

The correction term for the maximum of the log characteristic polynomial of CUE

Elliot Paquette

Weizmann Institute of Science (Rehovot, Israel)

Heilbronn Institute
10 May 2016

The CUE field

1. For $|z| \leq 1$,

$$\mathbf{U}(z) := \mathbf{U}_N(z) = \sum_{h=1}^N \log |1 - ze^{i\theta_h}|,$$

where $\{e^{i\theta_h}\}_{h=1}^N$ are the eigenvalues of an $N \times N$ Haar unitary matrix.

The CUE field

1. For $|z| \leq 1$,

$$\mathbf{U}(z) := \mathbf{U}_N(z) = \sum_{h=1}^N \log |1 - ze^{i\theta_h}|,$$

where $\{e^{i\theta_h}\}_{h=1}^N$ are the eigenvalues of an $N \times N$ Haar unitary matrix.

2. Maximum principle:

$$\sup_{|z| < 1} \mathbf{U}(z) = \sup_{|z|=1} \mathbf{U}(z),$$

CUE Theorems

Theorem (Arguin–Belius–Bourgade: 1511.07399)

The law of large numbers:

$$\frac{\max_{|z|=1} \mathbf{U}_N(z)}{\log N} \xrightarrow[N \rightarrow \infty]{\mathbb{P}} \mathbf{1},$$

CUE Theorems

Theorem (Arguin–Belius–Bourgade: 1511.07399)

The law of large numbers:

$$\frac{\max_{|z|=1} \mathbf{U}_N(z)}{\log N} \xrightarrow[N \rightarrow \infty]{\mathbb{P}} 1,$$

Theorem (P'–Zeitouni: 1602.08875)

The correction term:

$$\frac{\max_{|z|=1} \mathbf{U}_N(z) - \log N}{\log \log N} \xrightarrow[N \rightarrow \infty]{\mathbb{P}} -\frac{3}{4},$$

The Gaussian field

Let \mathcal{W} be

$$\mathcal{W}(\omega) = \text{w-lim}_{N \rightarrow \infty} \sum_{h=-N}^N \sqrt{h} \cdot Z_h \omega^h,$$

i.i.d. complex normals $\{Z_h\}_{h=1}^{\infty}$, $\mathbb{E}Z_h^2 = 0$, $\mathbb{E}|Z_h|^2 = 1$, $Z_{-h} = \overline{Z_h}$.

The Gaussian field

Let \mathcal{W} be

$$\mathcal{W}(\omega) = \text{w-lim}_{N \rightarrow \infty} \sum_{h=-N}^N \sqrt{h} \cdot Z_h \omega^h,$$

i.i.d. complex normals $\{Z_h\}_{h=1}^{\infty}$, $\mathbb{E}Z_h^2 = 0$, $\mathbb{E}|Z_h|^2 = 1$, $Z_{-h} = \overline{Z_h}$.

Define

$$\mathbf{G}(z) = \frac{1}{2\pi} \int_0^{2\pi} \log |1 - ze^{i\theta}| \cdot \mathcal{W}(e^{i\theta}) d\theta.$$

The Gaussian field

Let \mathcal{W} be

$$\mathcal{W}(\omega) = \text{w-lim}_{N \rightarrow \infty} \sum_{h=-N}^N \sqrt{h} \cdot Z_h \omega^h,$$

i.i.d. complex normals $\{Z_h\}_{h=1}^{\infty}$, $\mathbb{E}Z_h^2 = 0$, $\mathbb{E}|Z_h|^2 = 1$, $Z_{-h} = \overline{Z_h}$.

Define

$$\mathbf{G}(z) = \frac{1}{2\pi} \int_0^{2\pi} \log |1 - ze^{i\theta}| \cdot \mathcal{W}(e^{i\theta}) d\theta.$$

Then \mathbf{G} is a random Gaussian harmonic function on $\{z : |z| < 1\}$ with $\mathbf{G}(0) = 0$ almost surely and covariance

$$\mathbb{E}\mathbf{G}(z)\mathbf{G}(y) = -\frac{\log |1 - zy|}{2}.$$

U and G

Fix $\delta > 0$. **Fact:** considered as a random continuous function on $|z| < (1 - \delta)$,

$$\mathbf{U}_N(z) \xrightarrow[N \rightarrow \infty]{} \mathbf{G}(z)$$

(e.g. Strong Szegő theorem).

U and G

Fix $\delta > 0$. **Fact:** considered as a random continuous function on $|z| < (1 - \delta)$,

$$\mathbf{U}_N(z) \xrightarrow[N \rightarrow \infty]{} \mathbf{G}(z)$$

(e.g. Strong Szegő theorem).

Fact:

$$\max_{|z|=1-N^{-1}} \mathbf{G}(z) - \log N + \frac{3}{4} \log \log N \xrightarrow[N \rightarrow \infty]{} \xi,$$

for some distribution ξ (e.g. using Ding–Roy–Zeitouni '15+some analysis)

U and G

Fix $\delta > 0$. **Fact:** considered as a random continuous function on $|z| < (1 - \delta)$,

$$\mathbf{U}_N(z) \xrightarrow[N \rightarrow \infty]{} \mathbf{G}(z)$$

(e.g. Strong Szegő theorem).

Fact:

$$\max_{|z|=1-N^{-1}} \mathbf{G}(z) - \log N + \frac{3}{4} \log \log N \xrightarrow[N \rightarrow \infty]{} \xi,$$

for some distribution ξ (e.g. using Ding–Roy–Zeitouni '15+some analysis)

Hope: if \mathbf{U}_N and \mathbf{G} were *close enough* for $|z| \leq 1 - N^{-1}$, then their maximums would have the same order.

Plan

1. Branching structures.

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.

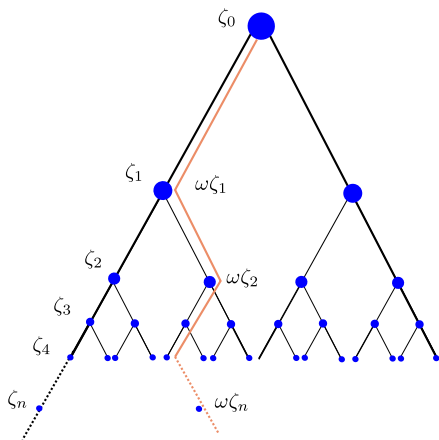
Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.

Plan

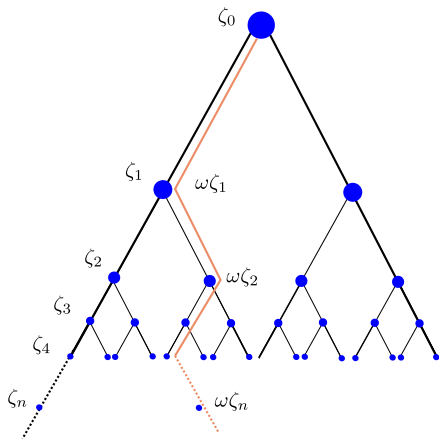
1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Branching random walk



1. $\{Z_e\}_e$ are i.i.d. $N(0, \frac{1}{2})$.

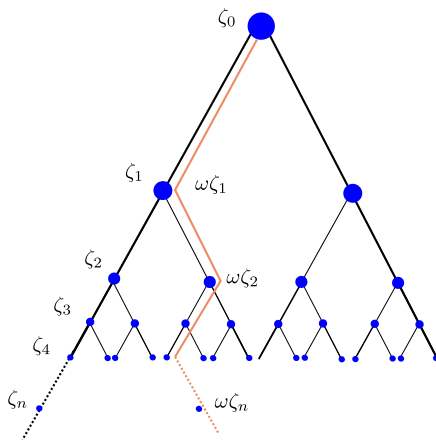
Branching random walk



1. $\{Z_e\}_e$ are i.i.d. $N(0, \frac{1}{2})$.
2. For every ray ω ,

$$X(\omega\zeta_n) = \sum_{k=1}^n Z_{(\omega\zeta_{k-1}, \omega\zeta_k)}.$$

Branching random walk

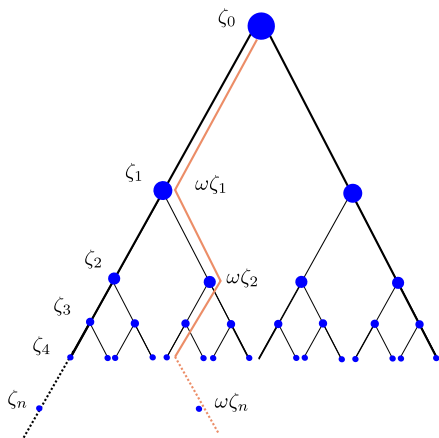


1. $\{Z_e\}_e$ are i.i.d. $N(0, \frac{1}{2})$.
2. For every ray ω ,

$$X(\omega\zeta_n) = \sum_{k=1}^n Z_{(\omega\zeta_{k-1}, \omega\zeta_k)}.$$

3. Every pair $\{(X(\omega_1\zeta_k))_1^\infty, (X(\omega_2\zeta_k))_1^\infty\}$ are Gaussian random walks, equal for the first $(\omega_1, \omega_2)\zeta_0$ steps and independent after.

Branching random walk



3. Every pair $\{(X(\omega_1\zeta_k))_1^\infty, (X(\omega_2\zeta_k))_1^\infty\}$ are Gaussian random walks, equal for the first $(\omega_1, \omega_2)_{\zeta_0}$ steps and independent after.

4. The *barrier method* uses this pair structure to show tightness for

$$\max_{\omega} X(\omega\zeta_n) - m_n$$

with

$$m_n = c_1 n + c_2 \log n.$$

Hyperbolic metric

1. Define $d_{\mathbb{H}}(0, z) = \log \left(\frac{1+|z|}{1-|z|} \right)$.

Hyperbolic metric

1. Define $d_{\mathbb{H}}(0, z) = \log \left(\frac{1+|z|}{1-|z|} \right)$.
2. $d_{\mathbb{H}}$ is invariant under all the maps:

$$T_y : \mathbb{D} \rightarrow \mathbb{D}, \quad z \mapsto \frac{z + y}{1 + z\bar{y}}.$$

Hyperbolic metric

1. Define $d_{\mathbb{H}}(0, z) = \log \left(\frac{1+|z|}{1-|z|} \right)$.
2. $d_{\mathbb{H}}$ is invariant under all the maps:

$$T_y : \mathbb{D} \rightarrow \mathbb{D}, \quad z \mapsto \frac{z + y}{1 + z\bar{y}}.$$

3. The process $z \mapsto \mathbf{G}(T_y(z)) - \mathbf{G}(y)$ has the same distribution as $z \mapsto \mathbf{G}(z)$.

Hyperbolic metric

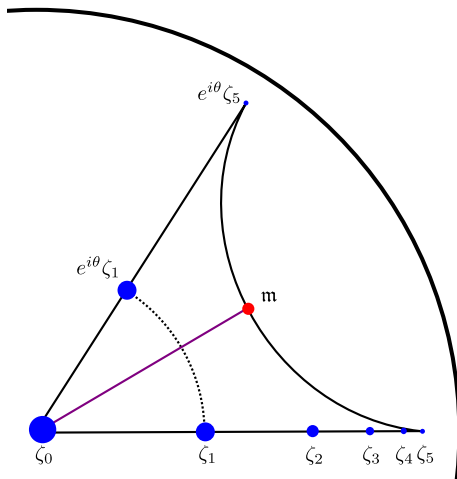
1. Define $d_{\mathbb{H}}(0, z) = \log \left(\frac{1+|z|}{1-|z|} \right)$.
2. $d_{\mathbb{H}}$ is invariant under all the maps:

$$T_y : \mathbb{D} \rightarrow \mathbb{D}, \quad z \mapsto \frac{z + y}{1 + z\bar{y}}.$$

3. The process $z \mapsto \mathbf{G}(T_y(z)) - \mathbf{G}(y)$ has the same distribution as $z \mapsto \mathbf{G}(z)$.
- 4.

$$\begin{aligned} \text{Var}(\mathbf{G}(z) - \mathbf{G}(y)) &= \log \left(\cosh \left(\frac{d_{\mathbb{H}}(z, y)}{2} \right) \right) \\ &= \frac{d_{\mathbb{H}}(z, y)}{2} - \log 2 + O(e^{-d_{\mathbb{H}}(z, y)}), \end{aligned}$$

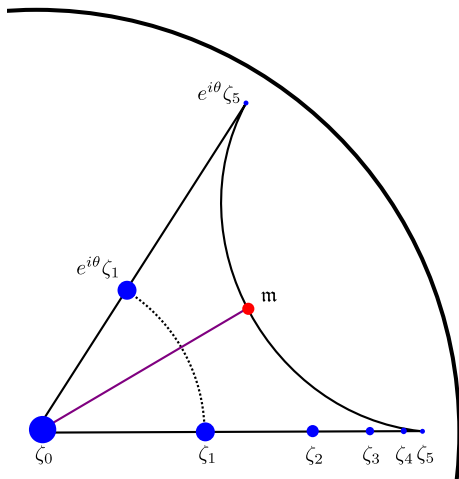
Hyperbolic branching



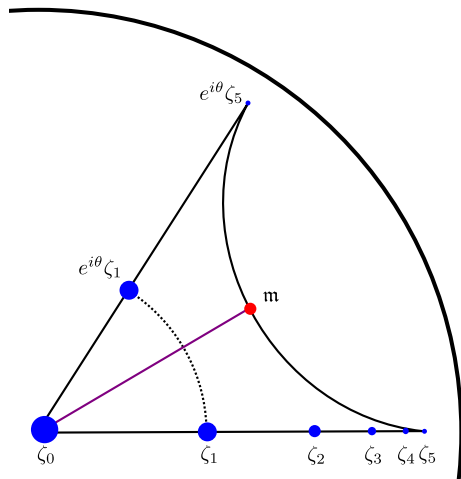
Hyperbolic branching

1. Let $\{\zeta_i\}_1^\infty \in \mathbb{R}_+$ have

$$d_{\mathbb{H}}(\zeta_i, \zeta_j) = |i - j|.$$



Hyperbolic branching

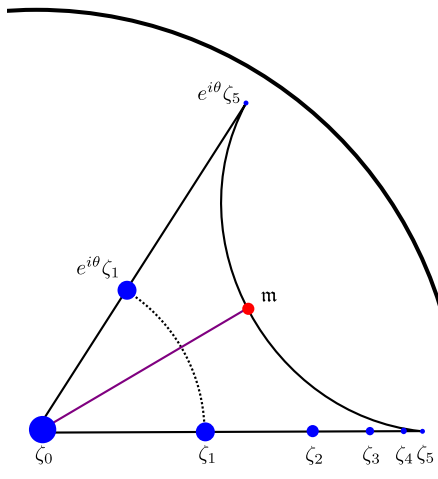


1. Let $\{\zeta_i\}_1^\infty \in \mathbb{R}_+$ have

$$d_{\mathbb{H}}(\zeta_i, \zeta_j) = |i - j|.$$

2. $m(\theta)$ is the midpoint of the hyperbolic segment $[1, e^{i\theta}]$.

Hyperbolic branching



1. Let $\{\zeta_i\}_1^\infty \in \mathbb{R}_+$ have

$$d_{\mathbb{H}}(\zeta_i, \zeta_j) = |i - j|.$$

2. $m(\theta)$ is the midpoint of the hyperbolic segment $[1, e^{i\theta}]$.

3. The pair

$$\{(\mathbf{G}(\zeta_k))_1^\infty, (\mathbf{G}(e^{i\theta}\zeta_k))_1^\infty\}$$

are Gaussian random walks, *roughly* coupled for the first $d_{\mathbb{H}}(0, m) \approx (1, e^{i\theta})_{\zeta_0}$ steps and roughly independent after.

Plan

1. Branching structures.
2. **The barrier method for \mathbf{G} .**
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Bounding the maximum of \mathbf{G}

1. Let $n = \lceil \log N \rceil$, so that *ideally* $\mathbf{U}_N \approx \mathbf{G}$ for z with $d_{\mathbb{H}}(0, z) \leq n$.

Bounding the maximum of \mathbf{G}

1. Let $n = \lceil \log N \rceil$, so that *ideally* $\mathbf{U}_N \approx \mathbf{G}$ for z with $d_{\mathbb{H}}(0, z) \leq n$.
2. Let $\mathbb{T}_n \subset \mathbb{D}$ be

$$\left\{ e^{2\pi i j [e^{-n}]} : 1 \leq j \leq e^n \right\}$$

Bounding the maximum of \mathbf{G}

1. Let $n = \lceil \log N \rceil$, so that *ideally* $\mathbf{U}_N \approx \mathbf{G}$ for z with $d_{\mathbb{H}}(0, z) \leq n$.
2. Let $\mathbb{T}_n \subset \mathbb{D}$ be

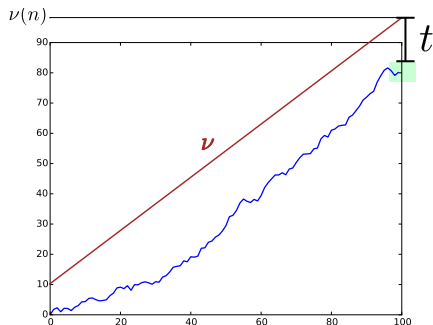
$$\left\{ e^{2\pi i j [e^{-n}]} : 1 \leq j \leq e^n \right\}$$

3. We will sketch the argument for

$$\frac{\max_{\omega \in \mathbb{T}_n} \mathbf{G}(\omega \zeta_n) - n}{\log n} \xrightarrow{\mathbb{P}} -\frac{3}{4}$$

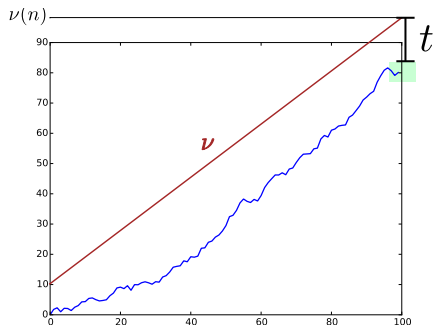
using the barrier method.

Bounding the maximum of \mathbf{G}



1. Let $\eta_m \rightarrow \infty$ grow sublogarithmically.

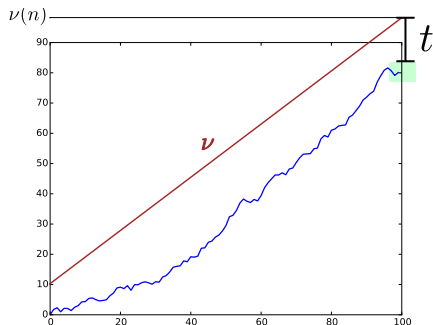
Bounding the maximum of \mathbf{G}



1. Let $\eta_m \rightarrow \infty$ grow sublogarithmically.
2. Define

$$\nu(i) = \frac{i}{n} \left(n - \frac{3}{4} \log n \right) + \eta_m.$$

Bounding the maximum of \mathbf{G}



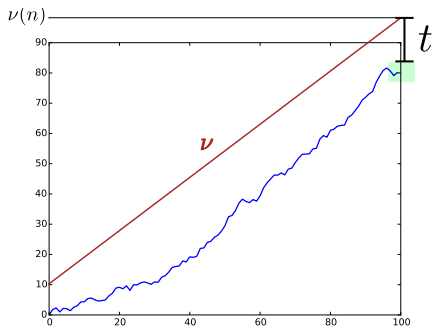
1. Let $\eta_m \rightarrow \infty$ grow sublogarithmically.
2. Define

$$\nu(i) = \frac{i}{n} \left(n - \frac{3}{4} \log n \right) + \eta_m.$$

3. Define $\mathcal{B}(\omega)$:

$$\{ \mathbf{G}(\omega_{\zeta_k}) \leq \nu(k), \forall 1 \leq k \leq n \}$$

Bounding the maximum of \mathbf{G}



2. Define

$$\nu(i) = \frac{i}{n} \left(n - \frac{3}{4} \log n \right) + \eta_n.$$

3. Define $\mathcal{B}(\omega)$:

$$\{ \mathbf{G}(\omega \zeta_k) \leq \nu(k), \forall 1 \leq k \leq n \}$$

4. Define $\mathcal{E}(\omega, t)$:

$$\{ \mathbf{G}(\omega \zeta_n) - \nu(n) + t \in [0, -1] \}.$$

Biassing

Because X is Gaussian,

$$\mathbb{E} \left[f(\left(\mathbf{G}(\zeta_k)\right)_1^n) e^{\alpha \mathbf{G}(\zeta_n)} \right] = \mathbb{E} \left[f\left(\left(\mathbf{G}(\zeta_k) + \frac{\alpha}{2} \mu(k)\right)_1^n\right) \right] \mathbb{E} \left[e^{\alpha \mathbf{G}(\zeta_n)} \right].$$

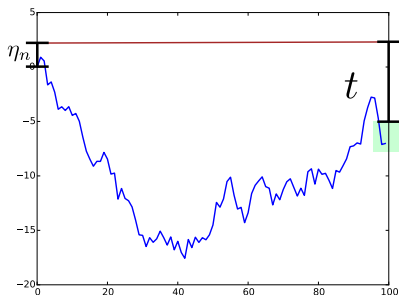
Biasing

Because X is Gaussian,

$$\mathbb{E} \left[f((\mathbf{G}(\zeta_k))_1^n) e^{\alpha \mathbf{G}(\zeta_n)} \right] = \mathbb{E} \left[f((\mathbf{G}(\zeta_k) + \frac{\alpha}{2} \mu(k))_1^n) \right] \mathbb{E} [e^{\alpha \mathbf{G}(\zeta_n)}].$$

$$\mu(k) = \mathbb{E} \mathbf{G}(\zeta_k) \mathbf{G}(\zeta_n) \approx k.$$

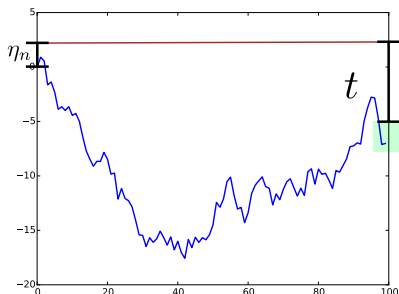
Biasing X



1. With $\alpha = 2 - \frac{3 \log n}{2n}$

$$\begin{aligned} & \mathbb{E} \left[\mathbf{1}_{\{\mathcal{B} \cap \mathcal{E}\}} e^{\alpha \mathbf{G}(\zeta_n)} \right] \\ &= \mathbb{E} \left[\mathbf{1}_{\{\mathcal{B}^F \cap \mathcal{E}^F\}} \right] \mathbb{E} \left[e^{\alpha \mathbf{G}(\zeta_n)} \right]. \end{aligned}$$

Biasing X



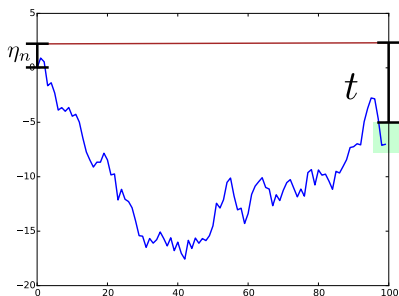
1. With $\alpha = 2 - \frac{3 \log n}{2n}$

$$\begin{aligned} & \mathbb{E} \left[\mathbf{1}_{\{\mathcal{B} \cap \mathcal{E}\}} e^{\alpha \mathbf{G}(\zeta_n)} \right] \\ &= \mathbb{E} \left[\mathbf{1}_{\{\mathcal{B}^F \cap \mathcal{E}^F\}} \right] \mathbb{E} \left[e^{\alpha \mathbf{G}(\zeta_n)} \right]. \end{aligned}$$

2. Here, $\mathcal{B}^F(\omega)$ is:

$$\left\{ \mathbf{G}(\omega \zeta_k) \leq \eta_n, \forall 1 \leq k \leq n \right\}.$$

Biasing X



1. With $\alpha = 2 - \frac{3 \log n}{2n}$

$$\begin{aligned} & \mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E} \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \\ &= \mathbb{E} \left[\mathbf{1} \{ \mathcal{B}^F \cap \mathcal{E}^F \} \right] \mathbb{E} \left[e^{\alpha \mathbf{G}(\zeta_n)} \right]. \end{aligned}$$

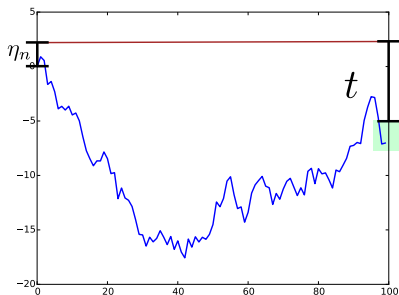
2. Here, $\mathcal{B}^F(\omega)$ is:

$$\{ \mathbf{G}(\omega \zeta_k) \leq \eta_m, \forall 1 \leq k \leq n \}.$$

3. And, $\mathcal{E}^F(\omega, t)$ is:

$$\{ \mathbf{G}(\omega \zeta_n) - \eta_m + t \in [0, -1] \}.$$

Biasing X



2. Here, $\mathcal{B}^F(\omega)$ is:

$$\{\mathbf{G}(\omega\zeta_k) \leq \eta_n, \forall 1 \leq k \leq n\}.$$

3. And, $\mathcal{E}^F(\omega, t)$ is:

$$\{\mathbf{G}(\omega\zeta_n) - \eta_n + t \in [0, -1]\}.$$

4. *Ballot theorem*: for $0 \leq t \leq \log n$,

$$\mathbb{P}(\mathcal{B}^F \cap \mathcal{E}^F(t)) = \Theta\left(\frac{\eta_n^2}{n^{3/2}}\right).$$

Upper bound

1. On the one hand,

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \asymp \mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} \right] e^{\alpha(n - 0.75 \log n + \eta_n - t)}$$

Upper bound

1. On the one hand,

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \asymp \mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} \right] e^{\alpha(n - 0.75 \log n + \eta_n - t)}$$

2. On the other hand, using **tilting** and the **ballot theorem**,

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \asymp \frac{\eta_n^2}{n^{1.5}} e^{0.25n\alpha^2}.$$

Upper bound

1. On the one hand,

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \asymp \mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} \right] e^{\alpha(n - 0.75 \log n + \eta_n - t)}$$

2. On the other hand, using **tilting** and the **ballot theorem**,

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B} \cap \mathcal{E}(t) \} e^{\alpha \mathbf{G}(\zeta_n)} \right] \asymp \frac{\eta_n^2}{n^{1.5}} e^{0.25n\alpha^2}.$$

3. Putting it together

$$\mathbb{E} \left[\mathbf{1} \{ \mathcal{B}(\omega) \cap \mathcal{E}(\omega, t) \} \right] \asymp \eta_n^2 e^{-n - 2\eta_n + 2t}$$

4. Formally, any choice of $\alpha = 2 + o(n^{-\epsilon})$ gives the same answer here.

Upper bound

3. Putting it together

$$\mathbb{E} [\mathbf{1} \{ \mathcal{B}(\omega) \cap \mathcal{E}(\omega, t) \}] \asymp \eta_m^2 e^{-n-2\eta_m+2t}$$

4. Formally, any choice of $\alpha = 2 + o(n^{-\epsilon})$ gives the same answer here.

Upper bound

3. Putting it together

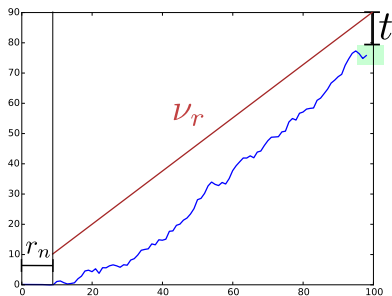
$$\mathbb{E}[\mathbf{1}\{\mathcal{B}(\omega) \cap \mathcal{E}(\omega, t)\}] \asymp \eta_n^2 e^{-n-2\eta_n+2t}$$

5. With $\eta_n = 0.75 \log n$, and $0 < t \leq 0.75 \log n - C \log \log n$ and ruling out the complement of the barrier gives

$$\max_{\omega \in \mathbb{T}_n} \mathbf{G}(\omega \zeta_n) \leq n - 0.75 \log n + C \log \log n$$

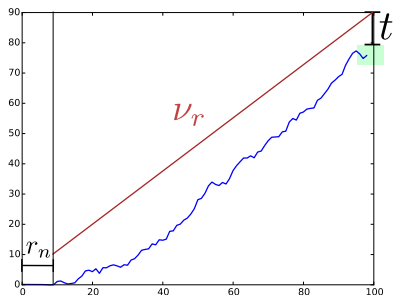
with $\mathbb{P} \rightarrow 1$.

Lower bound



1. Let $r_n \rightarrow \infty$ be another sublogarithmic sequence.

Lower bound

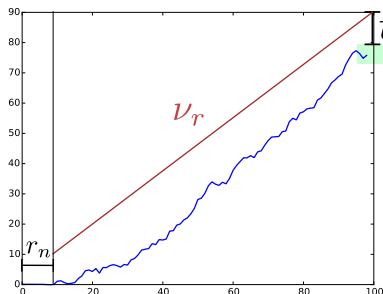


1. Let $r_n \rightarrow \infty$ be another sublogarithmic sequence.
2. Zero out the first r_n levels:

$$\nu_r(i) = \frac{i - r_n}{n} \left(n - \frac{3}{4} \log n \right) + \eta_n.$$

$$\mathbf{G}_r(\omega \zeta_k) = \mathbf{G}(\omega \zeta_k) - \mathbf{G}(\omega \zeta_{k \wedge r})$$

Lower bound



1. Let $r_n \rightarrow \infty$ be another sublogarithmic sequence.
2. Zero out the first r_n levels:

$$\nu_r(i) = \frac{i - r_n}{n} \left(n - \frac{3}{4} \log n \right) + \eta_n.$$

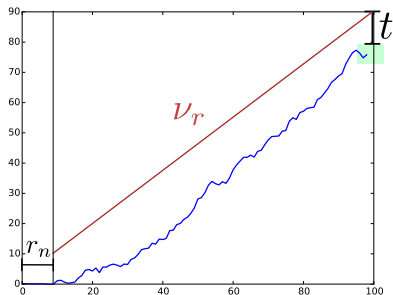
$$\mathbf{G}_r(\omega \zeta_k) = \mathbf{G}(\omega \zeta_k) - \mathbf{G}(\omega \zeta_{k \wedge r})$$

3. Redefine $\mathcal{B}(\omega)$:

$$\{ \mathbf{G}_r(\omega \zeta_k) \leq \nu_r(k), \forall r_n \leq k \leq n \}$$

Lower bound

2. Zero out the first r_n levels:



$$\nu_r(i) = \frac{i - r_n}{n} \left(n - \frac{3}{4} \log n \right) + \eta_n.$$

$$\mathbf{G}_r(\omega \zeta_k) = \mathbf{G}(\omega \zeta_k) - \mathbf{G}(\omega \zeta_{k \wedge r})$$

3. Redefine $\mathcal{B}(\omega)$:

$$\{ \mathbf{G}_r(\omega \zeta_k) \leq \nu_r(k), \forall r_n \leq k \leq n \}$$

4. Redefine $\mathcal{E}(\omega, t)$:

$$\{ \mathbf{G}_r(\omega \zeta_n) - \nu_r(n) + t \in [0, -1] \}.$$

Second moment setup

1. Define

$$\mathfrak{B}_\omega(\mathbf{F}) = 2\mathbf{F}(\omega\zeta_n), \quad Y(\omega) = e^{\mathfrak{B}_\omega(\mathbf{G}_r)} \mathbf{1}\{\mathcal{B}(\omega) \cap \mathcal{E}(\omega)\}.$$

Second moment setup

1. Define

$$\mathfrak{B}_\omega(\mathbf{F}) = 2\mathbf{F}(\omega\zeta_n), \quad Y(\omega) = e^{\mathfrak{B}_\omega(\mathbf{G}_r)} \mathbf{1}\{\mathcal{B}(\omega) \cap \mathcal{E}(\omega)\}.$$

2. In terms of this indicator, define the biased counting variable

$$Z = \sum_{\omega \in \mathbb{T}_n} Y(\omega).$$

Second moment setup

1. Define

$$\mathfrak{B}_\omega(\mathbf{F}) = 2\mathbf{F}(\omega\zeta_n), \quad Y(\omega) = e^{\mathfrak{B}_\omega(\mathbf{G}_r)} \mathbf{1}\{\mathcal{B}(\omega) \cap \mathcal{E}(\omega)\}.$$

2. In terms of this indicator, define the biased counting variable

$$Z = \sum_{\omega \in \mathbb{T}_n} Y(\omega).$$

3. We then have

$$\mathbb{E}[\mathbf{1}\{Z > 0\}] \geq \frac{(\sum_{\theta} \mathbb{E}[Y(\omega)])^2}{\sum_{\omega_1, \omega_2} \mathbb{E}[Y(\omega_1)Y(\omega_2)]},$$

Key estimates

A For a $1 - o(1)$ fraction of the pairs (ω_1, ω_2) ,

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \leq \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] (1 + o(1)).$$

Key estimates

A For a $1 - o(1)$ fraction of the pairs (ω_1, ω_2) ,

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \leq \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] (1 + o(1)).$$

B For (ω_1, ω_2) very close, we also need an estimate that is *not too bad*:

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \lesssim \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] |\omega_1 - \omega_2|^{-1}$$

Key estimates

A For a $1 - o(1)$ fraction of the pairs (ω_1, ω_2) ,

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \leq \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] (1 + o(1)).$$

B For (ω_1, ω_2) very close, we also need an estimate that is *not too bad*:

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \lesssim \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] |\omega_1 - \omega_2|^{-1}$$

1. By biasing,

$$\frac{\mathbb{E} [Y(\omega_1)Y(\omega_2)]}{\mathbb{E} [e^{\mathfrak{B}_{\omega_1} + \mathfrak{B}_{\omega_2}}]} = \mathbb{E} [(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)]$$

Key estimates

1. By biasing,

$$\frac{\mathbb{E}[Y(\omega_1)Y(\omega_2)]}{\mathbb{E}[e^{\mathfrak{B}_{\omega_1} + \mathfrak{B}_{\omega_2}}]} = \mathbb{E}[(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)]$$

2. If the midpoint $\mathfrak{m}(\omega_1, \omega_2)$ is closer to ζ_0 than r_n , then all the correlations between ray ω_1 and ω_2 are small.

Key estimates

1. By biasing,

$$\frac{\mathbb{E}[Y(\omega_1)Y(\omega_2)]}{\mathbb{E}[e^{\mathfrak{B}_{\omega_1} + \mathfrak{B}_{\omega_2}}]} = \mathbb{E}[(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)]$$

2. If the midpoint $\mathfrak{m}(\omega_1, \omega_2)$ is closer to ζ_0 than r_n , then all the correlations between ray ω_1 and ω_2 are small.

3. Applying [Slepian's lemma](#), we can get

$$\begin{aligned} & \mathbb{E}[(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)] \\ & \leq \mathbb{E}[(\mathbf{G}_r + \mu_1) \in \mathcal{B}(\omega_1) \cap \mathcal{E}(\omega_1)]^2 (1 + o(1)) \end{aligned}$$

Key estimates

1. By biasing,

$$\frac{\mathbb{E}[Y(\omega_1)Y(\omega_2)]}{\mathbb{E}[e^{\mathfrak{B}_{\omega_1} + \mathfrak{B}_{\omega_2}}]} = \mathbb{E}[(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)]$$

3. Applying **Slepian's lemma**, we can get

$$\begin{aligned} & \mathbb{E}[(\mathbf{G}_r + \mu) \in \mathcal{E}(\omega_1) \cap \mathcal{E}(\omega_2) \cap \mathcal{B}(\omega_1) \cap \mathcal{B}(\omega_2)] \\ & \leq \mathbb{E}[(\mathbf{G}_r + \mu_1) \in \mathcal{B}(\omega_1) \cap \mathcal{E}(\omega_1)]^2 (1 + o(1)) \end{aligned}$$

4. Conclude:

$$\mathbb{E}[Y(\omega_1)Y(\omega_2)] \leq \mathbb{E}[Y(\omega_1)]^2 (1 + o(1))$$

Key estimates

1. For a $1 - o(1)$ fraction of the pairs (ω_1, ω_2) ,

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \leq \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] (1 + o(1)).$$

2. For (ω_1, ω_2) very close, we also need an estimate that is *not too bad*:

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \lesssim \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] |\omega_1 - \omega_2|^{-1}$$

Key: Let $d = \lfloor d_{\mathbb{H}}(0, \mathbf{m}(\omega_1, \omega_2)) \rfloor$.

Key estimates

2. For (ω_1, ω_2) very close, we also need an estimate that is *not too bad*:

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \lesssim \mathbb{E} [Y(\omega_1)] \mathbb{E} [Y(\omega_2)] |\omega_1 - \omega_2|^{-1}$$

Key: Let $d = [d_{\mathbb{H}}(0, \mathbf{m}(\omega_1, \omega_2))]$.

On $\mathcal{B}(\omega_1)$, $\mathbf{G}_r(\omega_1 \zeta_d) \leq \nu(d)$:

Key estimates

Key: Let $d = [d_{\mathbb{H}}(0, \mathbf{m}(\omega_1, \omega_2))]$.

On $\mathcal{B}(\omega_1)$, $\mathbf{G}_r(\omega_1 \zeta_d) \leq \nu(d)$:

$$\mathbb{E} [Y(\omega_1)Y(\omega_2)] \leq \mathbb{E} \left[e^{\mathfrak{B}_{\omega_1} + \mathfrak{B}_{\omega_2} - \mathbf{G}_r(\omega_1 \zeta_d) + \nu(d)} \mathbf{1} \{ \mathcal{B} \dots \} \right].$$

Barrier method moral

For finite $\mathbf{z}, \mathbf{w} \subset \mathbb{D}$ and field \mathbf{F} , set

$$\mathfrak{B}(\mathbf{F}) = \sum_{z \in \mathbf{z}} 2\mathbf{F}(z) - \sum_{w \in \mathbf{w}} 2\mathbf{F}(w).$$

Barrier method moral

For finite $\mathbf{z}, \mathbf{w} \subset \mathbb{D}$ and field \mathbf{F} , set

$$\mathfrak{B}(\mathbf{F}) = \sum_{z \in \mathbf{z}} 2\mathbf{F}(z) - \sum_{w \in \mathbf{w}} 2\mathbf{F}(w).$$

It suffices to compute

$$\frac{\mathbb{E} [F(\mathbf{G})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}}$$

for bounded cylinder functions F with accuracy n^{-100} to conclude

$$\frac{\max_{\omega \in \mathbb{T}_n} \mathbf{G}(\omega \zeta_n) - n}{\log n} \xrightarrow{\mathbb{P}} -\frac{3}{4}$$

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. **Tilted, mesoscopic, and quantitative CLT.**
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;
2. for $F \in \mathcal{B}(\{\zeta_i \omega_j : 1 \leq i \leq n - m \log n, 1 \leq j \leq 2\})$;

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;
2. for $F \in \mathcal{B}(\{\zeta_i \omega_j : 1 \leq i \leq n - m \log n, 1 \leq j \leq 2\})$;

we have

$$\left| \frac{\mathbb{E} [F(\mathbf{U} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{U})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{U})}} - \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}} \right| \leq C(\log N)^{-K}.$$

Tilted, mesoscopic CLT

Let \mathbf{Z} on \mathbb{D} be white noise.

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;
2. for $F \in \mathcal{B}(\{\zeta_i \omega_j : 1 \leq i \leq n - m \log n, 1 \leq j \leq 2\})$;

we have

$$\left| \frac{\mathbb{E} [F(\mathbf{U} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{U})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{U})}} - \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}} \right| \leq C(\log N)^{-K}.$$

CLT commentary

1. It is not true that $\mathbb{E}e^{\mathfrak{B}(\mathbf{G})} = \mathbb{E}e^{\mathfrak{B}(\mathbf{U})}(1 + o(1))$ for all those \mathfrak{B} considered.

CLT commentary

1. It is not true that $\mathbb{E}e^{\mathfrak{B}(\mathbf{G})} = \mathbb{E}e^{\mathfrak{B}(\mathbf{U})}(1 + o(1))$ for all those \mathfrak{B} considered.
2. For any $K > 0$, there is an $m > 0$ sufficiently small so that conclusion of the CLT is false.

CLT commentary

1. It is not true that $\mathbb{E}e^{\mathfrak{B}(\mathbf{G})} = \mathbb{E}e^{\mathfrak{B}(\mathbf{U})}(1 + o(1))$ for all those \mathfrak{B} considered.
2. For any $K > 0$, there is an $m > 0$ sufficiently small so that conclusion of the CLT is false.
3. With $n_0 = n - m \log n$, using the CLT and the barrier method, it is possible to control

$$\max_{\omega \in \mathbb{T}} \mathbf{U}(\omega \zeta_{n_0})$$

CLT commentary

1. It is not true that $\mathbb{E}e^{\mathfrak{B}(\mathbf{G})} = \mathbb{E}e^{\mathfrak{B}(\mathbf{U})}(1 + o(1))$ for all those \mathfrak{B} considered.
2. For any $K > 0$, there is an $m > 0$ sufficiently small so that conclusion of the CLT is false.
3. With $n_0 = n - m \log n$, using the CLT and the barrier method, it is possible to control

$$\max_{\omega \in \mathbb{T}} \mathbf{U}(\omega \zeta_{n_0})$$

4. Corollary: for any $\epsilon > 0$

$$\frac{\max_{z \in \mathbb{D}} \mathbf{U}(z) - n_0}{\log n_0} \geq -\frac{3}{4} - \epsilon$$

with probability going to 1.

CLT commentary

3. With $n_0 = n - m \log n$, using the CLT and the barrier method, it is possible to control

$$\max_{\omega \in \mathbb{T}} \mathbf{U}(\omega \zeta_{n_0})$$

4. Corollary: for any $\epsilon > 0$

$$\frac{\max_{z \in \mathbb{D}} \mathbf{U}(z) - n_0}{\log n_0} \geq -\frac{3}{4} - \epsilon$$

with probability going to 1.

5. Quasi-corollary:

$$\max_{z \in \mathbb{T}_n} \mathbf{U}(z) \leq n - \frac{3}{4} \log n + C \log \log n$$

with probability going to 1.

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Basics

Let $f \in L^1(\mathcal{T})$. Define the N -th order Toeplitz determinant with symbol f by

$$D_N(f) = \det \left(\hat{f}(j-k) \right)_{1 \leq j, k \leq N},$$

with $\hat{f}(k)$ the k -th Fourier coefficient of f , i.e.

$$f(e^{i\theta}) = \sum_{k=-\infty}^{\infty} \hat{f}(k) e^{ik\theta}.$$

Basics

Let $f \in L^1(\mathcal{T})$. Define the N -th order Toeplitz determinant with symbol f by

$$D_N(f) = \det \left(\hat{f}(j-k) \right)_{1 \leq j, k \leq N},$$

with $\hat{f}(k)$ the k -th Fourier coefficient of f , i.e.

$$f(e^{i\theta}) = \sum_{k=-\infty}^{\infty} \hat{f}(k) e^{ik\theta}.$$

Heine identity:

$$\mathbb{E} \left[\prod_{h=1}^N f(e^{i\theta_h}) \right] = D_N(f).$$

Baxter-Cauchy identity

For some $|a_j| < 1$ and $|b_j| < 1$:

$$U_\ell(z) = \prod_{j=1}^{\ell} (1 - a_j z), \quad V_m(z) = \prod_{j=1}^m (1 - b_j z).$$

Baxter-Cauchy identity

For some $|a_j| < 1$ and $|b_j| < 1$:

$$U_\ell(z) = \prod_{j=1}^{\ell} (1 - a_j z), \quad V_m(z) = \prod_{j=1}^m (1 - b_j z).$$

Proposition (Baxter)

If $N \geq \ell$ or $N \geq m$,

$$D_N \left(\frac{1}{U_\ell(e^{-i\theta}) V_m(e^{i\theta})} \right) = \prod_{i=1}^m \prod_{j=1}^{\ell} \frac{1}{1 - a_j b_i}.$$

Baxter-Cauchy identity

For some $|a_j| < 1$ and $|b_j| < 1$:

$$U_\ell(z) = \prod_{j=1}^{\ell} (1 - a_j z), \quad V_m(z) = \prod_{j=1}^m (1 - b_j z).$$

Proposition (Baxter)

If $N \geq \ell$ or $N \geq m$,

$$D_N \left(\frac{1}{U_\ell(e^{-i\theta}) V_m(e^{i\theta})} \right) = \prod_{i=1}^m \prod_{j=1}^{\ell} \frac{1}{1 - a_j b_i}.$$

Also:

$$\prod_{i=1}^m \prod_{j=1}^{\ell} \frac{1}{1 - a_j b_i} = \exp \left(\frac{1}{2\pi i} \int_{S^1} \log \left(\frac{1}{V_m(z)} \right) \frac{d}{dz} \log(U_\ell(z^{-1})) dz \right).$$

Baxter-Cauchy identity

For some $|a_j| < 1$ and $|b_j| < 1$:

$$U_\ell(z) = \prod_{j=1}^{\ell} (1 - a_j z), \quad V_m(z) = \prod_{j=1}^m (1 - b_j z).$$

Proposition (Baxter)

If $N \geq \ell$ or $N \geq m$,

$$D_N \left(\frac{1}{U_\ell(e^{-i\theta}) V_m(e^{i\theta})} \right) = \prod_{i=1}^m \prod_{j=1}^{\ell} \frac{1}{1 - a_j b_i}.$$

Proof: apply partial fractions to the symbol.

Baxter-Cauchy identity

For some $|a_j| < 1$ and $|b_j| < 1$:

$$U_\ell(z) = \prod_{j=1}^{\ell} (1 - a_j z), \quad V_m(z) = \prod_{j=1}^m (1 - b_j z).$$

Proposition (Baxter)

If $N \geq \ell$ or $N \geq m$,

$$D_N \left(\frac{1}{U_\ell(e^{-i\theta}) V_m(e^{i\theta})} \right) = \prod_{i=1}^m \prod_{j=1}^{\ell} \frac{1}{1 - a_j b_i}.$$

Proof: apply partial fractions to the symbol. Observe the Toeplitz determinant is a product of vandermonde determinants.

Generalization

Proposition (P'–Zeitouni)

For any $N \geq \ell > k \geq 0$, any polynomial p nonvanishing on $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ of degree at most N :

$$\begin{aligned}
 D_N \left(\frac{p(e^{i\theta})e^{-ik\theta}}{U_\ell(e^{-i\theta})V_m(e^{i\theta})} \right) &= (-1)^{\binom{k}{2} + kN} p(0)^{N+k} \\
 &\cdot \exp \left(\frac{1}{2\pi i} \int_\gamma \log \left(\frac{p(z)}{p(0)V_m(z)} \right) \frac{d}{dz} \log(U_\ell(z^{-1})) dz \right) \\
 &\cdot \frac{1}{(2\pi i)^k} \int \cdots \int_\gamma \frac{\Delta(z_1, \dots, z_k)^2}{k!} \prod_{i=1}^k \frac{V_m(z_i)}{p(z_i)z_i^{N+k} U_\ell(z_i^{-1})} dz_i,
 \end{aligned}$$

where γ encloses $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ but no zeroes of p .

Generalization

Proposition (P'–Zeitouni)

For any $N \geq \ell > k \geq 0$, any polynomial p nonvanishing on $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ of degree at most N :

$$\begin{aligned}
 D_N \left(\frac{p(e^{i\theta})e^{-ik\theta}}{U_\ell(e^{-i\theta})V_m(e^{i\theta})} \right) &= (-1)^{\binom{k}{2} + kN} p(0)^{N+k} \\
 &\cdot \exp \left(\frac{1}{2\pi i} \int_\gamma \log \left(\frac{p(z)}{p(0)V_m(z)} \right) \frac{d}{dz} \log(U_\ell(z^{-1})) dz \right) \\
 &\cdot \frac{1}{(2\pi i)^k} \int \cdots \int_\gamma \frac{\Delta(z_1, \dots, z_k)^2}{k!} \prod_{i=1}^k \frac{V_m(z_i)}{p(z_i)z_i^{N+k} U_\ell(z_i^{-1})} dz_i,
 \end{aligned}$$

where γ encloses $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ but no zeroes of p .

Generalization

Proposition (P'-Zeitouni)

For any $N \geq \ell > k \geq 0$, any polynomial p nonvanishing on $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ of degree at most N :

$$\begin{aligned}
 D_N \left(\frac{p(e^{i\theta})e^{-ik\theta}}{U_\ell(e^{-i\theta})V_m(e^{i\theta})} \right) &= (-1)^{\binom{k}{2} + kN} p(0)^{N+k} \\
 &\cdot \exp \left(\frac{1}{2\pi i} \int_\gamma \log \left(\frac{p(z)}{p(0)V_m(z)} \right) \frac{d}{dz} \log(U_\ell(z^{-1})) dz \right) \\
 &\cdot \frac{1}{(2\pi i)^k} \int \cdots \int_\gamma \frac{\Delta(z_1, \dots, z_k)^2}{k!} \prod_{i=1}^k \frac{V_m(z_i)}{p(z_i)z_i^{N+k}U_\ell(z_i^{-1})} dz_i,
 \end{aligned}$$

where γ encloses $\{0\} \cup \{a_j\}_{j=1}^{\ell}$ but no zeroes of p .

Example

Consider evaluating the Toeplitz determinant with symbol
 $|1 - z_1 e^{i\theta}|^2 / |1 - z_2 e^{i\theta}|^2$.

Example

Consider evaluating the Toeplitz determinant with symbol $|1 - z_1 e^{i\theta}|^2 / |1 - z_2 e^{i\theta}|^2$.

This makes $p(x) = (1 - z_1 x)(x - \bar{z}_1)$, $U_1 = (1 - \bar{z}_2 x)$ and $V_1 = (1 - z_2 x)$.

Example

Consider evaluating the Toeplitz determinant with symbol $|1 - z_1 e^{i\theta}|^2 / |1 - z_2 e^{i\theta}|^2$.

This makes $p(x) = (1 - z_1 x)(x - \bar{z}_1)$, $U_1 = (1 - \bar{z}_2 x)$ and $V_1 = (1 - z_2 x)$.

Applying Proposition 2,

$$D_N \left(\frac{p(e^{i\theta})e^{-i\theta}}{U_\ell(e^{-i\theta})V_m(e^{i\theta})} \right) = \frac{|1 - \bar{z}_1 z_2|^2}{(1 - |z_1|^2)(1 - |z_2|^2)} - \frac{|z_1 - z_2|^2 |z_1|^{2N}}{(1 - |z_1|^2)(1 - |z_2|^2)},$$

for all N sufficiently large.

Main identity

Proposition (P'–Zeitouni)

Let $\mathbf{z}, \mathbf{y} \subset \mathbb{D}$, $|\mathbf{z}| = |\mathbf{y}|$. Let f_1 and f_2 be polynomials of degree strictly less than $N/2$ with $f_1(0) = f_2(0) = 1$ and without zeros in $\bar{\mathbb{D}}$ so that $z \mapsto f_1(z^{-1})$ does not vanish on \mathbf{z} .

$$\begin{aligned}
 & D_N \left(\frac{\prod_{z \in \mathbf{z}} |1 - ze^{i\theta}|^2}{f_1(e^{-i\theta}) f_2(e^{i\theta}) \prod_{y \in \mathbf{y}} |1 - ye^{i\theta}|^2} \right) = \mathbb{E} e^{\mathfrak{B}(\mathbf{G})} \\
 & \cdot \exp \left(\frac{1}{2\pi i} \int_{\gamma} \log \left(\frac{1}{f_2(z)} \right) \frac{d}{dz} \log(f_1(z^{-1})) dz \right) \cdot \frac{\prod_{z \in \mathbf{z}} f_1(z) f_2(\bar{z})}{\prod_{y \in \mathbf{y}} f_1(y) f_2(\bar{y})} \\
 & \cdot \sum_{\substack{S_1, S_2 \subset \mathbf{z} \\ |S_1| = |S_2|}} (-1)^{|S_1|} \cdot \left[\prod_{S_1} \frac{z^N f_1(z^{-1})}{f_2(z)} \right] \cdot \left[\prod_{S_2} \frac{\bar{z}^N f_2(\bar{z}^{-1})}{f_1(\bar{z})} \right] \cdot c^{\mathbf{y}}(S_1, S_2) \cdot c_{\mathbf{z}}(S_1, S_2).
 \end{aligned}$$

Main identity

Proposition (P'–Zeitouni)

Let $\mathbf{z}, \mathbf{y} \subset \mathbb{D}$, $|\mathbf{z}| = |\mathbf{y}|$. Let f_1 and f_2 be polynomials of degree strictly less than $N/2$ with $f_1(0) = f_2(0) = 1$ and without zeros in $\bar{\mathbb{D}}$ so that $z \mapsto f_1(z^{-1})$ does not vanish on \mathbf{z} .

$$\begin{aligned}
 & D_N \left(\frac{\prod_{z \in \mathbf{z}} |1 - ze^{i\theta}|^2}{f_1(e^{-i\theta}) f_2(e^{i\theta}) \prod_{y \in \mathbf{y}} |1 - ye^{i\theta}|^2} \right) = \mathbb{E} e^{\mathfrak{B}(\mathbf{G})} \\
 & \cdot \exp \left(\frac{1}{2\pi i} \int_{\gamma} \log \left(\frac{1}{f_2(z)} \right) \frac{d}{dz} \log(f_1(z^{-1})) dz \right) \cdot \frac{\prod_{z \in \mathbf{z}} f_1(z) f_2(\bar{z})}{\prod_{y \in \mathbf{y}} f_1(y) f_2(\bar{y})} \\
 & \cdot \sum_{\substack{S_1, S_2 \subset \mathbf{z} \\ |S_1| = |S_2|}} (-1)^{|S_1|} \cdot \left[\prod_{S_1} \frac{z^N f_1(z^{-1})}{f_2(z)} \right] \cdot \left[\prod_{S_2} \frac{\bar{z}^N f_2(\bar{z}^{-1})}{f_1(\bar{z})} \right] \cdot c^{\mathbf{y}}(S_1, S_2) \cdot c_{\mathbf{z}}(S_1, S_2).
 \end{aligned}$$

Main identity

Proposition (P'–Zeitouni)

Let $\mathbf{z}, \mathbf{y} \subset \mathbb{D}$, $|\mathbf{z}| = |\mathbf{y}|$. Let f_1 and f_2 be polynomials of degree strictly less than $N/2$ with $f_1(0) = f_2(0) = 1$ and without zeros in $\bar{\mathbb{D}}$ so that $z \mapsto f_1(z^{-1})$ does not vanish on \mathbf{z} .

$$\begin{aligned}
 & D_N \left(\frac{\prod_{z \in \mathbf{z}} |1 - ze^{i\theta}|^2}{f_1(e^{-i\theta})f_2(e^{i\theta}) \prod_{y \in \mathbf{y}} |1 - ye^{i\theta}|^2} \right) = \mathbb{E} e^{\mathfrak{B}(\mathbf{G})} \\
 & \cdot \exp \left(\frac{1}{2\pi i} \int_{\gamma} \log \left(\frac{1}{f_2(z)} \right) \frac{d}{dz} \log(f_1(z^{-1})) dz \right) \cdot \frac{\prod_{z \in \mathbf{z}} f_1(z)f_2(\bar{z})}{\prod_{y \in \mathbf{y}} f_1(y)f_2(\bar{y})} \\
 & \cdot \sum_{\substack{S_1, S_2 \subset \mathbf{z} \\ |S_1| = |S_2|}} (-1)^{|S_1|} \cdot \left[\prod_{S_1} \frac{z^N f_1(z^{-1})}{f_2(z)} \right] \cdot \left[\prod_{S_2} \frac{\bar{z}^N f_2(\bar{z}^{-1})}{f_1(\bar{z})} \right] \cdot c^{\mathbf{y}}(S_1, S_2) \cdot c_{\mathbf{z}}(S_1, S_2).
 \end{aligned}$$

Main identity

Proposition (P'–Zeitouni)

Let $\mathbf{z}, \mathbf{y} \subset \mathbb{D}$, $|\mathbf{z}| = |\mathbf{y}|$. Let f_1 and f_2 be polynomials of degree strictly less than $N/2$ with $f_1(0) = f_2(0) = 1$ and without zeros in $\bar{\mathbb{D}}$ so that $z \mapsto f_1(z^{-1})$ does not vanish on \mathbf{z} .

$$\begin{aligned}
 & D_N \left(\frac{\prod_{z \in \mathbf{z}} |1 - ze^{i\theta}|^2}{f_1(e^{-i\theta})f_2(e^{i\theta}) \prod_{y \in \mathbf{y}} |1 - ye^{i\theta}|^2} \right) = \mathbb{E} e^{\mathfrak{B}(\mathbf{G})} \\
 & \cdot \exp \left(\frac{1}{2\pi i} \int_{\gamma} \log \left(\frac{1}{f_2(z)} \right) \frac{d}{dz} \log(f_1(z^{-1})) dz \right) \cdot \frac{\prod_{z \in \mathbf{z}} f_1(z)f_2(\bar{z})}{\prod_{y \in \mathbf{y}} f_1(y)f_2(\bar{y})} \\
 & \cdot \sum_{\substack{S_1, S_2 \subset \mathbf{z} \\ |S_1| = |S_2|}} (-1)^{|S_1|} \cdot \left[\prod_{S_1} \frac{z^N f_1(z^{-1})}{f_2(z)} \right] \cdot \left[\prod_{S_2} \frac{\bar{z}^N f_2(\bar{z}^{-1})}{f_1(\bar{z})} \right] \cdot c^{\mathbf{y}}(S_1, S_2) \cdot c_{\mathbf{z}}(S_1, S_2).
 \end{aligned}$$

Main identity

Proposition (P'–Zeitouni)

Let $\mathbf{z}, \mathbf{y} \subset \mathbb{D}$, $|\mathbf{z}| = |\mathbf{y}|$. Let f_1 and f_2 be polynomials of degree strictly less than $N/2$ with $f_1(0) = f_2(0) = 1$ and without zeros in $\bar{\mathbb{D}}$ so that $z \mapsto f_1(z^{-1})$ does not vanish on \mathbf{z} .

$$\begin{aligned}
 & D_N \left(\frac{\prod_{z \in \mathbf{z}} |1 - ze^{i\theta}|^2}{f_1(e^{-i\theta})f_2(e^{i\theta}) \prod_{y \in \mathbf{y}} |1 - ye^{i\theta}|^2} \right) = \mathbb{E} e^{\mathfrak{B}(\mathbf{G})} \\
 & \cdot \exp \left(\frac{1}{2\pi i} \int_{\gamma} \log \left(\frac{1}{f_2(z)} \right) \frac{d}{dz} \log(f_1(z^{-1})) dz \right) \cdot \frac{\prod_{z \in \mathbf{z}} f_1(z)f_2(\bar{z})}{\prod_{y \in \mathbf{y}} f_1(y)f_2(\bar{y})} \\
 & \cdot \sum_{\substack{S_1, S_2 \subset \mathbf{z} \\ |S_1| = |S_2|}} (-1)^{|S_1|} \cdot \left[\prod_{S_1} \frac{z^N f_1(z^{-1})}{f_2(z)} \right] \cdot \left[\prod_{S_2} \frac{\bar{z}^N f_2(\bar{z}^{-1})}{f_1(\bar{z})} \right] \cdot c^{\mathbf{y}}(S_1, S_2) \cdot c_{\mathbf{z}}(S_1, S_2).
 \end{aligned}$$

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Perturbing the bias

Recall:

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;
2. for $F \in \mathcal{B}(\{\zeta_i \omega_j : 1 \leq i \leq n - m \log n, 1 \leq j \leq 2\})$;

we have

$$\left| \frac{\mathbb{E} [F(\mathbf{U} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{U})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{U})}} - \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}} \right| \leq C(\log N)^{-K}.$$

Perturbing the bias

Recall:

Theorem (P'–Zeitouni)

Fix $K > 0$. There exists $m > 100, C > 0$ so that:

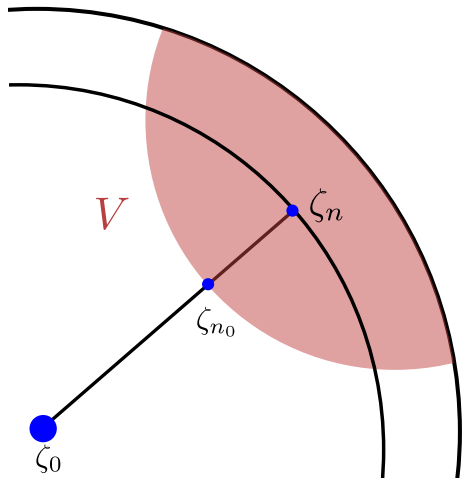
1. for $\mathbf{z}, \mathbf{w} \subset \mathbb{D}(1 - N^{-1})$;
2. for $F \in \mathcal{B}(\{\zeta_i \omega_j : 1 \leq i \leq n - m \log n, 1 \leq j \leq 2\})$;

we have

$$\left| \frac{\mathbb{E} [F(\mathbf{U} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{U})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{U})}} - \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}} \right| \leq C(\log N)^{-K}.$$

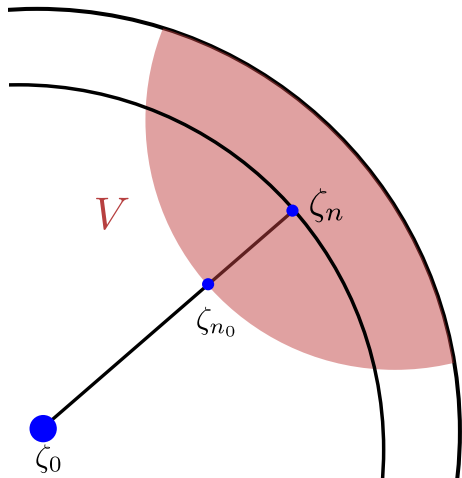
Perturbations of \mathfrak{B} only affect the mesoscopic distribution through a change of mean.

Perturbing the bias



1. Let V be the hyperbolic halfspace normal to $\{\zeta_j\}_1^\infty$.

Perturbing the bias

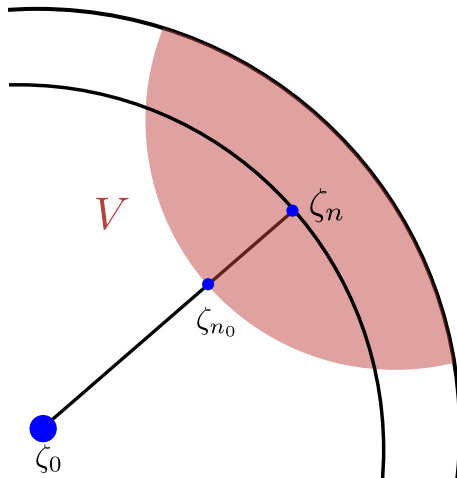


1. Let V be the hyperbolic halfspace normal to $\{\zeta_j\}_1^\infty$.
2. If $z, w \in V$ have $d_{\mathbb{H}}(z, w) \rightarrow 0$, then

$$\mathfrak{B}_x(\mathbf{F}) = \mathfrak{B}(\mathbf{F}) + \mathbf{F}(z) - \mathbf{F}(w)$$

induces the same mean (up to $o(1)$) on $\mathbf{G}(\zeta_k \omega)$ as \mathfrak{B} .

Perturbing the bias



2. If $z, w \in V$ have $d_{\mathbb{H}}(z, w) \rightarrow 0$, then

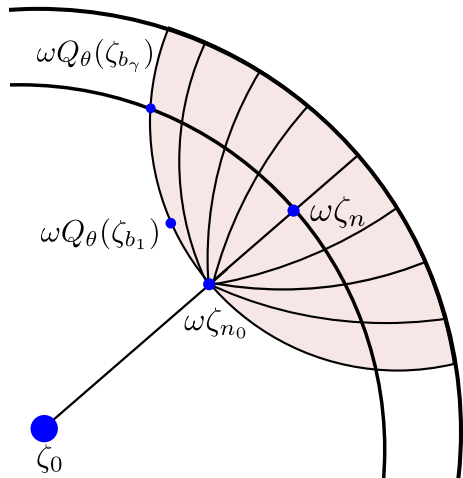
$$\mathfrak{B}_x(\mathbf{F}) = \mathfrak{B}(\mathbf{F}) + \mathbf{F}(z) - \mathbf{F}(w)$$

induces the same mean (up to $o(1)$) on $\mathbf{G}(\zeta_k w)$ as \mathfrak{B} .

3. Therefore,

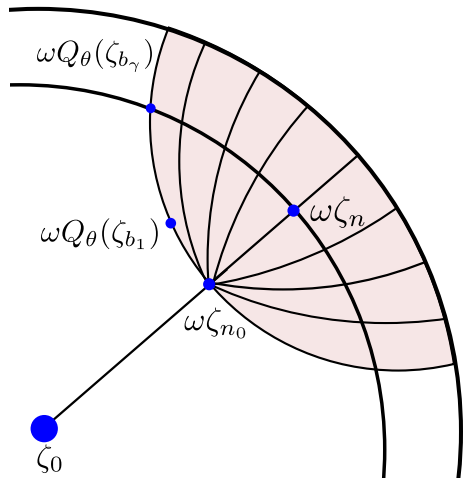
$$\begin{aligned} & \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}(\mathbf{G})}} \\ &= \frac{\mathbb{E} [F(\mathbf{G} + \mathbf{Z})e^{\mathfrak{B}_x(\mathbf{G})}]}{\mathbb{E} e^{\mathfrak{B}_x(\mathbf{G})}} + o(1) \end{aligned}$$

Microscopic landscape



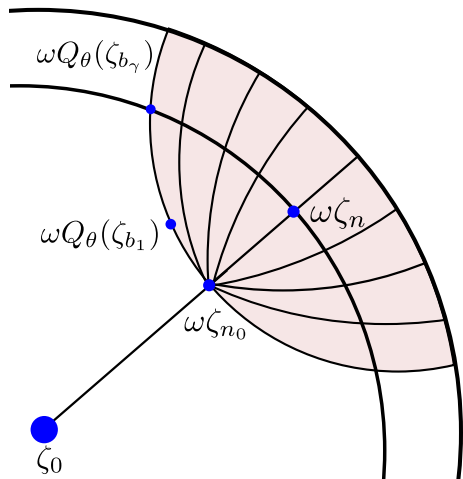
1. Let $\Theta = \{\theta\} \subset [-1, 1]$ be equally spaced of cardinality e^{n-n_0} .

Microscopic landscape



1. Let $\Theta = \{\theta\} \subset [-1, 1]$ be equally spaced of cardinality e^{n-n_0} .
2. Let Q_θ be the rotation about $\omega\zeta_{n_0}$ by angle θ .

Microscopic landscape



1. Let $\Theta = \{\theta\} \subset [-1, 1]$ be equally spaced of cardinality e^{n-n_0} .
2. Let Q_θ be the rotation about $\omega\zeta_{n_0}$ by angle θ .
3. Can we use only exponential moments to show

$$\frac{\max_{\theta \in \Theta} \mathbf{G}(Q_\theta(\zeta_n)) - \mathbf{G}(\zeta_{n_0})}{n - n_0} \xrightarrow{\mathbb{P}} 1?$$

Plan

1. Branching structures.
2. The barrier method for \mathbf{G} .
3. Tilted, mesoscopic, and quantitative CLT.
4. Baxter-Cauchy type Toeplitz determinant identities.
5. Controlling the microscopic field conditional on the mesoscopic.
6. Field moment calculus.

Final goal

1. By conformal covariance, we are back to bounding the maximum of $\mathbf{G}(\zeta_n)$ *using only exponential moments*.

Final goal

1. By conformal covariance, we are back to bounding the maximum of $\mathbf{G}(\zeta_n)$ *using only exponential moments*.
2. Let $\mathbb{T}_n \subset \mathbb{D}$ be

$$\left\{ e^{2\pi i j [e^{-n}]} : 1 \leq j \leq e^n \right\}$$

Final goal

1. By conformal covariance, we are back to bounding the maximum of $\mathbf{G}(\zeta_n)$ *using only exponential moments*.
2. Let $\mathbb{T}_n \subset \mathbb{D}$ be

$$\left\{ e^{2\pi i j [e^{-n}]} : 1 \leq j \leq e^n \right\}$$

3. We will sketch the argument for

$$\frac{\max_{\omega \in \mathbb{T}_n} \mathbf{G}(\omega \zeta_n)}{n} \xrightarrow{\mathbb{P}} 1$$

using only exponential moments.

Setup

1. Let $\{\eta_n\}$ be slowly growing (sublogarithmic).

Setup

1. Let $\{\eta_n\}$ be slowly growing (sublogarithmic).
2. $b_k = k \lfloor n/\eta \rfloor$.

Setup

1. Let $\{\eta_n\}$ be slowly growing (sublogarithmic).
2. $b_k = k \lfloor n/\eta \rfloor$.
- 3.

$$\mathcal{E}(\theta) = \{|\mathbf{F}(e^{i\theta} \zeta_n) - b_{\eta-1}| < \eta \cdot n^{1/2}\}$$

$$\mathcal{B}(\theta) = \{|\mathbf{F}(e^{i\theta} \zeta_{b_k}) - b_{k-1}| \leq \eta \cdot n^{1/2}, \forall 1 \leq k \leq \eta - 1\}$$

Second moment setup

Define

$$\mathfrak{B}_\theta(\mathbf{F}) = 2\mathbf{F}(e^{i\theta}\zeta_n) - 2\mathbf{F}(e^{i\theta}\zeta_{b_1}), \quad Y(\theta) = e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{1}\{\mathbf{G} \in \mathcal{B}(\theta) \cap \mathcal{E}(\theta)\}.$$

Second moment setup

Define

$$\mathfrak{B}_\theta(\mathbf{F}) = 2\mathbf{F}(e^{i\theta}\zeta_n) - 2\mathbf{F}(e^{i\theta}\zeta_{b_1}), \quad Y(\theta) = e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{1}_{\{\mathbf{G} \in \mathcal{B}(\theta) \cap \mathcal{E}(\theta)\}}.$$

In terms of this indicator, define the biased counting variable

$$Z = \sum_{\substack{h \in \mathbb{Z} \\ |h| \leq e^n}} Y(he^{-n}).$$

Second moment setup

Define

$$\mathfrak{B}_\theta(\mathbf{F}) = 2\mathbf{F}(e^{i\theta}\zeta_n) - 2\mathbf{F}(e^{i\theta}\zeta_{b_1}), Y(\theta) = e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{1}\{\mathbf{G} \in \mathcal{B}(\theta) \cap \mathcal{E}(\theta)\}.$$

In terms of this indicator, define the biased counting variable

$$Z = \sum_{\substack{h \in \mathbb{Z} \\ |h| \leq e^n}} Y(he^{-n}).$$

We then have

$$\mathbb{E}[\mathbf{1}\{Z > 0\}] \geq \frac{(\sum_\theta \mathbb{E}[Y(\theta)])^2}{\sum_{\theta_1, \theta_2} \mathbb{E}[Y(\theta_1)Y(\theta_2)]},$$

Key estimates

We want that for a $1 - o(1)$ fraction of the pairs (θ_1, θ_2) ,

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} [Y(\theta_1)] \mathbb{E} [Y(\theta_2)] (1 + o(1)).$$

Key estimates

We want that for a $1 - o(1)$ fraction of the pairs (θ_1, θ_2) ,

$$\mathbb{E}[Y(\theta_1)Y(\theta_2)] \leq \mathbb{E}[Y(\theta_1)] \mathbb{E}[Y(\theta_2)] (1 + o(1)).$$

Aside: For (θ_1, θ_2) very close, we also need an estimate that is *not too bad*:

$$\mathbb{E}[Y(\theta_1)Y(\theta_2)] \lesssim \mathbb{E}[Y(\theta_1)] \mathbb{E}[Y(\theta_2)] |\theta_1 - \theta_2|^{-1}$$

This is the reason for the $\mathcal{B}(\theta)$ event.

Key estimates

We want that for a $1 - o(1)$ fraction of the pairs (θ_1, θ_2) ,

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} [Y(\theta_1)] \mathbb{E} [Y(\theta_2)] (1 + o(1)).$$

Trivially bound:

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} \left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G}) + \mathfrak{B}_{\theta_2}(\mathbf{G})} \right].$$

Key estimates

We want that for a $1 - o(1)$ fraction of the pairs (θ_1, θ_2) ,

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} [Y(\theta_1)] \mathbb{E} [Y(\theta_2)] (1 + o(1)).$$

Trivially bound:

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} \left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G}) + \mathfrak{B}_{\theta_2}(\mathbf{G})} \right].$$

Using branching, for $|\theta_1 - \theta_2| = \omega(\eta^{-1})$

$$\mathbb{E} \left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G}) + \mathfrak{B}_{\theta_2}(\mathbf{G})} \right] = \mathbb{E} \left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G})} \right] \mathbb{E} \left[e^{\mathfrak{B}_{\theta_2}(\mathbf{G})} \right] (1 + o(1)).$$

Key estimates

Combining everything, for most pairs (θ_1, θ_2) :

$$\mathbb{E} [Y(\theta_1)Y(\theta_2)] \leq \mathbb{E} \left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G})} \right] \mathbb{E} \left[e^{\mathfrak{B}_{\theta_2}(\mathbf{G})} \right] (1 + o(1))$$

Key estimates

Combining everything, for most pairs (θ_1, θ_2) :

$$\mathbb{E}[Y(\theta_1)Y(\theta_2)] \leq \mathbb{E}\left[e^{\mathfrak{B}_{\theta_1}(\mathbf{G})}\right] \mathbb{E}\left[e^{\mathfrak{B}_{\theta_2}(\mathbf{G})}\right] (1 + o(1))$$

This reduces the problem to showing:

$$\mathbb{E}[Y(\theta)] \geq \mathbb{E}\left[e^{\mathfrak{B}_{\theta}(\mathbf{G})}\right] (1 - o(1)).$$

Key estimates

This reduces the problem to showing:

$$\mathbb{E}[Y(\theta)] \geq \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] (1 - o(1)).$$

Why is this possible?

Key estimates

This reduces the problem to showing:

$$\mathbb{E}[Y(\theta)] \geq \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] (1 - o(1)).$$

Why is this possible?

$$\mu_\theta(e^{i\theta} b_j) := \frac{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} b_j) \right]}{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right]} \approx b_{j-1}.$$

Key estimates

This reduces the problem to showing:

$$\mathbb{E}[Y(\theta)] \geq \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] (1 - o(1)).$$

Why is this possible?

$$\mu_\theta(e^{i\theta} b_j) := \frac{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} b_j) \right]}{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right]} \approx b_{j-1}.$$

Recall, $Y(\theta) > 0$ requires all $|\mathbf{G}(e^{i\theta} \zeta_{b_j}) - b_{j-1}| \leq \eta n^{1/2}$.

Key estimates

This reduces the problem to showing:

$$\mathbb{E}[Y(\theta)] \geq \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] (1 - o(1)).$$

Why is this possible?

$$\mu_\theta(e^{i\theta} b_j) := \frac{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} b_j) \right]}{\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right]} \approx b_{j-1}.$$

Recall, $Y(\theta) > 0$ requires all $|\mathbf{G}(e^{i\theta} \zeta_{b_j}) - b_{j-1}| \leq \eta n^{1/2}$.

Ideally, we could estimate:

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_{b_j})^k \right]$$

for $k = 1, 2$ and $j = 1, 2, \dots, \eta$.

Field moment calculus

Field moment calculus

Suppose we want to estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right].$$

Field moment calculus

Suppose we want to estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right].$$

Let $\{0 = t_0 < t_1 < t_2 < \dots < t_k = t\}$ be an equally spaced grid, and write

$$\sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} (\mathbf{G}(ie^{i\theta} \zeta_{t_j}) - \mathbf{G}(ie^{i\theta} \zeta_{t_{j-1}})) \right].$$

Field moment calculus

Suppose we want to estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right].$$

Let $\{0 = t_0 < t_1 < t_2 < \dots < t_k = t\}$ be an equally spaced grid, and write

$$\sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} (\mathbf{G}(ie^{i\theta} \zeta_{t_j}) - \mathbf{G}(ie^{i\theta} \zeta_{t_{j-1}})) \right].$$

Now use that

$$\mathbf{G}(e^{i\theta} \zeta_{t_j}) - \mathbf{G}(e^{i\theta} \zeta_{t_{j-1}}) \leq \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2}.$$

Field moment calculus

Suppose we want to estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right].$$

Let $\{0 = t_0 < t_1 < t_2 < \dots < t_k = t\}$ be an equally spaced grid, and write

$$\sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} (\mathbf{G}(ie^{i\theta} \zeta_{t_j}) - \mathbf{G}(ie^{i\theta} \zeta_{t_{j-1}})) \right].$$

Now use that

$$\mathbf{G}(e^{i\theta} \zeta_{t_j}) - \mathbf{G}(e^{i\theta} \zeta_{t_{j-1}}) \leq \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2}.$$

Then we can estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \leq \sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2} \right].$$

Field moment calculus

1. Then we can estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{L}_N(e^{i\theta} \zeta_t) \right] \leq \sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2} \right].$$

Field moment calculus

1. Then we can estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{L}_N(e^{i\theta} \zeta_t) \right] \leq \sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2} \right].$$

2. Provided that $|t_j - t_{j-1}| = o(1)$, the smoothness of the field \mathbf{G} implies that

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \lesssim \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \sum_{j=1}^k (\mu_\theta(e^{i\theta} \zeta_{t_j}) - \mu_\theta(e^{i\theta} \zeta_{t_{j-1}})) \right].$$

Field moment calculus

1. Then we can estimate

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{L}_N(e^{i\theta} \zeta_t) \right] \leq \sum_{j=1}^k \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \frac{e^{2\mathbf{G}(e^{i\theta} \zeta_{t_j}) - 2\mathbf{G}(e^{i\theta} \zeta_{t_{j-1}})} - 1}{2} \right].$$

2. Provided that $|t_j - t_{j-1}| = o(1)$, the smoothness of the field \mathbf{G} implies that

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \lesssim \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] \sum_{j=1}^k (\mu_\theta(e^{i\theta} \zeta_{t_j}) - \mu_\theta(e^{i\theta} \zeta_{t_{j-1}})).$$

3. Provided that $|t_j - t_{j-1}| = o(1)$, the smoothness of the field \mathbf{T} implies that

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \lesssim \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] \mu_\theta(e^{i\theta} \zeta_t).$$

Field moment calculus

3. Provided that $|t_j - t_{j-1}| = o(1)$, the smoothness of the field \mathbf{T} implies that

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \lesssim \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] \mu_\theta(e^{i\theta} \zeta_t).$$

4. $\mu_\theta(e^{i\theta} \zeta_t) \approx (t - b_1)_+$ is the mean of \mathbf{G} after biasing by $e^{\mathfrak{B}_\theta(\mathbf{G})}$.

Field moment calculus

3. Provided that $|t_j - t_{j-1}| = o(1)$, the smoothness of the field \mathbf{T} implies that

$$\mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \mathbf{G}(e^{i\theta} \zeta_t) \right] \lesssim \mathbb{E} \left[e^{\mathfrak{B}_\theta(\mathbf{G})} \right] \mu_\theta(e^{i\theta} \zeta_t).$$

4. $\mu_\theta(e^{i\theta} \zeta_t) \approx (t - b_1)_+$ is the mean of \mathbf{G} after biasing by $e^{\mathfrak{B}_\theta(\mathbf{G})}$.
5. A similar technique can be used for the second moments and lower bounds.

Side comment

Side theorem: using these techniques, as well as the [Fyodorov-Strahov identity](#) and [DKMVZ orthogonal polynomial asymptotics](#):

Theorem (Lambert–P' 16+)

Let $\mathbf{L}_N(z)$ be the centered log-potential of a $\beta = 2$ random matrix, with potential V real analytic on \mathbb{R} , one-cut and with regular boundary conditions. Then:

$$\frac{\max_{|z|=1} \mathbf{L}_N(z)}{\log N} \xrightarrow[N \rightarrow \infty]{\mathbb{P}} 1,$$

Side comment

Side theorem: using these techniques, as well as the [Fyodorov-Strahov identity](#) and [DKMVZ orthogonal polynomial asymptotics](#):

Theorem (Lambert–P' 16+)

Let $\mathbf{L}_N(z)$ be the centered log-potential of a $\beta = 2$ random matrix, with potential V real analytic on \mathbb{R} , one-cut and with regular boundary conditions. Then:

$$\frac{\max_{|z|=1} \mathbf{L}_N(z)}{\log N} \xrightarrow[N \rightarrow \infty]{\mathbb{P}} 1,$$

Thanks!