdots & \cdots & \ddots & \chi_2 & g_{N-1} & \chi_1 \\ 0 & \cdots & \cdots & 0 & \chi_1 & g_N \end{pmatrix}$$

$g_1, \dots, g_N$  independent standard normal

$\chi_{N-1}, \dots, \chi_1$  independent chi-variables

same GOE eigenvalues

H. Trotter (1984), A. Edelman, I. Dimitriu (2002)

## tridiagonal models ( $\beta$ -ensembles)

GOE, GUE, Wishart models

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## tridiagonal representation

$$\begin{pmatrix} g_1 & \chi_{(N-1)\beta} & 0 & \cdots & \cdots & 0 \\ \chi_{(N-1)\beta} & g_2 & \chi_{(N-2)\beta} & 0 & \cdots & \vdots \\ 0 & \chi_{(N-2)\beta} & g_3 & \chi_{(N-3)\beta} & \ddots & \vdots \\ \vdots & 0 & \ddots & \ddots & \ddots & 0 \\ \vdots & \cdots & \ddots & \chi_{2\beta} & g_{N-1} & \chi_\beta \\ 0 & \cdots & \cdots & 0 & \chi_\beta & g_N \end{pmatrix}$$

$g_1, \dots, g_N$  independent standard normal

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same GOE eigenvalues

H. Trotter (1984), A. Edelman, I. Dimitriu (2002)

similar for the  $\beta$ -ensembles , Wishart models

## A. Ramirez, B. Rider, B. Virág (2007)

new proof of **Tracy-Widom** theorem (every  $\beta$ )

new description of the **Tracy-Widom** distribution

$$TW_\beta = \sup_f \left\{ \frac{2}{\sqrt{\beta}} \int_0^\infty f^2(x) dB(x) - \int_0^\infty [f'(x)^2 + x f^2(x)] dx \right\}$$

$$f(0) = 0, \quad \int_0^\infty f^2(x) dx = 1$$

$$\int_0^\infty [f'(x)^2 + x f^2(x)] dx < \infty$$

Painlevé equation ?

**bounds on the largest eigenvalue**  $\lambda_N^N = \max_{1 \leq k \leq N} \lambda_k^N$

## tridiagonal representation

$$\begin{pmatrix} g_1 & \chi_{N-1} & 0 & \cdots & \cdots & 0 \\ \chi_{N-1} & g_2 & \chi_{N-2} & 0 & \cdots & \vdots \\ 0 & \chi_{N-2} & g_3 & \chi_{N-3} & \ddots & \vdots \\ \vdots & 0 & \ddots & \ddots & \ddots & 0 \\ \vdots & \cdots & \ddots & \chi_2 & g_{N-1} & \chi_1 \\ 0 & \cdots & \cdots & 0 & \chi_1 & g_N \end{pmatrix}$$

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**bounds on the largest eigenvalue**  $\lambda_N^N = \max_{1 \leq k \leq N} \lambda_k^N$

$$\lambda_N^N = \sup_{|v|=1} \left( \sum_{i=1}^N g_i v_i^2 + 2 \sum_{i=1}^{N-1} \chi_{N-i} v_i v_{i+1} \right)$$

explicit computations

$$\mathbb{P}(\hat{\lambda}_N^N \geq 2 + \varepsilon) \leq C e^{-N\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq 1$$

$$\mathbb{P}(\hat{\lambda}_N^N \leq 2 - \varepsilon) \leq C e^{-N^2\varepsilon^3/C}, \quad 0 < \varepsilon \leq 1$$

$\beta$ -ensembles (GOE, GUE, Wishart models)

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$\beta$ -ensembles (GOE, GUE, Wishart models)

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finite  $N$  variance bounds

$$\text{Var}(\hat{\lambda}_N^N) \leq C N^{-4/3}$$

$N^{2/3}[\hat{\lambda}_N^N - 2] \rightarrow F_{\text{TW}}$  Tracy-Widom distribution

$\beta$ -ensembles (GOE, GUE, Wishart models)

open for general Wigner matrices