

# On the eigenvalues of Brownian motion on $\mathbb{U}(N)$

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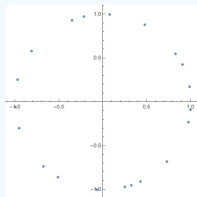
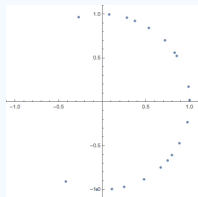
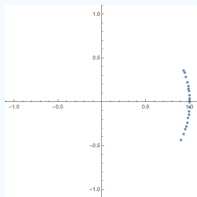
$$\langle A, B \rangle = N \operatorname{Tr}(B^* A) \quad A, B \in \mathfrak{u}(N).$$

We may define a standard Brownian motion  $\{U_t^N\}_{t \geq 0}$  on  $\mathbb{U}(N)$  to be a solution to

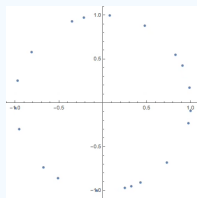
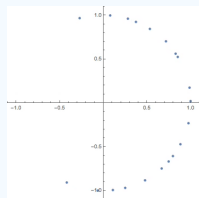
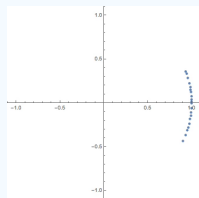
$$\begin{aligned} dU_t^N &= U_t^N \circ dW_t^N \\ &= U_t^N dW_t^N - \frac{1}{2} U_t^N dt \end{aligned}$$

with  $U_0^N = I_N$  and  $W_t^N$  a standard B.M. on  $\mathfrak{u}(N)$ .

# The empirical spectral measure



# The empirical spectral measure



Suppose that  $M$  is an  $n \times n$  random matrix with eigenvalues  $\lambda_1, \dots, \lambda_n$ .

The **empirical spectral measure**  $\mu$  of  $M$  is the (random) measure

$$\mu := \frac{1}{n} \sum_{k=1}^n \delta_{\lambda_k}.$$

# The empirical spectral measure

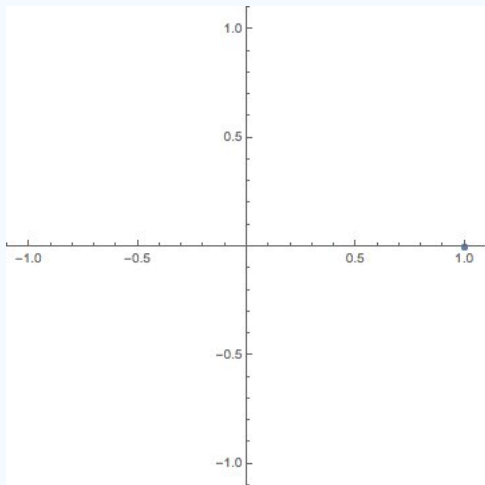
If  $\{U_t^N\}_{t \geq 0}$  is a Brownian motion on  $\mathbb{U}(N)$ , then for each  $t$ ,  $U_t^N$  has  $N$  eigenvalues

$$z_{t,1}, \dots, z_{t,N}$$

on the unit circle, and an associated spectral measure

$$\mu_t^N = \frac{1}{N} \sum_{j=1}^N \delta_{z_{t,j}}.$$

The process  $\{\mu_t^N\}_{t \geq 0}$





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Theorem (Biane, 1997)

*There is a deterministic family of measures  $\{\nu_t\}_{t \geq 0}$  on  $\mathbb{S}^1$  such that, for each  $t \geq 0$ , the spectral measure of  $U_t^N$  converges weakly almost surely to  $\nu_t$ :*

*for all  $f \in C(\mathbb{S}^1)$ ,*

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The measure  $\nu_t$  represents in some sense the spectral distribution of a “free unitary Brownian motion”.

The measures  $\nu_t$  are characterized in terms of their moments. They have densities (symmetric about 1) on the circle, and are supported on symmetric arcs until time  $t = 4$ , when their support becomes the whole circle.

# Non-asymptotic theory

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$L^1$ -Kantorovich distance:

For Borel probability measures  $\mu$  and  $\nu$  on a Polish space  $(\mathcal{X}, \rho)$ ,

$$W_1(\mu, \nu) = \inf \left\{ \int \rho(x, y) d\pi(x, y) : \pi \text{ a coupling of } \mu, \nu \right\}$$

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$$\begin{aligned} W_1(\mu, \nu) &= \inf \left\{ \int \rho(x, y) d\pi(x, y) : \pi \text{ a coupling of } \mu, \nu \right\} \\ &= \sup \left\{ \int f d\mu - \int f d\nu : |f|_{Lip} \leq 1 \right\}. \end{aligned}$$

# Levels of randomness



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Limit theorems for the ensemble-averaged spectral measure  $\mathbb{E}\mu_n$ :

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The quenched setting:

Almost sure (or a.a.s.) bounds on the random variable  $d(\mu_n, \nu)$

# Distance to the ensemble average

## Theorem (M.–Melcher)

Let  $\{U_t^N\}_{N \in \mathbb{N}, t \geq 0}$  be such that for each  $N$ ,  $U_t^N$  is a Brownian motion on  $\mathbb{U}(N)$ , with spectral measure  $\mu_t^N$ .

There is a constant  $C > 0$  such that with probability one, for all  $N \in \mathbb{N}$  sufficiently large and for all  $t > 0$ ,

$$W_1(\mu_t^N, \mathbb{E}\mu_t^N) \leq C \left( \frac{t}{N^2} \right)^{1/3}.$$

Moreover, for all  $N \in \mathbb{N}$  sufficiently large and all  $t \geq 8(\log N)^2$ ,

$$W_1(\mu_t^N, \mathbb{E}\mu_t^N) \leq \frac{C}{N^{2/3}}.$$

# Paths of measures

## Theorem (M.–Melcher)

There are constants  $c, C$  such that for any  $T \geq 0$  and for all  $x \geq c \frac{T^{2/5} \log(N)}{N^{2/5}}$ ,

$$\mathbb{P} \left( \sup_{0 \leq t \leq T} W_1(\mu_t^N, \nu_t) > x \right) \leq C \left( \frac{T}{x^2} + 1 \right) e^{-\frac{N^2 x^2}{T}}.$$

In particular, with probability one for  $N$  sufficiently large

$$\sup_{0 \leq t \leq T} W_1(\mu_t^N, \nu_t) \leq c \frac{T^{2/5} \log(N)}{N^{2/5}}.$$

# Concentration of measure

A standard argument using the fact that  $\mathbb{U}(N)$  has nonnegative Ricci curvature implies that for  $F : \mathbb{U}(N) \subseteq \mathbb{M}_N \rightarrow \mathbb{R}$  a 1-Lipschitz function with respect to  $\sqrt{N} \|\cdot\|_{H.S.}$ ,

$$\mathbb{P} \left[ |F(U_t^N) - \mathbb{E}F(U_t^N)| > r \right] \leq 2e^{-\frac{r^2}{t}}.$$

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On the other hand, at stationarity (i.e., if  $U$  is distributed according to Haar measure on  $\mathbb{U}(N)$ ), a clever coupling argument shows that

$$\mathbb{P}\left[|F(U) - \mathbb{E}F(U)| > r\right] \leq 2e^{-cr^2}.$$

# Concentration of measure

## Proposition (M.–Melcher)

There is a constant  $C > 0$  such that for all  $N \in \mathbb{N}$ ,  $t \geq 8(\log N)^2$  and  $r > 0$ , and all 1-Lipschitz functions  $F : \mathbb{U}(N) \rightarrow \mathbb{R}$ ,

$$\mathbb{P}(|F(U_t) - \mathbb{E}F(U_t)| > r) \leq Ce^{-\frac{r^2}{4}}.$$



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$$\mathbb{P}(|F(U_t) - \mathbb{E}F(U_t)| > r) \leq Ce^{-\frac{r^2}{4}}.$$

The proof takes advantage of

$$\left. \begin{array}{l} z_t = e^{\frac{ib_t}{N}} \\ \quad (b_t \text{ a standard BM on } \mathbb{R}) \\ \\ V_t \text{ a standard BM on } \mathbb{SU}(N) \\ \quad (\text{independent of } b_t) \end{array} \right\} \implies z_t V_t \text{ is a standard BM on } \mathbb{U}(N).$$

# Concentration of the empirical spectral measure

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If  $M$  is an  $n \times n$  normal matrix with spectral measure  $\mu_M$  and  $\nu$  is any reference measure,

$$M \mapsto W_1(\mu_M, \nu)$$

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$$\implies \mathbb{P}[W_1(\mu_t^N, \nu_t) > \mathbb{E}W_1(\mu_t^N, \nu_t) + r] \leq Ce^{-\frac{cN^2r^2}{(t)}}.$$


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 To show  $W_1(\mu_t^N, \nu_t)$  is typically small, it's enough to show that  $\mathbb{E}W_1(\mu_t^N, \nu_t)$  is small.

# Average distance to average

One approach: consider the stochastic process

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The concentration inequality for Lipschitz functions of  $U_t^N$  implies that  $\{X_f\}_f$  satisfies a sub-Gaussian increment condition:

$$\mathbb{P} [ |X_f - X_g| > r ] \leq 2e^{-\frac{cN^2 r^2}{|f-g|_L^2(t)}}.$$

Dudley's entropy bound together with approximation theory, truncation arguments, etc., leads to the bound

$$\begin{aligned} & \mathbb{E} W_1(\mu_t^N, \mathbb{E}\mu_t^N) \\ &= \mathbb{E} \left( \sup_{|f|_L \leq 1} X_f \right) \leq C \begin{cases} \left(\frac{t}{N^2}\right)^{1/3}, & \text{all } t > 0; \\ \left(\frac{1}{N^2}\right)^{1/3}, & t \geq 8(\log(N))^2. \end{cases} \end{aligned}$$

## Theorem (M.–Melcher)

Let  $\mu_t^N$  be the spectral measure of  $U_t$ , where  $\{U_t\}_{t \geq 0}$  is a Brownian motion on  $\mathbb{U}(n)$  with  $U_0 = I$ .

For any  $t, x > 0$ ,

$$\mathbb{P} \left( W_1(\mu_t^N, \mathbb{E}\mu_t^N) > c \left( \frac{t}{N^2} \right)^{1/3} + x \right) \leq 2e^{-\frac{N^2 x^2}{t}}.$$

For  $x > 0$  and  $t \geq 8(\log(N))^2$ ,

$$\mathbb{P} \left( W_1(\mu_t^N, \mathbb{E}\mu_t^N) > c \left( \frac{1}{N^2} \right)^{1/3} + x \right) \leq 2e^{-cN^2 x^2}.$$

Almost sure bounds on  $W_1(\mu_t^N, \mathbb{E}\mu_t^N)$  are immediate from the Borel–Cantelli lemma.



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Given  $f : \mathbb{S}^1 \rightarrow \mathbb{R}$  a 1-Lipschitz function, let

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Observe that

$$\int z^k d\mu_t^N(z) = \frac{1}{N} \left[ \text{Tr}((U_t)^k) \right],$$

so that

$$\begin{aligned} \left| \int S_m d\mathbb{E}\mu_t^N - \int S_m d\nu_t \right| &= \left| \sum_{1 \leq |k| < m} \hat{f}(k) \left( \frac{1}{N} \mathbb{E}[\text{Tr}(U_t^k)] - \int z^k d\nu_t \right) \right| \\ &\leq \sum_{1 \leq |k| < m} \frac{\pi}{2k} \left| \frac{1}{N} \mathbb{E}[\text{Tr}(U_t^k)] - \int z^k d\nu_t \right|. \end{aligned}$$

# Convergence of $\mathbb{E}\mu_t^N$ to $\nu_t$

Collins–Dahlqvist–Kemp '18:

$$\left| \frac{1}{N} \mathbb{E}[\text{Tr}(U_t^k)] - \int z^k d\nu_t \right| \leq \frac{t^2 k^4}{N^2}.$$

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Using this estimate together with the classical fact that

$$\|f - S_m\|_\infty \leq C \left( \frac{\log(m)}{m} \right)$$

and optimizing over  $m$  leads to

$$W_1(\mathbb{E}\mu_t^N, \nu_t) \leq C \frac{t^{2/5} \log N}{N^{2/5}}.$$

Convergence of paths: Continuity of  $\{\mathbb{E}\mu_t^N\}_{t \geq 0}$ :

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$$\begin{aligned} W_1(\mathbb{E}\mu_t^N, \mathbb{E}\mu_s^N) &= \sup_{|f|_L \leq 1} \mathbb{E} \left[ \int f d\mu_t^N - \int f d\mu_s^N \right] \\ &\leq \frac{\mathbb{E} \|U_t - U_s\|_N}{N} \\ &= \frac{\mathbb{E} \|I_N - U_{t-s}\|_N}{N}. \end{aligned}$$

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General properties of Brownian motion on manifolds together with estimates on volume ratios of balls in  $\mathbb{U}(N)$  yield a concentration inequality for  $\|I_N - U_{t-s}\|_N$  and ultimately,

$$W_1(\mathbb{E}\mu_t^N, \mathbb{E}\mu_s^N) \leq 3\sqrt{t-s} + \frac{1}{N}.$$



# Convergence of paths: Continuity of $\{\nu_t\}_{t \geq 0}$

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Using

- ▶ the established convergence of  $\mathbb{E}\mu_t^N$  to  $\nu_t$
- ▶ the continuity of  $\{\mathbb{E}\mu_t^N\}_{t \geq 0}$

if  $0 < s < t$ ,

$$\begin{aligned} W_1(\nu_t, \nu_s) &\leq W_1(\nu_t, \mathbb{E}\mu_t^N) + W_1(\nu_s, \mathbb{E}\mu_s^N) + W_1(\mathbb{E}\mu_t^N, \mathbb{E}\mu_s^N) \\ &\leq C \frac{(t^{2/5} + s^{2/5}) \log N}{N^{2/5}} + 3\sqrt{t-s} + \frac{1}{N}. \end{aligned}$$

Letting  $N \rightarrow \infty$  yields

$$W_1(\nu_t, \nu_s) \leq 3\sqrt{t-s}.$$

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Fix  $T > 0$ , let  $m \in \mathbb{N}$ , and for  $j = 1, \dots, m$ , let  $t_j := \frac{jT}{m}$ .

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$$x > 9\sqrt{\frac{T}{m}}$$

$$\begin{aligned} & \mathbb{P} \left( \sup_{0 \leq t \leq T} W_1(\mu_t^N, \nu_t) > x \right) \\ & \leq \mathbb{P} \left( \max_{1 \leq j \leq m} \sup_{|t-t_j| < \frac{T}{m}} W_1(\mu_t^N, \mu_{t_j}^N) > \frac{x}{3} \right) \\ & \quad + \mathbb{P} \left( \max_{1 \leq j \leq m} W_1(\mu_{t_j}^N, \nu_{t_j}) > \frac{x}{3} \right) \end{aligned}$$

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# Convergence of paths

## Theorem (M.–Melcher)

Let  $T \geq 0$ . There are constants  $c, C$  such that for all  $x \geq c \frac{T^{2/5} \log(N)}{N^{2/5}}$ ,

$$\mathbb{P} \left( \sup_{0 \leq t \leq T} W_1(\mu_t^N, \nu_t) > x \right) \leq C \left( \frac{T}{x^2} + 1 \right) e^{-\frac{N^2 x^2}{T}}.$$

*In particular, with probability one for  $N$  sufficiently large*

$$\sup_{0 \leq t \leq T} W_1(\mu_t^N, \nu_t) \leq c \frac{T^{2/5} \log(N)}{N^{2/5}}.$$

Thank you.

