

Math 275D Lecture 25 Notes

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1 Dyson Brownian Motion

1.1 The drifting term in the rate of change of the eigenvalues

Let's say we have Hermitian matrices $H^t = H^0 + H_G^t$, where $H^t = (H^t)^\dagger$ and $H^0 = (H^0)^\dagger$. Let λ_k^t be the k -th largest eigenvalue of H^t and u_k^t be such that $\|U_k^t\|_2 = 1$ and $H^t u_k^t = \lambda_k^t u_k^t$ (eigenvector).

We let $[H_G^t]_{i,j} \sim B(t)$ for $i < j$ and $[H_G^t]_{i,i} \sim B(2t)$. The entries of H^t are independent. with $H_G^t = (H_G^t)^\dagger$. Then $H_{i,j}^t = H_{i,j}^0 + B_{i,j}(t)$ is a process for each $i \leq j$. So we can consider eigenvalues as a function of these processes: $\lambda_k^t(\{H_{i,j}\}_{i \leq j})$. What if we change the eigenvalues a little bit?

$$\begin{aligned} d\lambda_k^t &= \sum_{i,j} \frac{\partial \lambda_k}{\partial H_{i,j}^t} dB_{i,j}^t + \frac{1}{2} \sum_{i,j,i',j'} \frac{\partial^2 \lambda_k}{\partial H_{i,j} \partial H_{i',j'}} dB_{i,j} dB_{i',j'} \\ &= \nabla \lambda_k \cdot dH^t + dH^t \cdot \text{Hess}(\lambda_k) dH^t. \end{aligned}$$

Since these $N \times N$ matrices are Hermitian, we have N^2 processes which are not all independent.

Since $dB_{i,j} dB_{i',j'} = dt$ if $\{i,j\} = \{i',j'\}$ and 0 otherwise, we get

$$d\lambda_k^t = \sum_{i,j} \frac{\partial \lambda_k}{\partial H_{i,j}^t} dB_{i,j}^t + \frac{1}{2} \sum_{i,j} \left[\frac{\partial \lambda_k}{\partial H_{i,j}^2} + \frac{\partial^2 \lambda_k}{\partial H_{i,j} \partial H_{j,i}} \right] dt$$

There are results that say

$$\frac{\partial \lambda_k}{\partial H_{i,j}} = u_k(i)u_k(j), \quad \frac{\partial u_k}{\partial H_{i,j}} = \sum_{\ell \neq k} \frac{u_k(i)u_\ell(j)}{\lambda_\ell - \lambda_k} u_\ell.$$

So the right term, the “**drifting term**”, is

$$\sum_{\ell} \frac{-1}{\lambda_\ell - \lambda_k} dt.$$

So if λ_k is very close to λ_{k+1} , we get a very large negative term. So there is some large force to pull λ_k down, away from λ_{k+1} . This gives us a surprising property: the order of the eigenvalues never changes! That is, if we plot all the eigenvalues with t , the curves never intersect.

1.2 The diffusion term

Let's do some calculations with the first term, the “**diffusion term**”. This is

$$\sum_{i,j} u_k(i)u_k(j) dB_{i,j}(t)$$

Now define

$$B_k(t) = \int \sum_{i,j} u_k^t(i)u_k(j) dB_{i,j}(t).$$

This is a process, and we can check that for any fixed t , $B_k(t) \sim \mathcal{N}(0, t)$: We can calculate

$$\mathbb{E}[B_k^2(t)] = \sum_{i,j} u_k^t(i)u_k^t(j)u_k^t(i)u_k^t(j) \cdot t = t.$$

We can also calculate for $\ell \neq k$,

$$\mathbb{E}[B_k(t)B_\ell(t)] = \sum_{i,j} u_k^t(i)u_k^t(j)u_\ell(i)u_\ell(j) = 0$$

because the u_k, u_ℓ are orthogonal to each other. In fact, we can prove that

$$\mathbb{E}[B_k^m(t)] = \mathbb{E}[\mathcal{N}(0, t)^m].$$

Moreover, $B_k(\cdot)$ is a Brownian motion, and $\{B_k\}_{k=1}^N$ are independent.

So we have that

$$d\lambda_k = dB_k + \sum_{\ell} \frac{-1}{\lambda_\ell - \lambda_k}.$$

What if I define some other process

$$d\tilde{\lambda}_k = d\tilde{B}_k + \sum_{\ell} \frac{-1}{\tilde{\lambda}_\ell - \tilde{\lambda}_k}$$

with the same distribution? This doesn't come from a matrix, but we can still use it to study the distribution of λ_k .