The maximum of the characteristic polynomial for a random permutation matrix

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Nick Cook, UCLA Based on joint work with Ofer Zeitouni

Model and previous work

- Let P_N be a uniform random $N \times N$ permutation matrix and let $\chi_N(z) = \det(zI_N P_N)$ denote its characteristic polynomial.
- Consider the random field $X_N : \mathbb{R}/\mathbb{Z} \to [-\infty, \infty)$

$$X_N(t) = \log |\chi_N(e(-t))| = \log |\det(I_N - e(t)P_N)|$$

where $e(t) := \exp(2\pi i t)$.

 Hambly-Keevash-O'Connell-Stark '99 obtained a CLT: For fixed t ∈ ℝ/ℤ of finite type,

$$rac{X_{N}(t)}{\sqrt{rac{\pi^{2}}{12}\log N}}
ightarrow N(0,1)$$

(and similarly for the imaginary part of log χ_N).

- Note X_N is badly behaved at rational points (atom at $-\infty$).
- Multidimensional CLT obtained by Dang-Zeindler '13.

Numerical simulations



Figure 1: Simulated $X_N(t)$ with $N = 10^4$ for $t \in [0.1, 0.11]$. Cycle structure of P_N : 6310, 1914, 909, 668, 79, 47, 33, 19, 12, 5, 3, 1. (Generated using the Chinese restaurant process.)

Numerical simulations



Figure 2: Simulated $X_N(t)$ with $N = 10^9$ for $t \in [0.2, 0.35]$. Cycle structure of P_N : 892,060,223, 78,087,020, 19,479,718, 9,152,317, 630,684, 352,623, 114,502, 104,059, 8,973, 8,193, 1,641, 33, 5, 3, 2, 2, 1, 1.

$$X_N(t) = \log |\det(I_N - e(t)P_N)|.$$

Theorem (C., Zeitouni '18)

$$rac{1}{\log N} \sup_{t \in \mathbb{R}/\mathbb{Z}} X_N(t) o x_c pprox 0.677$$
 in probability.

(Informally:

$$\sup_{|z|=1} |\chi_N(z)| = N^{x_c + o(1)} \qquad \text{with high probability.})$$

Related work: Maximum of the CUE field

Replacing P_N with a Haar unitary U_N , we obtain the CUE field:

$$X_N^{cue}(t) = \log |\det(I_N - e(t)U_N)|.$$

Conjecture (Fyodorov–Hiary–Keating '12):

$$M_N := \sup_{t \in \mathbb{R}/\mathbb{Z}} X_N^{ ext{cue}}(t) - \left(\log N - rac{3}{4}\log\log N
ight)$$
 converges in distribution.

$$\sup_{t \in \mathbb{R}/\mathbb{Z}} X_N^{\text{cue}}(t) = \log N + o_P(\log N)$$
Arguin–Belius–Bourgade '15

Related work of Arguin–Belius–Bourgade–Soundararajan–Radziwiłł'16 on the Riemann ζ function.

All proofs proceed by exposing an underlying branching structure.

Note that, unlike X_N , distribution of X_N^{cue} is invariant under rotations.

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$$= \log N - \frac{3}{4} \log \log N + o_P(\log \log N) \qquad \text{Paquette-Zeitouni '15}$$

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$$= \log N - \frac{3}{4} \log \log N + O_P(1)$$
Chhaibi-Madaule-Najnudel '16

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Maximal displacement for branching random walk (BRW)

- X_N and X^{cue}_N are logarithmically correlated fields
- Archetypical log-correlated field is (binary, Gaussian) BRW, which we view as a random field X^{brw}_n(t), t ∈ [0, 1].
- For each $t \in [0,1]$, $X_n^{\mathrm{brw}}(t) \sim N(0,n)$.



$$\operatorname{Cov}(X_n(s) - X_m(s), \, X_n(t) - X_m(t)) = 0$$
 for any $\operatorname{anc}(s, t) < m < n.$



Maximal displacement for branching random walk (BRW)

Theorem (Hammersley '74, Kingman '75, Biggins '77) $\frac{1}{n} \sup_{t \in [0,1)} X_n^{\text{brw}}(t) \to b_c = \sqrt{2 \log 2} \quad \text{in probability.}$

- We recall some key proof ideas going back to Bramson '78.
- Let $T_n = \{k2^{-n} \in [0,1)\}$ and put $S_n(b) = \{t \in T_n : X_n^{brw}(t) \ge bn\}.$
- Upper bound. First moment method (union bound):

$$\mathbb{E} \left| \mathcal{S}_n(b_c + \epsilon) \right| = 2^n \mathbb{P}(X_n^{\mathsf{brw}}(1) \ge (b_c + \epsilon)n) \le e^{-c(\epsilon)n}.$$

Then apply Markov's inequality.

Lower bound. Same computation shows E |S_n(b_c − ε)| ≥ e^{c'(ε)n}, so we'd be done if we can establish concentration, i.e.

$$\frac{\operatorname{Var}|\mathcal{S}_n(b_c-\epsilon)|}{(\mathbb{E}|\mathcal{S}_n(b_c-\epsilon)|)^2} = o(1).$$

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But this is false. (See the board...)

Maximal displacement for branching random walk (BRW)

- Modified second moment: Rather than counting high points S_n(b), we should only count points that make steady progress toward b_cn.
- Toy computation: Let $X_n^{(1)} = X_{n/2}^{\text{brw}}$, $X_n^{(2)} = X_n^{\text{brw}} X_{n/2}^{\text{brw}}$, and put

$$\mathcal{E}_2(t) = \Big\{ X_n^{(1)}(t), \, X_n^{(2)}(t) \ge \frac{bn}{2} \Big\}.$$

For pairs (s, t) with anc(s, t) < n/2, we can bound

$$\mathbb{P}(\mathcal{E}_2(s)\cap\mathcal{E}_2(t))=\mathbb{P}\left(X_{n/2}^{(1)}(s),X_{n/2}^{(1)}(t)\geq rac{bn}{2}
ight)\mathbb{P}\left(X_n^{(2)}(s),X_n^{(2)}(t)\geq rac{bn}{2}
ight)\ \leq\mathbb{P}\left(X_{n/2}^{(1)}(s)\geq rac{bn}{2}
ight)\mathbb{P}\left(X_n^{(2)}(s)\geq rac{bn}{2}
ight)^2$$

where we used the decorrelation of increments.

• Proceeding in a similar manner, we can show the greatest contribution to the second moment of the number of "steadily advancing" particles comes from pairs (*s*, *t*) whose lineages split early on.

First steps: Discretize and pass to Poisson model

$$X_N(t) = \log |\det(I_N - e(t)P_N)|, \ t \in \mathbb{R}/\mathbb{Z}.$$

• It is enough to show

$$\max_{t\in T_N} X_N(t) = (1+o(1))\log N \quad \text{w.h.p.}$$

where T_N is a mesh for \mathbb{R}/\mathbb{Z} of size $|T_N| = O(N)$.

• X_N only depends on the cycle structure of P_N :

$$X_N(t) = \sum_{1 \leq \ell \leq N} C_\ell(P_N) \log |1 - e(\ell t)|,$$

where $C_{\ell}(P_N)$ is the number of cycles of length ℓ in P_N .

• Arratia–Tavare '92: Let Z_{ℓ} be independent $\mathsf{Poi}(1/\ell)$ variables. Then

$$d_{\mathsf{TV}}((C_\ell(P_N))_{\ell\leq L}, (Z_\ell)_{\ell\leq L}) \to 0 \quad \text{ if } L = o(N).$$

First steps: Discretize and pass to Poisson model

• Let $\omega(1) \leq W \leq N^{o(1)}$ be a slowly growing function of N and split

$$egin{aligned} X_N(t) &= \sum_{\ell \leq N/W} C_\ell(P_N) \log |1 - e(\ell t)| + \sum_{N/W < \ell \leq N} C_\ell(P_N) \log |1 - e(\ell t)| \ &=: X_N^\leq(t) + X_N^>(t). \end{aligned}$$

Letting

$$Y_N(t) = \sum_{\ell \leq N} Z_\ell \log |1 - e(\ell t)|,$$

we have $X_N^{\leq}(t) \approx Y_{N/W}(t)$ in distribution.

Second moment computation shows

$$X_N^>(t) \leq \sum_{N/W < \ell \leq N} C_\ell(P_N) \leq O(\log W) = o(\log N)$$
 w.h.p.

so for the upper bound it suffices to consider the Poisson field Y_N .

Upper bound: first attempt

• Let $S(Y_N, x) = \{t \in T_N : Y_N(t) \ge x \log N\}$. We want to show that for any fixed $\epsilon > 0$,

$$|\mathcal{S}(Y_N, x_c + \epsilon)| = 0$$
 w.h.p.

First moment:

$$\mathbb{E} |\mathcal{S}(Y_N, x)| = \sum_{t \in \mathcal{T}_N} \mathbb{P}(Y_N(t) \ge x \log N).$$

• To estimate tail events, estimate MGFs:

$$\mathbb{E} e^{\beta Y_N(t)} = \prod_{\ell \le N} \mathbb{E} \exp(\beta Z_\ell \log |1 - e(\ell t)|) = \exp\left(\sum_{\ell \le N} \frac{1}{\ell} \left(|1 - e(\ell t)|^\beta - 1 \right) \right)$$

- For t irrational, the sum is $\sim (\log N) (\int_0^1 |1 e(u)|^\beta du 1).$
- This eventually shows $\mathbb{E} |\mathcal{S}(Y_N, \mathbf{1} + \epsilon)| = o(1)$.

Fix: Condition on a "typical" distribution of cycle lengths

- Problem: First moment is thrown off by "clumping" events, where there is an atypically large number of cycles of comparable length (the corresponding waves t → log |1 - e(ℓt)| add constructively).
- Solution: Consider a sequence of windows I_k = [e^{δk}, e^{δ(k+1)}) ⊂ [N] with δ small (actually o(1)) and put

$$\mathcal{N}(I_k) = \sum_{\ell \in I_k} Z_\ell, \qquad Y_{I_k}(t) = \sum_{\ell \in I_k} Z_\ell \log |1 - e(\ell t)|.$$

Expect $\mathbb{E}\mathcal{N}(I_k) \approx \delta$ cycles in each window. We condition on a "typical" sequence $(\mathcal{N}(I_k))_{k\geq 1}$ (with $\mathcal{N}(I_k) \in \{0,1\}$ for most k).

Let K = {k : N(I_k) = 1}. Then |K| ≈ log N, and for k ∈ K, the increments

$$Y_{I_k}(t) = \log |1 - e(\ell_k t)|, \qquad \mathbb{P}(\ell_k = \ell) \propto rac{1_{\ell \in I_k}}{\ell}$$

are still independent and have comparable contribution to the fluctuations of Y_N .

Upper bound

• Under the conditioning on $(\mathcal{N}(I_k))_k$ we have

$$\mathbb{E} e^{\beta Y_N(t)} \approx \prod_{k \in \mathcal{K}} \mathbb{E} \exp\left(\beta Y_{l_k}(t)\right) = \prod_{k \in \mathcal{K}} \left(\frac{1}{\sum_{\ell \in I_k} \frac{1}{\ell}} \sum_{\ell \in I_k} \frac{|1 - e(\ell t)|^{\beta}}{\ell}\right).$$

• If t is irrational then

$$\frac{1}{\sum_{\ell \in I_k} \frac{1}{\ell}} \sum_{\ell \in I_k} \frac{|1 - e(\ell t)|^{\beta}}{\ell} = \int_0^1 |1 - e(u)|^{\beta} du + o_{\beta,t}(1).$$

 As we'll be taking a union bound over t ∈ T_N, we need to quantify dependence of the error on t. We get satisfactory errors outside low-frequency Bohr sets

$$B_{\xi}(\eta) = \{t \in \mathbb{R}/\mathbb{Z} : \|\xi t\|_{\mathbb{R}/\mathbb{Z}} \leq \eta\}.$$

We call $t \in Maj(\xi_0, \eta) := \bigcup_{\xi \leq \xi_0} B_{\xi}(\eta)$ major arc points.

Upper bound: Major and minor arcs

• While $Y_N(t)$ is badly behaved on major arcs, we can show

$$\sup_{t\in {\sf Maj}(\xi_0,\eta)} Y_{{\sf N}}(t) \leq -c(\xi_0,\eta)\log^2 {\sf N} \quad {\sf w.h.p.}$$

• On the complementary set of minor arcs, Y_N behaves like a sequence of iid variables:

$$Y_{I_k}(t) = \log |1 - e(\ell_k t)| \approx \log |1 - e(\boldsymbol{u}_k)|, \qquad \boldsymbol{u}_k \sim \textit{Unif}(\mathbb{R}/\mathbb{Z}).$$

Letting λ(β) = log E exp (β log |1 − e(u)|) = log ∫₀¹ |1 − e(u)|^β du, for major arcs we have a first moment estimate for |S(Y_N, x) \ Maj |:

$$\sum_{t \in \mathcal{T}_N \setminus \mathsf{Maj}(\xi_0, \eta)} \mathbb{P}\left(Y_N(t) \ge x \log N\right) \approx |\mathcal{T}_N| \exp\left(-\lambda^*(x) |\mathcal{K}|\right) \approx N^{1-\lambda^*(x)}.$$

where λ^* is the Legendre transform of λ .

• $x_c \approx 0.677$ is the unique solution to $\lambda^*(x) = 1$.

Lower bound: Early, middle, and late generations

For $M \leq N$, denote the truncated field

$$X_{\leq M}(t) = \sum_{\ell \leq M} C_{\ell}(P_N) \log |1 - e(\ell t)|,$$

and the set of "survivors"

$$\mathcal{S}_{\leq M}(x) = \{t \in T_N : X_{\leq M}(t) \geq x \log M\}.$$

We show $S_{\leq N}(x_c - \epsilon)$ is nonempty by tracking the population of survivors across three epochs:

- 1. Early generations: $|S_{\leq N^c}(-C)| \gtrsim N$ w.h.p.
- 2. Middle generations: $|S_{\leq N/W}(x_c \epsilon)| \geq N^{c(\epsilon)}$ w.h.p.
- 3. Late generations: $|S_{\leq N}(x_c 2\epsilon)| \geq 1$ w.h.p.

For early and middle generations we can replace $X_{\leq M}$ with Y_M .

Lower bound: Middle generations

- We follow the second moment argument for BRW. Requires approximate decorrelation of tail events for increments $\Delta_{m,n}(t) := \sum_{k \in \mathcal{K} \cap [m,n)} Y_{l_k}(t).$
- For $\xi_0 \in \mathbb{N}$ put

$$d_{\xi_0}(s,t) = \min_{\xi,\xi'\in\{-\xi_0,\ldots,\xi_0\}\setminus\{0\}} \|\xi s+\xi't\|_{\mathbb{R}/\mathbb{Z}}.$$

Lower bound on $d_{\xi_0}(s, t)$ gives quantitative linear independence of 1, s, t over \mathbb{Z} .

• If $m = \omega(1)$, $s, t \notin Maj(\xi_0, \eta)$ and $d_{\xi_0}(s, t) > \eta$ for $\xi_0 = \omega(n^2)$ and $\eta \gg e^{-\delta m}$, then we have an estimate of the form

$$\mathbb{P}\left(\Delta_{m,n}(s),\Delta_{m,n}(t)\geq x
ight)\leq (1+o(1))\,\mathbb{P}(\Delta_{m,n}(s)\geq x)^{2}.$$

Lower bound: Late generations $(|S_{\leq N}(x_c - 2\epsilon)| \geq 1)$

- Poisson model is no longer available. We condition on the N^{c(ε)} survivors of Middle Ages S_{≤N/W} := S_{≤N/W}(x_c − ε) and want to show they aren't all wiped out by the high frequency tail X_N[>] = X_N − X_{≤N/W}.
- Note that $\log |1 e(\ell t)| \le -\epsilon \log N$ for t in the Bohr set

$$B_{\ell}(N^{-\epsilon}) = \{t \in \mathbb{R}/\mathbb{Z} : \|\ell t\|_{\mathbb{R}/\mathbb{Z}} \leq N^{-\epsilon}\}.$$

So if $S_{\leq N/W} \subset B_{\ell}(N^{-\varepsilon})$ for some $\ell \in (N/W, N]$, there is a chance the whole population gets wiped out.

- We employ a structural dichotomy for $S_{\leq N/W}$: Either
 - (Structured case) $S_{\leq N/W}$ has large overlap with $B_{\ell}(N^{-\epsilon})$ for $\geq N/W^3$ different frequencies $\ell \in (N/W, N]$, or
 - (Unstructured case) it doesn't.

In the unstructured case, we can show it is unlikely any of the exceptional frequencies are selected (using the switchings method).

(Structured case)

 $S_{\leq N/W}$ has large overlap with $B_{\ell}(N^{-\epsilon})$ for $\geq N/W^3$ different frequencies $\ell \in (N/W, N]$.

In this case we can use pigeonholing and a Vinogradov-type lemma to show the $S_{\leq N/W}$ must actually contain an element that is major arc, specifically an element of $B_{\xi}(\eta)$ for some

$$\xi = W^{O(1)}, \qquad \eta = W^{O(1)} N^{-\epsilon - 1}.$$

But such major arc points are unlikely to be in $S_{\leq N/W}$ in the first place.