

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

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recent study 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...

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

invariant random matrix models

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

$X = X^N$ real symmetric or Hermitian $N \times N$ matrix

$v(x) = \beta x^2/4$ Gaussian Orthogonal ($\beta = 1$) Unitary ($\beta = 2$) Ensembles

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

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

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

random growth models

directed last passage percolation

w_{ij} , $1 \leq i, j \leq N$, iid geometric random variables

$$W_N = \max_{\pi} \sum_{(i,j) \in \pi} w_{ij}$$

π up/right paths from $(1, 1)$ to (N, N)

K. Johansson (2000)

$$P(W_N \leq t) = \int_{\{\max x_i \leq t+N-1\}} \frac{1}{Z} \Delta_N(x)^2 \prod_{i=1}^N d\mu(x_i)$$

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

law of $(\lambda_1^N, \dots, \lambda_N^N)$: $P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{i=1}^N d\mu(x_i)$

global regime : spectral measure

spectral measure : $\widehat{\lambda}_i^N = \lambda_i^N$ suitably normalized

$$\widehat{\mu}^N = \frac{1}{N} \sum_{i=1}^N \delta_{\widehat{\lambda}_i^N} \rightarrow \text{equilibrium measure}$$

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

weighted logarithmic potential theory

law of $(\lambda_1^N, \dots, \lambda_N^N)$: $P^N(dx) = \frac{1}{Z} |\Delta_N(x)|^\beta \prod_{i=1}^N d\mu(x_i)$

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global regime : spectral measure

large deviation asymptotics (every β)

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global regime : spectral measure

$$P^N(dx) = \frac{1}{Z} \exp \left(- \sum_{i=1}^N v(x_i) + \beta \sum_{i < j} \log |x_i - x_j| \right) dx$$

spectral measure : $\hat{\lambda}_i^N = \lambda_i^N$ suitably normalized

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

weighted logarithmic potential theory

local regime : individual behavior (spacings, extreme values)

$\beta = 2$ **orthogonal polynomial method**

$P_\ell, \ell \in \mathbb{N}$ orthogonal polynomials for μ

marginals of P^N : determinants of the kernel

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

orthogonal polynomial ensemble

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$$\Delta_N(x) = C_n \det(P_{\ell-1}(x_k))_{1 \leq k, \ell \leq N}$$

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orthogonal polynomial ensemble

local regime : individual behavior (spacings, extreme values)

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P_ℓ , $\ell \in \mathbb{N}$ orthogonal polynomials for μ

$$\Delta_N(x)^2 = C_n^2 \det^2(P_{\ell-1}(x_k))_{1 \leq k, \ell \leq N}$$

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P^N determinantal random point field structure

marginals of P^N : determinants of the kernel

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

orthogonal polynomial ensemble

(common) **asymptotics of orthogonal polynomials**

universal local regime : spacings and edge behavior

top edge behavior

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

limiting **Tracy-Widom** distribution F_{TW}

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)

$$d\mu(x) = e^{-x^2/2} \frac{dx}{\sqrt{2\pi}}, \quad K_N \text{ Hermite kernel}, \quad \widehat{\lambda}_i^N = \lambda_i^N / \sqrt{N}$$

spectral measure

$$\widehat{\mu}^N = \frac{1}{N} \sum_{i=1}^N \delta_{\widehat{\lambda}_i^N} \rightarrow \frac{1}{2\pi} \sqrt{4-x^2} dx \quad \text{on } (-2, +2)$$

Wigner semi-circle law

$$N^{1/6} [\lambda_N^N - 2\sqrt{N}] \rightarrow F_{\text{TW}} \quad \text{Tracy-Widom distribution}$$

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

$$u'' = 2u^3 + xu \quad \text{Painlevé II equation}$$

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Wigner semi-circle law

largest eigenvalue λ_N^N

$N^{1/6} [\lambda_N^N - 2\sqrt{N}] \rightarrow F_{\text{TW}}$ **Tracy-Widom** distribution

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Wigner semi-circle law

largest eigenvalue $\lambda_N^N / \sqrt{N} \rightarrow 2$

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

μ 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)

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

K. Johansson (2000)

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largest "particle" $\lambda_N^N \sim W_N$

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K. Johansson (2000)

Markov operator tools

for **classical** orthogonal polynomial ensembles
of random matrix and random growth models

non asymptotic results

- **spectral description** (universal arcsine law)
- **recurrence equations for moments** (map enumeration)
- **recurrence equations for real symmetric models** (Gaussian Orthogonal Ensemble)
- **non asymptotic tail inequalities for largest eigenvalues** (optimal rate)

classical orthogonal polynomial ensembles

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

differential equations and operators

mean spectral measure $\mu^N = E\left(\frac{1}{N} \sum_{i=1}^N \delta_{\lambda_i^N}\right)$

$$\langle f, \mu^N \rangle = \int f(x) \frac{1}{N} K_N(x, x) d\mu(x) = \int f \frac{1}{N} \sum_{\ell=0}^{N-1} P_\ell^2 d\mu$$

example : Gaussian Unitary Ensemble (GUE)

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

normalized in $L^2(\mu)$

$$\langle f, \mu^N \rangle = \int f \frac{1}{N} \sum_{\ell=0}^{N-1} P_{\ell}^2 d\mu$$

preliminary investigation : measure $P_N^2 d\mu$

Laplace transform : $\varphi(t) = \int e^{tx} P_N^2 d\mu, \quad t \in \mathbb{R}$

Hermite operator : $Lf = f'' - x f'$

Gaussian integration by parts : $\int f(-Lg) d\mu = \int f' g' d\mu$

$-LP_N = NP_N$ eigenvectors

Laplace transform : $\varphi(t) = \int e^{tx} P_N^2 d\mu$

second order differential equation

$$t\varphi'' + \varphi' - t(t^2 + 4N + 2)\varphi = 0$$

under the GUE scaling : $t \mapsto t/\sqrt{N}$ ($\widehat{\lambda}_i^N = \lambda_i^N/\sqrt{N}$)

limiting differential equation

$$t\Phi'' + \Phi' - 4t\Phi = 0$$

Φ Laplace transform of the **arcsine law**

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

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

for the classical orthogonal polynomials

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

continuous variable : Hermite, Laguerre, Jacobi

discrete variable : Charlier, Meixner, Krawtchouk, Hahn

varying parameters (in N)

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

spectral measure : averaging procedure

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

$$\widehat{\mu}^N = \frac{1}{N} \sum_{i=1}^N \delta_{\widehat{\lambda}_i^N} \rightarrow \sqrt{U} \xi$$

U uniform, ξ arcsine, independent

$\sqrt{U} \xi$ semi-circle law on $(-2, +2)$

Meixner example (μ geometric) : equilibrium measure

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

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

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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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Meixner example (μ geometric) : equilibrium measure

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

moment recursion formula

$$E(e^{t \operatorname{Tr}(X^N)}) = \psi(t) = \int e^{tx} \sum_{\ell=0}^{N-1} P_{\ell}^2 d\mu, \quad t \in \mathbb{R}$$

differential equation

$$\text{GUE} \quad t\psi'' + 3\psi' - t(t^2 + 4N)\psi = 0$$

U. Haagerup, S. Thorbjørnsen (2003)

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

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

J. Harer, D. Zagier (1986)

map enumeration (Wick products)

$$a_p^N = \sum_{g \geq 0} \varepsilon_g(p) N^{p+1-2g}$$

genus series (oriented case)

$\varepsilon_0(p)$ Catalan numbers

Markov operator technology

similar recursion formulas

for the **classical** orthogonal polynomial ensembles

continuous variable : Hermite, Laguerre, Jacobi

discrete variables : Charlier, Meixner, Krawtchouk, Hahn

explicit expressions for the (factorial) moments

Gaussian Orthogonal Ensemble $\beta = 1$

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

$$d\mu(x) = e^{-x^2/2} \frac{dx}{\sqrt{2\pi}}$$

Pfaffian instead of determinant

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

Tracy-Widom edge asymptotics

moment recursion formula ?

mean spectral measure

$$E\left(\text{Tr}(f(X^N))\right) = \int f(x) \mu_{\text{GOE}}^N(x) d\mu(x)$$

$P_\ell, \ell \in \mathbb{N}$ Hermite polynomials

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

Markov operator tools : differential equation
on the Laplace transform of the spectral measure

moment recursion formula

$$b_p^N = E\left(\text{Tr}\left((X^N)^{2p}\right)\right), \quad p \in \mathbb{N}$$

recursion formula coupled with the moments a_p^N of the GUE

$$\begin{aligned} b_p^N &= (4N - 2)b_{p-1}^N + 4(2p - 2)(2p - 3)b_{p-2}^N \\ &\quad + a_p^N - (4N - 3)a_{p-1}^N - (2p - 2)(2p - 3)a_{p-2}^N \end{aligned}$$

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

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

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

recursion formula coupled with the moments a_p^N of the GUE

$$\begin{aligned} b_p^N &= (4N - 2)b_{p-1}^N + 4(2p - 2)(2p - 3)b_{p-2}^N \\ &\quad + a_p^N - (4N - 3)a_{p-1}^N - (2p - 2)(2p - 3)a_{p-2}^N \end{aligned}$$

b_p^N **five term** recurrence equation

closed form : **I. Goulden, D. Jackson (1997)**

map enumeration : unoriented case

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

recursion formulas lead to

sharp moment bounds

Gaussian Unitary Ensemble (GUE)

three term recurrence equation

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

similar bounds for

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

Gaussian Orthogonal Ensemble (GOE) ($\beta = 1$)

non asymptotic small deviation inequalities

$$P(\lambda_N^N \leq 2\sqrt{N} + sN^{-1/6}) \rightarrow F_{\text{TW}}(s), \quad s \in \mathbb{R}$$

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

moment bounds (GUE, GOE)

$$P(\lambda_N^N \geq 2\sqrt{N} + \varepsilon) \leq C e^{-N^{1/4}\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq \sqrt{N}$$

fits the **Tracy-Widom** asymptotics $(\varepsilon = sN^{-1/6})$

deviation inequality on last passage percolation W_N

K. Johansson (2000)

moment comparison

Wigner matrices (independent entries)

example : sign matrix $Y^N = Y_{ij}^N = \pm 1$

asymptotic result : **Y. Sinai, A. Soshnikov (1998-99)**

$$P(\lambda_N^N(Y) \leq 2\sqrt{N} + s N^{-1/6}) \rightarrow F_{\text{TW}}(s), \quad s \in \mathbb{R}$$

$$E\left(\text{Tr}((Y^N)^{2p})\right) \leq E\left(\text{Tr}((X^N)^{2p})\right), \quad X^N \text{ GOE}$$

$$P(\lambda_N^N(Y) \geq 2\sqrt{N} + \varepsilon) \leq C e^{-N^{1/4}\varepsilon^{3/2}/C}, \quad 0 < \varepsilon \leq \sqrt{N}$$