

Propagation of singular behavior in UE and GUE sums

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with

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**Random matrix theory and strongly correlated
systems**

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1. Unitary Ensemble plus GUE

Unitary ensemble (UE)

M is random $n \times n$ Hermitian matrix from UE

$$\frac{1}{Z_n} e^{-n \operatorname{Tr} V(M)} dM$$

- Eigenvalues have joint p.d.f.

$$\frac{1}{Z_n} \Delta_n(x)^2 \prod_{j=1}^n e^{-nV(x_j)}, \quad \Delta_n(x) = \prod_{1 \leq i < j \leq n} (x_j - x_i)$$

- Equilibrium measure μ_V is the minimizer of

$$\iint \log \frac{1}{|s-t|} d\mu(s) d\mu(t) + \int V(t) d\mu(t)$$

- μ_V is limiting eigenvalue distribution as $n \rightarrow \infty$

H is $n \times n$ (scaled) GUE matrix with distribution

$$\frac{1}{Z_n} e^{-\frac{n}{2} \text{Tr} H^2} dH$$

- Eigenvalues of $\sqrt{\tau}H$ follow semi-circle law with variance τ

$$d\lambda_\tau(s) = \frac{1}{2\pi\tau} \sqrt{4\tau - s^2} ds, \quad s \in [-2\sqrt{\tau}, 2\sqrt{\tau}].$$

Sum of UE and GUE

We are interested in eigenvalues of

$$X = M + \sqrt{\tau}H$$

with M from a unitary ensemble, H from a (scaled) GUE, independent from M , and $\tau > 0$.

- Eigenvalues are determinantal point process
- Interpretation as non-intersecting paths
- Free probability:

$$\mu_V \boxplus \lambda_\tau$$

is limiting distribution of eigenvalues as $n \rightarrow \infty$

Determinantal point process

Brézin and Hikami (1998), Zinn-Justin (1998), Johansson (2001)

- If M is fixed with eigenvalues a_1, \dots, a_n , then eigenvalues of $M + \sqrt{\tau}H$ have joint density

$$\propto \frac{1}{\Delta_n(a)} \cdot \Delta_n(x) \cdot \det \left[e^{-\frac{n}{2\tau}(x_k - a_j)^2} \right]_{j,k=1}^n$$

- If M is random from UE , then after averaging over a_1, \dots, a_n ,

$$\propto \Delta_n(x) \cdot \det \left[\int_{-\infty}^{\infty} a^{j-1} e^{-\frac{n\tau}{2}(x_k - a)^2} e^{-nV(a)} da \right]_{j,k=1}^n$$

- It is polynomial ensemble (special case of DPP)

Non-intersecting paths

Johansson (2001), Bleher and Kuijlaars (2004):

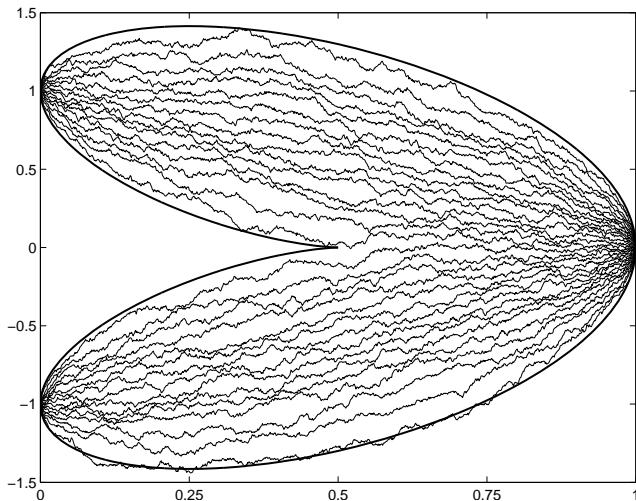
- Non-intersecting **Brownian bridges** with starting positions a_1, \dots, a_n at time $t = 0$ and ending positions b_1, \dots, b_n at time $t = T$
- Joint density for particles at time $t \in (0, T)$:

$$\propto \frac{1}{\Delta_n(a)\Delta_n(b)} \cdot \det \left[e^{-\frac{n}{2t}(a_j - x_k)^2} \right]_{j,k=1}^n \cdot \det \left[e^{-\frac{n}{2(T-t)}(x_j - b_k)^2} \right]_{j,k=1}^n$$

- In limit when all $b_k \rightarrow 0$

$$\propto \frac{1}{\Delta_n(a)} \cdot \det \left[e^{-\frac{n}{2t}(a_j - x_k)^2} \right]_{j,k=1}^n \cdot \Delta_n(x) \cdot \prod_{j=1}^n e^{-\frac{n}{2(T-t)}x_j^2}$$

Figure



Picture if all $a_j \rightarrow \pm 1$.

Random starting points

- If a_j are random eigenvalues of matrix from UE ensemble $\frac{1}{Z_n} e^{-nV(M)} dM$ then, after averaging over a_1, \dots, a_n ,

$$\propto \Delta_n(x) \cdot \det \left[\int_{-\infty}^{\infty} a^{j-1} e^{-\frac{n}{2t}(x_k - a)^2} e^{-nV(a)} da \right]_{j,k=1}^n \cdot \prod_{j=1}^n e^{-\frac{n}{2(T-t)} x_j^2}$$

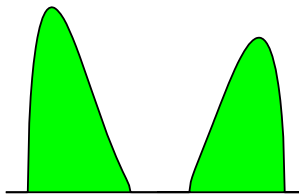
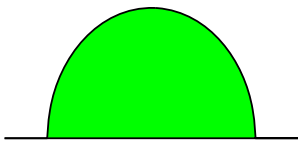
- For $T \rightarrow \infty$, this is **exactly the same** as eigenvalues of $M + \sqrt{t}H$.

2. Singular potential

Typical behavior of equilibrium measure

Suppose V is **real analytic**.

- μ_V is supported on finite union of intervals with density ψ_V
- Generically ψ_V has **square root vanishing** at endpoints, and is positive in the interior of each of the intervals.



Singular behavior

Singular interior point

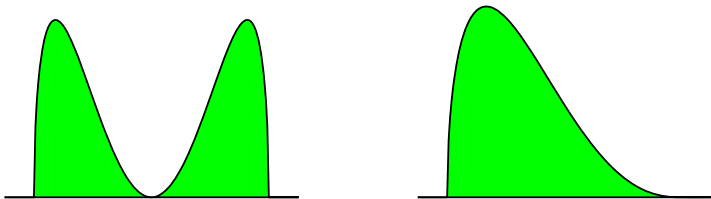
- ψ_V vanishes at interior point x^*

$$\psi_V(x) \sim (x - x^*)^\kappa, \quad \kappa = 2k \text{ for integer } k \geq 1.$$

Singular edge point

- ψ_V vanishes to higher order at edge point x^*

$$\psi_V(x) \sim |x - x^*|^\kappa, \quad \kappa = 2k + \frac{1}{2} \text{ for integer } k \geq 1.$$



Correlation kernels

Limiting correlation kernels

- **Sine kernel** at regular interior point
- **Airy kernel** at regular edge point
- **Painlevé kernels** at singular points

3. Results

- (a) Propagation of singular density for $\tau < \tau_{cr}$
- (b) Propagation of correlation kernel
- (c) Density at critical τ_{cr}

4 Propagation of singular density

Propagation of singular density

Suppose μ_V has density

$$\psi_V(s) = c_0^{\kappa+1} |s - x^*|^\kappa h(s), \quad h(x^*) = 1,$$

where h is real analytic at x^* . Define

$$\tau < \tau_{cr} = \left[\int \frac{d\mu_V(s)}{(s - x^*)^2} \right]^{-1} \quad \text{and} \quad x_\tau^* = x^* - \tau \int \frac{d\mu_V(s)}{s - x^*}$$

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Theorem

$\mu_\tau = \mu_V \boxplus \lambda_\tau$ has density ψ_τ satisfying

$$\psi_\tau(s) = c_\tau^{\kappa+1} |s - x_\tau^*|^\kappa h_\tau(s), \quad c_\tau = \frac{\tau_{cr}}{\tau_{cr} - \tau} c_0, \quad h_\tau(x^*) = 1$$

where h_τ is real analytic at x_τ^*

Propagation of singular density

The singularity at an interior point or edge point propagates in the model $M + \sqrt{\tau}H$.

- Interpretation in terms of non-intersecting Brownian paths
- Connection with two-matrix model

Duits (2014)

About the proof

Biane (1997) describes how to calculate the density of

$$\mu_\tau = \mu_V \boxplus \lambda_\tau$$

The mapping

$$w \mapsto w + \tau \int \frac{d\mu_V(s)}{w - s}$$

has an inverse $z \mapsto F_\tau(z)$ that extends continuously and injectively to $\overline{\mathbb{C}^+}$. Then

$$\psi_\tau(x) = -\frac{1}{\pi} \operatorname{Im} \left(\int \frac{d\mu_V(s)}{F_\tau(x) - s} \right), \quad x \in \mathbb{R}.$$

About the proof

$$\psi_\tau(x) = -\frac{1}{\pi} \operatorname{Im} \left(\int \frac{d\mu_V(s)}{F_\tau(x) - s} \right), \quad x \in \mathbb{R}.$$

- **Expand** $w + \tau \int \frac{d\mu_V(s)}{w-s}$ **around** $w = x^*$
- **Expand its inverse** $F_\tau(z)$ **around** $z = x_\tau^*$
- **Combine this to find first non-zero term in expansion of** $\psi_\tau(x)$ **around** $x = x_\tau^*$.

5 Propagation of correlation kernel

Propagation of correlation kernel

- M is from **unitary ensemble** with eigenvalue correlation kernel $K_n^M(x, y)$ and scaling limit

$$\lim_{n \rightarrow \infty} \frac{1}{c_0 n^\gamma} K_n^M \left(x^* + \frac{x}{c_0 n^\gamma}, x^* + \frac{y}{c_0 n^\gamma} \right) = \mathcal{K}_{crit, \kappa}(x, y)$$

with $\gamma = (\kappa + 1)^{-1}$

- $X = M + \sqrt{\tau}H$ has correlation kernel $K_n^X(x, y)$

Propagation of correlation kernel

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- $X = M + \sqrt{\tau}H$ has correlation kernel $K_n^X(x, y)$

Theorem

Under these conditions

$$\lim_{n \rightarrow \infty} \frac{e^{-H_n(x) + H_n(y)}}{c_\tau n^\gamma} K_n^X \left(x_\tau^* + \frac{x}{c_\tau n^\gamma}, x_\tau^* + \frac{y}{c_\tau n^\gamma} \right) = \mathcal{K}_{\text{crit}, \kappa}(x, y)$$

for certain function H_n

About the proof

For $X = M + \sqrt{\tau}H$,

$$K_n^X(x, y) = \frac{n}{2\pi i\tau} \int_{x^*-i\infty}^{x^*+i\infty} ds \int_{-\infty}^{\infty} dt K_n^M(s, t) \\ \times e^{\frac{n}{2}(V(s)-V(t))} e^{\frac{n}{2\tau}((s-x)^2-(t-y)^2)}$$

Claeys-K-Wang (2015)

- Proof is basically a **steepest descent analysis** of this double integral.

Separate

$$K_n^X(x, y) = K_{n,loc}^X(x, y) + K_{n,rest}^X(x, y)$$

with

$$K_{n,loc}^X(x, y) = \frac{n}{2\pi i\tau} \int_{x^* - iRn^{-\gamma}}^{x^* + iRn^{-\gamma}} ds \int_{x^* - Rn^{-\gamma}}^{x^* + Rn^{-\gamma}} dt K_n^M(s, t) \\ \times e^{\frac{n}{2}(V(s) - V(t))} e^{\frac{n}{2\tau}((s-x)^2 - (t-y)^2)}$$

- **R is large, but fixed constant, independent of n , but it will depend on x and y .**

Analysis of local part

After change of variables

$$\begin{aligned} & \frac{1}{c_T n^\gamma} K_{n,loc}^X \left(x_\tau^* + \frac{x}{c_T n^\gamma}, x_\tau^* + \frac{y}{c_T n^\gamma} \right) \\ &= \frac{n^{1-2\gamma}}{2\pi i c_0 c_T} \int_{x^*-i c_0 R}^{x^*+i c_0 R} ds \int_{x^*-c_0 R}^{x^*+c_0 R} dt e^{F_n(s,x)-F_n(t,y)} \\ & \quad \times \underbrace{\frac{1}{c_0 n^\gamma} K_n^M \left(x^* + \frac{s}{c_0 n^\gamma}, x^* + \frac{t}{c_0 n^\gamma} \right)}_{\rightarrow \mathcal{K}_{crit,\kappa}(s,t)} \end{aligned}$$

with

$$F_n(s,x) = \frac{n}{2} V \left(x^* + \frac{s}{c_0 n^\gamma} \right) + \frac{n}{2\tau} \left(\frac{s}{c_0 n^\gamma} - \frac{\tau}{2} V'(x^*) - \frac{x}{c_T n^\gamma} \right)$$

Taylor expansion of F_n

Expand as $n \rightarrow \infty$

$$F_n(s, x) = \frac{n}{2} V(x^*) + \frac{\tau n}{8} V'(x^*)^2 + \frac{n^{1-\gamma}}{2c_\tau} V'(x^*)x + \frac{n^{1-2\gamma}}{4c_0 c_\tau} V''(x^*)x^2 \\ + \frac{n^{1-2\gamma}}{2\tau c_0 c_\tau} [(s-x)^2 + O(n^{-\gamma})]$$

- No term $sn^{-\gamma}$ because of choice for x_τ^*
- Complete square $(s-x)^2$ because of choice for c_τ
- Saddle point equation $\frac{\partial F_n}{\partial s} = 0$ gives the saddle

$$x + O(n^{-\gamma})$$

Analysis of rest

$$K_{n,rest}^X(x, y) = K_n^X(x, y) - K_{n,loc}^X(x, y)$$

Scaled version

$$e^{-H_n(x)+H_n(y)} K_{n,rest}^X \left(x_\tau^* + \frac{x}{c_\tau n^\gamma}, x_\tau^* + \frac{y}{c_\tau n^\gamma} \right)$$

becomes $O(e^{-cn^{1-\gamma}})$ as $n \rightarrow \infty$ if R is large enough.

- **Proof uses deformation of t -contour into the complex plane, and uses bounds on orthogonal polynomials that come from **steepest descent analysis of Riemann-Hilbert problem****

6 Density at critical T_{cr}

Density at critical τ_{cr}

What about the density for critical $\tau_{cr} = \left[\frac{d\mu_V(s)}{(s - x^*)^2} \right]^{-1}$?

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We have results for

Singular interior point $\psi_V(s) \sim (s - x^*)^{2k}$ as $s \rightarrow x^*$

- $\kappa = 2k = 2$

- $\kappa = 2k \geq 4$ and $\int \frac{d\mu_V(s)}{(s - x^*)^3} = 0$

- $\kappa = 2k \geq 4$ and $\int \frac{d\mu_V(s)}{(s - x^*)^3} \neq 0$

Singular edge point $\psi_V(s) \sim (x^* - s)^{2k + \frac{1}{2}}$ as $s \rightarrow x^* -$

Density at critical τ_{cr}

Singular interior point with exponent $\kappa = 2k = 2$

Theorem

$$\psi_{\tau_{cr}}(s) = A_{\pm} |s - x_{\tau_{cr}}^*|^{1/2} + O(s - x_{\tau_{cr}}^*), \quad \text{as } s \rightarrow x_{\tau_{cr}}^* \pm$$

where

$$\begin{pmatrix} A_- \\ A_+ \end{pmatrix} = \frac{1}{(\pi \tau_{cr} c_0)^{3/2} (1 + (\frac{1}{\pi} \int \frac{h(s)}{s - x^*} ds)^2)^{1/4}} \begin{pmatrix} \cos \theta \\ \sin \theta \end{pmatrix}$$

$$\theta = \frac{\pi}{4} + \arctan \left(\frac{1}{\pi} \int \frac{h(s)}{s - x^*} ds \right)$$

Density at critical τ_{cr}

Singular interior point with exponent $\kappa = 2k \geq 4$.

Theorem

Suppose $\int \frac{d\mu_V(s)}{(s - x^*)^3} ds = 0$. **Then**

$$\psi_{\tau_{cr}}(s) = A |s - x_{\tau_{cr}}^*|^{1/3} + O(|s - x_{\tau_{cr}}^*|^{2/3}), \quad \text{as } s \rightarrow x_{\tau_{cr}}^*$$

where

$$A = \frac{\sqrt{3}}{2\pi\tau_{cr}^{4/3} \left(\int \frac{d\mu_V(s)}{(s - x^*)^4} ds \right)^{1/3}}$$

Density at critical τ_{cr}

Singular interior point with exponent $\kappa = 2k \geq 4$.

Theorem

Suppose $\int \frac{d\mu_V(s)}{(s-x^*)^3} ds > 0$. **Then**

$$\psi_{\tau_{cr}}(s) = \begin{cases} A |s - x_{\tau_{cr}}^*|^{1/2} + O(s - x_{\tau_{cr}}^*), & \text{as } s \rightarrow x_{\tau_{cr}}^* + \\ B |s - x_{\tau_{cr}}^*|^{k-1/2} + O((s - x_{\tau_{cr}}^*)^k) & \text{as } s \rightarrow x_{\tau_{cr}}^* - \end{cases}$$

where

$$A = \frac{1}{\pi \tau_{cr}^{3/2} \left(\int \frac{d\mu_V(s)}{(s-x^*)^3} ds \right)^{1/2}}, \quad B = \frac{c_0^{2k+1}}{2 \tau_{cr}^{k+1/2} \left(\int \frac{d\mu_V(s)}{(s-x^*)^3} ds \right)^{k+1/2}}$$

Density at critical τ_{cr}

Singular right edge point with exponent $\kappa = 2k + \frac{1}{2}$.

Theorem

Suppose $\int \frac{d\mu_V(s)}{(x^* - s)^3} ds > 0$. **Then**

$$\psi_{\tau_{cr}}(s) = A (x_{\tau_{cr}}^* - s)^{1/2} + O((s - x_{\tau_{cr}}^*)^{3/4}) \quad \text{as } s \rightarrow x_{\tau_{cr}}^* -$$

where

$$A = \frac{1}{\tau_{cr}^{3/2} c_0^{1/4} \left(\int \frac{d\mu_V(s)}{(s - x^*)^3} ds \right)^{1/2}}$$

Speculations about correlation kernel

We have no results yet for the correlation kernels at critical τ_{cr}

- Case $\kappa = 2k = 2$ leads to exponent $1/2$:
Tacnode kernel of **Duits-Geudens (2013)** !?

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Pearcey kernel !??

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- Case $\kappa = 2k \geq 4$ with $\int \frac{d\mu_V(s)}{(s - x^*)^3} \leq 0$ leads to two exponents $1/2$ and $k - 1/2$:
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- Case $\kappa = 2k + 1/2$ leads to exponent $1/2$:
Airy kernel ??

Thank you for your attention