

*Markov operators, classical orthogonal  
polynomial ensembles, and random matrices*

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recent study of

## random matrix and random growth models

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new asymptotics

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

universal limiting **Tracy-Widom** distribution

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random matrices, longest increasing subsequence,  
random growth models, last passage percolation...

**P. Forrester, C. Tracy, H. Widom, J. Baik, P. Deift, K. Johansson**

# invariant random matrix models

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$$P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{i=1}^N d\mu(x_i), \quad x = (x_1, \dots, x_N) \in \mathbb{R}^N$$

$$\Delta_N(x) = \prod_{i < j} (x_i - x_j), \quad d\mu(x) = e^{-v(x)} dx$$

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$\mu(\{x\}) = (1 - q)q^x$ ,  $x \in \mathbb{N}$  geometric distribution

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**large deviation asymptotics** (every  $\beta$ )

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minimizer of  $2 \int v d\nu - \beta \iint \log |x - y| d\nu(x) d\nu(y)$

weighted logarithmic potential theory

**local regime** : individual behavior (spacings, extreme values)

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**orthogonal polynomial ensemble**

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largest eigenvalue (particle) fluctuates at the rate  $(\text{mean})^{1/3}$

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C. Tracy, H. Widom, P. Deift, X. Zhou, J. Baik, K. Johansson,  
T. Kriecherbauer, K. McLaughlin, P. Miller, S. Venakides, M. Vanlessen,  
D. Gioev, P. Bleher, A. Its, A. Kuijlaars, M. Shcherbina, L. Pastur...

example : **Gaussian Unitary Ensemble (GUE)**

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

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

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- **spectral description** (universal arcsine law)

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- **non asymptotic tail inequalities for largest eigenvalues** (optimal rate)

## **classical** orthogonal polynomial ensembles

Hermite, Laguerre, Jacobi, Charlier, Meixner, Krawtchouk, Hahn

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$P_\ell, \ell \in \mathbb{N}$  Hermite polynomials for  $d\mu(x) = e^{-x^2/2} \frac{dx}{\sqrt{2\pi}}$

normalized in  $L^2(\mu)$

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$-LP_N = NP_N$  eigenvectors

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second order differential equation

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Φ Laplace transform of the **arcsine law**

$$\frac{dx}{\pi \sqrt{4 - x^2}} \quad \text{on } (-2, +2)$$

**common** behavior, with the limiting **arcsine law**

**for** the classical orthogonal polynomials

of (normalized) measures  $P_N^2 d\mu$

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varying parameters (in  $N$ )

compact case : **A. Maté, P. Nevai, V. Totik (1985)**

## spectral measure : averaging procedure

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GUE

$$E(\langle f, \widehat{\mu}^N \rangle) = \frac{1}{N} \sum_{\ell=0}^{N-1} \int f\left(\sqrt{\frac{\ell}{N}} \cdot \frac{x}{\sqrt{\ell}}\right) P_{\ell}^2 d\mu$$

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$$E(\langle f, \widehat{\mu}^N \rangle) = \frac{1}{N} \sum_{\ell=0}^{N-1} \int f\left(\sqrt{\frac{\ell}{N}} \cdot \frac{x}{\sqrt{\ell}}\right) P_{\ell}^2 d\mu$$

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Meixner example ( $\mu$  geometric) : equilibrium measure

example : last passage percolation  $W_N = \max_{\pi} \sum_{(i,j) \in \pi} w_{ij}$

$\mu$  geometric,  $K_N$  Meixner kernel,  $\widehat{\lambda}_i^N = \lambda_i^N / N$

spectral measure

$\widehat{\mu}^N = \frac{1}{N} \sum_{i=1}^N \delta_{\widehat{\lambda}_i^N} \rightarrow$  equilibrium measure on  $(a, b)$

largest "particle"  $\lambda_N^N \sim W_N / N \rightarrow b$

$N^{-1/3} [W_N - bN] \rightarrow F_{\text{TW}}$  Tracy-Widom distribution

K. Johansson (2000)

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$$\frac{1}{1-q} \left( \sqrt{qU(1+U)} \xi + [U + q(1+U)] \right)$$

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recurrence equation on moments

$$E\left(\operatorname{Tr}\left((X^N)^p\right)\right) = \int x^p \sum_{\ell=0}^{N-1} P_{\ell}^2 d\mu, \quad p \in \mathbb{N}$$

$$a_p^N = E\left(\text{Tr}\left((X^N)^{2p}\right)\right), \quad X^N \text{ GUE}$$

**three term** recurrence equation

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

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$\varepsilon_0(p)$  Catalan numbers

Markov operator technology

**similar recursion formulas**

for the **classical** orthogonal polynomial ensembles

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continuous variable : Hermite, Laguerre, Jacobi

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discrete variables : Charlier, Meixner, Krawtchouk, Hahn

explicit expressions for the (factorial) moments

# Gaussian Orthogonal Ensemble $\beta = 1$

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$$P^N(dx) = \frac{1}{Z} |\Delta_N(x)| \prod_{i=1}^N d\mu(x_i), \quad x = (x_1, \dots, x_N) \in \mathbb{R}^N$$

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limiting spectral distribution : **Wigner** semi-circle law

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$$\begin{aligned} \mu_{\text{GOE}}^N &= \sum_{\ell=0}^{N-1} P_\ell^2 + \frac{e^{x^2/4} P_{N-1}}{\int e^{x^2/4} P_{N-1} d\mu} \mathbf{1}_{N \text{ odd}} \\ &+ \sqrt{\frac{\pi N}{8}} e^{x^2/4} P_{N-1} \int \text{sgn}(x-y) e^{x^2/4}(y) P_N(y) d\mu(y) \end{aligned}$$

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## effective values

GUE

$$\begin{aligned}a_0^N &= N \\a_1^N &= N^2 \\a_2^N &= 2N^3 + N \\a_3^N &= 5N^4 + 10N^2 \\a_4^N &= 14N^5 + 70N^3 + 21N\end{aligned}$$

GOE

$$\begin{aligned}b_0^N &= N \\b_1^N &= N^2 + N \\b_2^N &= 2N^3 + 5N^2 + 5N \\b_3^N &= 5N^4 + 22N^3 + 52N^2 + 41N \\b_4^N &= 14N^5 + 93N^4 + 374N^3 + 690N^2 + 509N\end{aligned}$$

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duality with Symplectic Ensemble ( $\beta = 4$ )

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similar bounds for

classical orthogonal polynomial ensembles ( $\beta = 2$ )

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deviation inequality on last passage percolation  $W_N$

**K. Johansson (2000)**

moment comparison

**Wigner matrices** (independent entries)

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