

Section 6

Dyson Brownian Motion and Bulk Universality

I. Various notions of Bulk universality

The local eigenvalue statistics can either be expressed in terms of "local correlation functions" rescaled around some energy E or the "gap statistics" for a gap $\lambda_{j+1} - \lambda_j$ with a given label j . They are called "fixed energy" and "fixed gap" universalities, and they do not coincide. In fact, eigenvalues fluctuate on a scale much larger than the typical eigenvalue spacing, the label j of the eigenvalue λ_j closest to a fixed energy E is not a deterministic function of E . Moreover, the two concepts both have natural averaged versions, which are generally easier to establish.

Remark: Recall that $\int_{\delta_j}^{\delta_{j+1}} \rho_{sc}(x) dx = \frac{1}{N} \Rightarrow \delta_{j+1} - \delta_j \sim \frac{1}{N \rho_{sc}(\delta_j)}$. Hence, the ~~fluctuation~~ gaps and correlation functions need to be rescaled by the local density ρ_{sc} to get an universal limit. This holds in more general setups, such as sample covariance matrices.

① Fixed energy universality: $\forall n \in \mathbb{N}$, $F: \mathbb{R}^n \rightarrow \mathbb{R}$ is C_c^∞ . For any const $K > 0$, we have that uniformly in $E \in [-2+K, 2-K]$,

$$\lim_{N \rightarrow \infty} \frac{1}{\rho_{sc}(E)^n} \int_{\mathbb{R}^n} d\vec{\alpha} F(\vec{\alpha}) p_N^{(n)}\left(E + \frac{\vec{\alpha}}{N \rho_{sc}(E)}\right) = \int_{\mathbb{R}^n} d\vec{\alpha} F(\vec{\alpha}) q_{GOE/GUE}^{(n)}(\vec{\alpha}),$$

where $\vec{\alpha} = (\alpha_1, \dots, \alpha_n)$, $p_N^{(n)}$ is the n -point correlation function, and $q_{GOE/GUE}^{(n)}(\vec{\alpha})$

② Averaged bulk universality (on scale $N^{-1+\epsilon}$):

$$\lim_{N \rightarrow \infty} \frac{1}{\rho_{sc}(E)^n} \int_{E-b}^{E+b} \frac{dx}{2b} \int_{\mathbb{R}^n} d\vec{\alpha} F(\vec{\alpha}) p_N^{(n)}\left(x + \frac{\vec{\alpha}}{N \rho_{sc}(E)}\right) = \int_{\mathbb{R}^n} d\vec{\alpha} F(\vec{\alpha}) q_{GOE/GUE}^{(n)}(\vec{\alpha})$$

where $b = N^{-1+\epsilon}$ \forall const $\epsilon > 0$.

$= \det(S(\alpha_i - \alpha_j))_{i,j=1}^n$ is the determinant of sine-kernel we derived before.

③ Fixed gap universality: Fixed any small constant $\delta > 0$ and $n \in \mathbb{N}$. For any $F: \mathbb{R}^n \rightarrow \mathbb{R}$, $F \in C_c^\infty$ and any $k, m \in [\delta N, (1-\delta)N]$, we have that

$$\lim_{N \rightarrow \infty} \left| \mathbb{E}_{HN} F\left(N \rho_{sc}(\delta_k) (\lambda_{k+1} - \lambda_k), \dots, N \rho_{sc}(\delta_k) (\lambda_{k+n} - \lambda_k)\right) - \mathbb{E}_{GOE/GUE} F\left(N \rho_{sc}(\delta_m) (\lambda_{m+1} - \lambda_m), \dots, N \rho_{sc}(\delta_m) (\lambda_{m+n} - \lambda_m)\right) \right| = 0.$$

④ Averaged gap universality: For $l = N^\epsilon$, \forall const $\epsilon > 0$,

$$\lim_{N \rightarrow \infty} \left| \frac{1}{2l+1} \sum_{j=k-l}^{k+l} \mathbb{E}_{HN} F\left(N \rho_{sc}(\delta_k) (\lambda_{j+1} - \lambda_j), \dots, N \rho_{sc}(\delta_k) (\lambda_{j+n} - \lambda_j)\right) - \mathbb{E}_{GOE/GUE} F\left(N \rho_{sc}(\delta_m) (\lambda_{m+1} - \lambda_m), \dots, N \rho_{sc}(\delta_m) (\lambda_{m+n} - \lambda_m)\right) \right| = 0. \quad (49)$$

Rmk: Fixed energy \Rightarrow Averaged energy, Fixed gap \Rightarrow Averaged gap,

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We will focus on proving the averaged energy universality.

Theorem 6.1 The averaged energy universality holds on scales $N^{-1+\epsilon} \forall 0 < \epsilon < 1$.

This is a version of the famous "Wigner-Dyson-Mehta conjecture".

II. The three-step strategy

Step 1: Local semicircle law. At this step, we get precise estimates on the matrix elements of the resolvent, the rigidity of eigenvalues, and delocalization of eigenvectors.

Step 2: Universality for Gaussian divisible ~~ensembles~~ ensembles.

Gaussian divisible ~~se~~ ensembles are random matrices that can be written as $H_t = H_0 + \sqrt{t} H_0^G$, where H_0 is Wigner, H_0^G is GOE/GUE ~~and~~ independent of H_0 , and $t > 0$ is a parameter.

A convenient way to generate H_t is the "matrix Ornstein-Uhlenbeck (OU) Process":

$$dH_t = \frac{1}{\sqrt{N}} dB_t - \frac{1}{2} H_t dt, \quad H_t = H_0,$$

where B_t is a matrix Brownian motion whose entries are independent BMs up to symmetry $B_t^* = B_t$, and $\frac{1}{\sqrt{N}} B_t \stackrel{\text{law}}{=} \sqrt{t} \text{GOE/GUE}$. For each entry,

$$dh_{ij}(t) = \frac{1}{\sqrt{N}} db_{ij}(t) - \frac{1}{2} h_{ij}(t) dt.$$

It has a unique strong solution:

$$h_{ij}(t) = h_{ij}(0) e^{-t/2} + \int_0^t \frac{e^{-(t-t')/2}}{\sqrt{N}} db_{ij}(t').$$

Note that $\int_0^t \frac{e^{-(t-t')/2}}{\sqrt{N}} db_{ij}(t')$ is centered Gaussian of variance $\frac{1}{N} \int_0^t e^{-(t-t')} dt' = \frac{1}{N} (1 - e^{-t})$.

Hence, with a slight abuse of notation, we write it as $h_{ij}^G \cdot \sqrt{1 - e^{-t}}$.

This gives a solution

$$H_t \stackrel{\text{law}}{=} e^{-t/2} H_0 + \sqrt{1 - e^{-t}} H_0^G.$$

A big advantage of this form is that variances are preserved throughout the process:

$$\mathbb{E} |h_{ij}(t)|^2 = e^{-t} \mathbb{E} |h_{ij}(0)|^2 + (1 - e^{-t}) \mathbb{E} |h_{ij}^G|^2 = \mathbb{E} |h_{ij}(0)|^2,$$

for $\mathbb{E} |h_{ij}(0)|^2 = 1 + \delta_{ij}$ in the real case, and $\mathbb{E} |h_{ij}(0)|^2 = 2$ in the complex case.

The purpose of Step 2 is to show ~~that~~ the bulk universality of H_t for $t = N^{-1+\epsilon}$ for any $0 < \epsilon < 1$.

Approximation by a Gaussian ~~div~~ divisible ensemble

Step 3: Given a Wigner matrix, H , there exists a Wigner H_0 such that H_t has asymptotically ~~the~~ identical local eigenvalue statistics as H . This is usually done through a Green's function comparison argument by using certain moment matching conditions. Alternatively, one can also use a continuity estimate of ~~a~~ the matrix OU process.

The "three-step strategy" is now (one of) the most standard in proving the bulk universality of random matrices (for edge universality, the Step 2 sometimes is not necessary). Here, Step 1 is model-specific and generally is "hardest" step. The Steps 2 and 3 are more standard, ~~where~~ ^{where} general methods / proofs / arguments are known and work for "most" models. In particular, the strongest result for Step 2 has been established for very general initial conditions H_0 (not necessarily a random matrix).

III. Dyson Brownian Motion

The matrix Brownian motion introduces ~~the~~ SPDE tools to study the evolution of the eigenvalues of H_t : $\lambda_1(t) \leq \lambda_2(t) \leq \dots \leq \lambda_N(t)$. A classical theorem below will guarantee that the eigenvalues are simple and continuous functions of t . So the labelling is preserved along the evolution.

In principle, the eigenvalues $\{\lambda_i(t)\}$ and eigenvectors $\{\vec{u}_i(t)\}$ of H_t are correlated strongly, and we expect a couple system of stochastic differential equations for them (which is indeed the case if B_t is not chosen to have the law of \mathbb{R} GOE / GUE).

But, Dyson observe that the eigenvalues themselves satisfy an autonomous system of SDEs that does not involve eigenvectors, which is called the Dyson Brownian Motion (DBM).

Theorem 6.2 The eigenvalues $\{\lambda_i(t)\}$ of H_t satisfy the following system of SDEs:

$$d\lambda_i = \frac{\sqrt{2}}{\sqrt{\beta N}} dB_i(t) + \left(-\frac{\lambda_i}{2} + \frac{1}{N} \sum_{j \neq i} \frac{1}{\lambda_i - \lambda_j} \right) dt, \quad 1 \leq i \leq N, \quad \begin{cases} \beta=1 & \text{for real,} \\ \beta=2 & \text{for complex.} \end{cases}$$

$\{B_i(t) : 1 \leq i \leq N\}$ is a collection of independent BMs. The solution to the above equation is called ^a DBM (where there is not necessarily an underlying matrix model).

Proof: Let $\lambda_a^{(t)}$ be an eigenvalue of $H(t) = (h_{ij}(t))$ with eigenvector $\vec{u}_a(t)$. Almost surely, all eigenvalues are simple. We apply Ito's formula to $\lambda_a(t)$ to derive the DBM. We only consider the real case with $\beta=1$.

Differentiating: $H \vec{u}_a = \lambda_a \vec{u}_a$, $\vec{u}_a^* \vec{u}_\beta = \delta_{a\beta}$, we obtain that

$$(1) \quad \frac{\partial H}{\partial h_{ij}} \vec{u}_a + H \frac{\partial \vec{u}_a}{\partial h_{ij}} = \frac{\partial \lambda_a}{\partial h_{ij}} \vec{u}_a + \lambda_a \frac{\partial \vec{u}_a}{\partial h_{ij}},$$

$$(2) \quad \frac{\partial \vec{u}_\alpha^*}{\partial h_{ij}} \vec{u}_\beta + \vec{u}_\alpha^* \frac{\partial \vec{u}_\beta}{\partial h_{ij}} = 0, \quad \vec{u}_\alpha^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}} = 0.$$

Taking inner product with \vec{u}_α (1) and using (2), we get

$$\vec{u}_\alpha^* \frac{\partial H}{\partial h_{ij}} \vec{u}_\alpha + \lambda_\alpha \vec{u}_\alpha^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}} = \frac{\partial \lambda_\alpha}{\partial h_{ij}} + \lambda_\alpha \vec{u}_\alpha^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}}.$$

$$\Rightarrow \frac{\partial \lambda_\alpha}{\partial h_{ij}} = \vec{u}_\alpha^* \frac{\partial H}{\partial h_{ij}} \vec{u}_\alpha. \quad (*)$$

Taking inner product \vec{u}_β with (1) and using (2), we get (for $\beta \neq \alpha$)

$$\vec{u}_\beta^* \frac{\partial H}{\partial h_{ij}} \vec{u}_\alpha + \lambda_\beta \vec{u}_\beta^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}} = \lambda_\alpha \vec{u}_\beta^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}}.$$

$$\text{This implies that } \vec{u}_\alpha \frac{\partial \vec{u}_\alpha}{\partial h_{ij}} = \sum_{\beta \neq \alpha} \vec{u}_\beta (\vec{u}_\beta^* \frac{\partial \vec{u}_\alpha}{\partial h_{ij}}) = \sum_{\beta \neq \alpha} \vec{u}_\beta \frac{1}{\lambda_\alpha - \lambda_\beta} (\vec{u}_\beta^* \frac{\partial H}{\partial h_{ij}} \vec{u}_\alpha) \quad (†)$$

For (*), it writes: $\frac{\partial \lambda_\alpha}{\partial h_{ij}} = (2 - \delta_{ij}) u_{\alpha(i)} u_{\alpha(j)}$. For (†), it writes

$$\frac{\partial u_{\alpha(k)}}{\partial h_{ij}} = \sum_{\beta \neq \alpha} \frac{u_{\beta(i)} u_{\alpha(j)} + u_{\beta(j)} u_{\alpha(i)} (1 - \delta_{ij})}{\lambda_\alpha - \lambda_\beta} u_{\beta(k)}.$$

With these two formulas, we can also compute the second order partial derivatives:

$$\frac{\partial^2 \lambda_\alpha}{\partial h_{ik} \partial h_{ej}} = (2 - \delta_{ik}) \left[\frac{\partial u_{\alpha(i)}}{\partial h_{ej}} u_{\alpha(k)} + u_{\alpha(i)} \frac{\partial u_{\alpha(k)}}{\partial h_{ej}} \right]$$

$$= (2 - \delta_{ik}) \sum_{\beta \neq \alpha} \frac{1}{\lambda_\alpha - \lambda_\beta} \left[(u_{\beta(j)} u_{\alpha(l)} + u_{\beta(l)} u_{\alpha(j)} (1 - \delta_{lj})) u_{\beta(i)} u_{\alpha(k)} \right. \\ \left. + (u_{\beta(l)} u_{\alpha(j)} + u_{\beta(j)} u_{\alpha(l)} (1 - \delta_{lj})) u_{\beta(k)} u_{\alpha(i)} \right]$$

$$= (2 - \delta_{ik}) \sum_{\beta \neq \alpha} \frac{1}{\lambda_\alpha - \lambda_\beta} (u_{\beta(j)} u_{\alpha(l)} + u_{\beta(l)} u_{\alpha(j)} (1 - \delta_{lj})) (u_{\beta(i)} u_{\alpha(k)} + u_{\beta(k)} u_{\alpha(i)}).$$

Now, using Ito's formula, we obtain that

$$d\lambda_\alpha = \sum_{i,k} \frac{\partial \lambda_\alpha}{\partial h_{ik}} dh_{ik} + \frac{1}{2} \sum_{i,k} \sum_{j,l} \frac{\partial^2 \lambda_\alpha}{\partial h_{ik} \partial h_{jl}} [dh_{ik}, dh_{jl}]$$

$$= \sum_{i,k} u_{\alpha(i)} u_{\alpha(k)} \left[\frac{db_{ik}}{\sqrt{N}} - \frac{1}{2} h_{ik} dt \right] + \sum_{i,k} \sum_{\alpha \neq \beta} \frac{1}{2N} \frac{1}{\lambda_\alpha - \lambda_\beta} [u_{\beta(i)} u_{\alpha(k)} + u_{\beta(k)} u_{\alpha(i)}]^2 dt$$

$$= \frac{1}{\sqrt{N}} \sum_{i,k} u_{\alpha(i)} u_{\alpha(k)} db_{ik} - \frac{1}{2} \lambda_\alpha dt + \frac{1}{N} \sum_{\alpha \neq \beta} \frac{1}{\lambda_\alpha - \lambda_\beta} dt,$$

where we used $[dh_{ik}, dh_{ej}] = \frac{1}{N} \delta_{ie} \delta_{kj} (1 + \delta_{ik}) dt$, and $\sum_k h_{ik} u_{\alpha(k)} = \lambda_\alpha u_{\alpha(i)}$.

Now, we define a new Gaussian process: $\tilde{B}_\alpha := \sum_{i,k} u_{\alpha(i)} u_{\alpha(k)} B_{ik}$. Clearly, $\mathbb{E} \tilde{B}_\alpha = 0$.

We now calculate its covariance: