Corrections to scaled limits in random matrix theory

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- Riemann zero data and Painléve transcendents
- ► Finite size corrections to Riemann zero data
- ► Finite size corrections in RMT scaling limits







Gap probabilities and Painléve transcedents

Painléve transcendents — solutions of one of six nonlinear 2nd order differential equations, each with the property that no moveable singular points are essential singularities.

There is a (non-autonomous) Hamiltonian theory (Malmquist). The Hamiltonian satisfies a so-called Painléve equation in sigma form, e.g. σPII , $(\sigma'')^2 + 4\sigma' ((\sigma')^2 - t\sigma' + \sigma) - a^2 = 0$.

Since the work of the Kyoto school (1980) it has been known that **gap probabilities** in random matrix theory permit evaluations in terms of sigma Painléve transcendents, e.g. (F. & Witte, 2004) $p(s;0;\xi)|_{\beta=2} = \frac{\pi^2}{3}s^2 \exp \int_0^{2\pi s} u(t;\xi) \frac{dt}{t}$, where

$$\left(su''(s)\right)^{2} + \left(su'(s) - u(s)\right)\left(su'(s) - u(s) - 4 + 4(u'(s))^{2}\right) - 16\left(u'(s)\right)^{2} = 0$$

subject to the boundary condition

$$u(s;\xi) \underset{s \to 0^+}{\sim} -\frac{1}{15}s^2 + O(s^4) - \frac{\xi}{8640\pi} (s^5 + O(s^7)).$$

Riemann zero data

Montgomery-Odlyzko law: the statistics of the large **Riemann zeros** have the same distribution as the bulk eigenvalues of a large **complex Hermitian** random matrix.

Veracity can be probed using a high precision, big data set due to Odlyzko (\sim 2000). The data set begins with zero number $10^{23} + 985,531,550$, and lists the next 10^9 .

This occurs at the point s = 1/2 + iE in the complex *s*-plane with *E* equal to

 $13066434408793621120027.3961465854 \approx 1.30664344 \times 10^{22}.$



A grand challenge/ suggestion — underlying determinantal point process for RZ?

Keating and Snaith (2000) hypothesised a U(N) random matrix model for the leading corrections. Further developed by Bogomolny and collaborators.

The eigenvalues $e^{i\theta} = e^{i2\pi x/N}$ of U(N) matrices form a determinantal point process:

$$\rho_{(k)}(x_1, \dots, x_k) = \det \left[K_N(x_i, x_j) \right]_{i,j=1}^k, \quad K_N(x, y) = \frac{1}{N} \frac{\sin \pi (x - y)}{\sin \pi (x - y)/N}.$$

With $N = \frac{1}{\sqrt{12\Lambda}} \log \left(\frac{E}{2\pi} \right), \ \alpha = 1 + \frac{C}{\log(E/2\pi)}$, all results known to

date are consistent with

$$\rho_{(k)}^{\mathrm{RZ}}(x_1,\ldots,x_k) = \det \left[\mathcal{K}_N^{\mathrm{RZ}}(x_i,x_j) \right]_{i,j=1}^k$$

where

$$K_N^{\rm RZ}(x,y) = \frac{\sin \pi (x-y)}{\pi (x-y)} + \frac{\pi (x-y) \sin(\pi (2\alpha - 1)(x-y))}{6N^2} + O\Big(\frac{1}{N^4}\Big).$$

The implied RMT challenge: a theory of finite size corrections



Results from F. & Witte (2004) give

$$\frac{2\pi}{N}p^{N}(2\pi s/N;0;\xi) = \frac{1}{3}(N^{2}-1)\sin^{2}\frac{\pi s}{N}\exp\left(-\int_{0}^{\pi s/N}V(\cot\phi;\xi)\,d\phi\right)$$

where V satisfies a particular $\tilde{\sigma}$ PVI equation. Hence

$$\frac{2\pi}{N} p^{N} (2\pi s/N; 0; \xi) = \frac{\pi^{2} s^{2}}{3} \exp\left(\int_{0}^{2\pi s} \frac{u^{(0)}(X)}{X} dX\right) \\ \times \left(1 - \frac{1}{N^{2}} - \frac{\pi^{2} s^{2}}{3N^{2}} + \frac{1}{N^{2}} \int_{0}^{2\pi s} \frac{u^{(1)}(X)}{X} dX + O\left(\frac{1}{N^{4}}\right)\right)$$

A second order linear DE with σPV coefficients

We find that $u^{(1)}(X)$ satisfies the second order, linear differential equation

$$ilde{A}(s)y''(s)+ ilde{B}(s)y'(s)+ ilde{C}(s)y(s)= ilde{D}(s),$$

where, with $u(s) = u^{(0)}(s)$, $\tilde{A}(s) = 8s^2u''(s)$. The other coefficients are also explicit polynomials in $\{u(s), u'(s), u''(s), s\}$. The equation must be solved subject to the $s \to 0^+$ boundary condition

$$u^{(1)}(s) = rac{4}{15}s^2 - rac{13}{6300}s^4 + rac{\xi}{1728\pi}s^5 + O(s^6).$$

Universality results give that *all* random matrices in a certain class (for example complex Hermitian with independent entries from the same zero mean, unit variance distribution) have, after scaling, the same large N statistical properties. What about finite N corrections?

Setting up the question in the bulk

Option 1

- Choose a unitary invariant ensemble, or a complex Wigner ensemble. The simplest choice would be to consider GUE matrices.
- For complex Wigner ensembles the eigenvalue density will be to leading order given by the Wigner semi-circle law, supported to leading order on $(-\sqrt{2N}, \sqrt{2N})$. Use this density to **unfold** the eigenvalues.
- Now compute the averaged spacing distribution for some finite fraction of the eigenvalues about the origin.
- ► The task is to compute the large *N* form of this averaged spacing distribution.

Option 2

Instead of averaging over a finite fraction of the eigenvalues, ask specifically about the spacing between, say, the two middle eigenvalues.

Gap probabilities and correlations

Let E(n; J) denote the probability that the interval J contains exactly n eigenvalues. Define the generating function $G(J; \xi)$ by

$$G(J;\xi) := \sum_{n=0}^{\infty} (1-\xi)^n E(n;J).$$

Specifically, $G(J;\xi)|_{\xi=1} = E(0;J).$

In terms of the k-point correlations $\rho_{(k)}$,

$$G(J;\xi) = 1 + \sum_{k=1}^{\infty} \frac{(-\xi)^k}{k!} \int_{a_1}^{a_2} dx_1 \cdots \int_{a_1}^{a_2} dx_k \,\rho_{(k)}(x_1,\ldots,x_k)$$

Remarks

- ► The sums terminate at *N* for a finite system.
- Can interpret G(J; ξ) as the probability that the interval J is free of eigenvalues, in the setting that each eigenvalue has been deleted independently with probability (1 − ξ).

Spacing distributions and correlations

The PDF for the event that, given there is an eigenvalue at a_1 , the next eigenvalue to the right is at a_2 , is given by

$$p((a_1, a_2)) = -\frac{1}{\rho_{(1)}(a_1)} \frac{\partial^2}{\partial a_1 \partial a_2} E(0; (a_1, a_2))$$
$$= \frac{1}{\rho_{(1)}(a_1)} \rho_{(2)}(a_1, a_2) + \cdots$$

Note that this is relevant to Option 1 via the averaged quantity

$$ar{p}(s) := \int p((a,a+s)) \, da = \int rac{1}{
ho_{(1)}(a)}
ho_{(2)}(a,a+s) \, da + \cdots$$

Difficult. Ask instead a related question: what are the properties of the finite size correction at the soft edge (i.e. neighbourhood of the largest eigenvalue).

Soft edge leading corrections

We have

$$p_{\max}(s) = \frac{d}{ds} \Big(\sum_{k=1}^{\infty} \frac{(-1)^k}{k!} \int_s^{\infty} dx_1 \cdots \int_s^{\infty} dx_k \rho_{(k)}(x_1, \ldots, x_k)$$
$$= \rho_{(1)}(s) - \int_s^{\infty} \rho_{(2)}(s, x) dx + \cdots$$

Introduce the scaled variable

$$s_{N,t} = \begin{cases} \sqrt{2N} + t/\sqrt{2}N^{1/6}, & \text{GUE} \\ 4N + 2a + 2(2N)^{1/3}t, & \text{LUE} \end{cases}$$

F. & Trinh (2017) have shown that

$$p_{\max,N}(t;\xi) = p_{\max,\infty}(t;\xi) + \frac{1}{N^{2/3}}\tilde{p}(t;\xi) + O\left(\frac{1}{N}\right).$$

We know that, with $q'' = sq + 2q^3$,

$$egin{aligned} p_{\max,\infty}(t;\xi) &= rac{d}{dt} \det(\mathbb{I} - \xi \mathbb{K}^{ ext{Airy}}_{(t,\infty)}) \ &= rac{d}{dt} \exp\Big(- \int_t^\infty (x-t) q^2(x;\xi) \, dx \Big) \end{aligned}$$

Weak universality?

The function $\tilde{p}(t; \xi)$ can be characterised as the solution of a 2nd order linear differential equation with σ PII coefficients. It is *different* for the GUE and LUE.

Note subtlety in relation to the Laguerre case $s_{N,t} = 4N + 2(2N)^{1/3}t + 2a$. What would happen if instead we set $s_{N,t} = 4N + 2(2N)^{1/3}t$?

$$p_{\max,N}(t+\frac{c}{N^{1/3}};\xi)=p_{\max,\infty}(t;\xi)+\frac{c}{N^{1/3}}\frac{d}{dt}p_{\max,\infty}(t;\xi)+O\Big(\frac{1}{N^{2/3}}\Big).$$

Hence, a **non-optimal** choice of scaling can be detected by the "leading" correction term being related to the scaled limit as a derivative.

Question: Can a soft edge scaling variable always be chosen so that the optimal correction term is $O\left(\frac{1}{N^{2/3}}\right)$?

A particular Wigner ensemble

Define X so that its entries are chosen independently and uniformly from the set of four values $\frac{1}{\sqrt{2}}(\pm 1 \pm i)$.

In terms of X define the Hermitian matrix $Y = \frac{1}{2}(X + X^{\dagger})$. Off diagonal entries have mean zero and variance $\frac{1}{2}$. Hence to leading order largest eigenvalue is equal to $\sqrt{2N}$.

Plot histograms associated with the scaled random variable $t = \sqrt{2}N^{1/6}(\lambda_{\max} - \sqrt{2N})$, minus $p_{\max,\infty}(t)$ multiplied by $N^{1/3}$ (left). Except for shifting by c = 1/2, this appears to be the graph for $\frac{d}{dt}p_{\max,\infty}(t)$. Replacing t by $t - 1/(2N^{1/3})$ leads to a $N^{-2/3}$ correction.



Future work

Study optimal soft edge scaling for the density in Gaussian and Laguerre beta ensemble with beta even. This is possible due to certain duality formulas, e.g.

$$\left\langle \prod_{l=1}^{N} \left(x - \sqrt{\frac{2}{\beta}} x_l \right)^n \right\rangle_{\mathsf{GE}_{\beta,N}} = \left\langle \prod_{j=1}^{n} (x - i x_j)^N \right\rangle_{\mathsf{GE}_{4/\beta,n}}$$

Find in the Gaussian case that choosing

$$\lambda = \sqrt{2N} + \frac{1}{\sqrt{2}N^{1/6}} \left(x + \left(\frac{1}{2} - \frac{1}{\beta}\right) \frac{1}{N^{1/3}} \right)$$

gives optimal O($N^{-2/3}$) correction. For the Laguerre case, the same (a dependent) scaling as for $\beta = 2$ is optimal.

For the unitary invariant ensemble with potential $e^{-Nx^{2m}}$ a result of Dieft and Gioev establishes *m* dependent values of *b* and γ so that $\lambda = b + \frac{x}{N^{2/3}\gamma}$ gives optimal O($N^{-2/3}$) correction.

Presently studying similar questions at the hard edge.