

Random Matrix Theory: Lecture 6

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CINVESTAV

$A_n(t) = P(\text{there are no eigenvalues inside } (-t, t)) =$

$$1 + \sum_{k=1}^n \frac{(-1)^k}{k!} \int_{-t}^t \dots \int_{-t}^t \det[K_n(x_i, x_j)]_{i,j=1}^k dx_1 \dots dx_k$$

Rescaling with $1/\sqrt{n}$, as $n \rightarrow \infty$, $A_n(\theta)$ converges to

$$A(t) = 1 + \sum_{k=1}^{\infty} \frac{(-1)^k}{k!} \int_{-t}^t \dots \int_{-t}^t \det[k(x_i, x_j)]_{i,j=1}^k dx_1 \dots dx_k,$$

where

$$k(x, y) = \frac{\sin(x - y)}{\pi(x - y)}.$$

Fredholm Determinant

Consider the operator K defined through a kernel k

$$Kf(x) = \int_s^t k(x, y)f(y)dy,$$

acting on function in $L_2([s, t])$.

In the pursuit of solving the equation $f - Kf = h$, Fredholm proposed studying the expression

$$D = 1 + \sum_{k=1}^{\infty} \frac{(-1)^k}{k!} \int_s^t \dots \int_s^t \det[k(x_i, x_j)]_{i,j=1}^k dx_1 \dots dx_k.$$

This is now called the Fredholm determinant of K , denoted $\det(I - K)$, it is finite!.

The resolvent

It turns out that one can find the inverse of the operator $I - K$. Using the notation of Fredholm, define

$$k \begin{pmatrix} x_1 & \cdots & x_n \\ y_1 & \cdots & y_n \end{pmatrix} = \det[k(x_i, y_j)]_{i,j=1}^n.$$

Now, we can define the so-called resolvent kernel

$$r(x, y) := \sum_{k=0}^{\infty} \frac{(-1)^{k+1}}{k!} \int_s^t \cdots \int_s^t k \begin{pmatrix} x & x_1 & \cdots & x_k \\ y & x_1 & \cdots & x_k \end{pmatrix} dx_1 \cdots dx_k.$$

Lemma

It holds the following two identities

$$\begin{aligned} r(x, y) - Dk(x, y) &= \int_s^t k(x, z)r(z, y)dz \\ &= \int_s^t r(x, z)k(z, y)dz. \end{aligned}$$

Derivative of A

Notice that $\frac{dA_n(t)}{dt} =$

$$-2 \sum_{k=1}^n \frac{(-1)^k}{(k-1)!} \int_{-t}^t \cdots \int_{-t}^t K_n \begin{pmatrix} t & x_1 & \cdots & x_{k-1} \\ t & x_1 & \cdots & x_{k-1} \end{pmatrix} dx_1 \cdots dx_{k-1}$$

where

$$K_n \begin{pmatrix} x_1 & \cdots & x_n \\ y_1 & \cdots & y_n \end{pmatrix} = \det[K_n(x_i, y_j)]_{i,j=1}^n$$

Hence

$$\frac{dA}{dt} = \frac{dD}{dt} = -2r(t, t),$$

where r is the resolvent kernel associated to k .

If $R(x, y) = r(x, y)/D$,

$$\frac{d \log A}{dt} = -2R(t, t), \quad (1)$$

Aim: characterize $\sigma(t) = tR(t, t)$.

Towards a differential equation

Notation: $H = \int h(x, y)$ is the integral operator with kernel h .
Let $T = \int R(x, y)$. • the identity can be seen as an integral operator with kernel the delta Dirac

$$h(x) = \int \delta(x - y)h(y)dy$$

If $f(x) = \sin(x)/\sqrt{\pi}$ and $g(x) = \cos(x)/\sqrt{\pi}$, then

$$k(x, y) = \frac{f(x)g(y) - f(y)g(x)}{x - y}.$$

Also we define

$$Q(x) = (I + T)f(x) = f(x) + \int_{-t}^t R(x, y)f(y)dy \quad (2)$$

and

$$P(x) = (I + T)g(x) = g(x) + \int_{-t}^t R(x, y)g(y)dy. \quad (3)$$

Step 1: The commutator

$$[T_1, T_2] := T_1 T_2 - T_2 T_1$$

Proposition Let M be the multiplication operator by the argument, that is $Mf(x) = xf(x)$. Then $[M, H] = \int (x - y)h(x, y)$.

Proposition Suppose that $I - T_2$ is invertible, then it holds

$$[T_1, (I - T_2)^{-1}] = (I - T_2)^{-1}[T_1, T_2](I - T_2)^{-1}.$$

Corollary $[M, I + T] = (I + T)[M, K](I + T)$.

Theorem It holds that

$$R(x, y) = \frac{Q(x)P(y) - P(x)Q(y)}{x - y}.$$

Proposition

$$R(-t, t) = \frac{Q(t)P(t)}{t}.$$

Step 2: x, y derivatives

We work on the interval $[t_1, t_2]$.

Lemma It holds

$$-\left(\frac{dR}{dx} + \frac{dR}{dy}\right) = R(x, t_2)R(t_2, y) - R(x, t_1)R(t_1, y),$$

Proposition

$$Q'(x) = P(x) + R(x, t_1)Q(t_1) - R(x, t_2)Q(t_2)$$

and

$$P'(x) = -Q(x) + R(x, t_1)P(t_1) - R(x, t_2)P(t_2).$$

Hence

$$R(t, t) = P^2(t) + Q^2(t) - 2\frac{Q^2(t)P^2(t)}{t}$$

Step 3: t derivatives

Lemma

$$\frac{dR}{dt}(x, y) = R(x, t)R(t, y) + R(x, -t)R(-t, y).$$

Proposition

$$\frac{dR}{dt}(t, t) = 2R^2(-t, t).$$

Proposition

$$\frac{dQ(t)}{dt} = P(t) - 2\frac{P(t)Q^2(t)}{t}$$

and

$$\frac{dP(t)}{dt} = Q(t) + 2\frac{Q(t)P^2(t)}{t}.$$

Theorem

σ satisfies the Painlevé V differential equation

$$(t\sigma''(t))^2 = 8 ((\sigma'(t))^2 + 2t\sigma'(t) - 2\sigma(t)) (t\sigma'(t) - \sigma(t)).$$

Tracy-Widom distribution (F_2)

Using the other asymptotics for the Hermite polynomials, we can study the limit behaviour of eigenvalues around the edge of the semi-circle law, that around 2 or -2 .

It turns out that

$$\frac{1}{n^{1/6}} K_n \left(2\sqrt{n} + \frac{x}{n^{1/6}}, 2\sqrt{n} + \frac{y}{n^{1/6}} \right) \\ \sim \frac{Ai(x)Ai'(y) - Ai(y)Ai'(x)}{x - y}$$

And the distribution of the largest eigenvalue is a Fredholm determinant with this kernel:

$$\lim_{n \rightarrow \infty} n^{2/3} (\lambda_n / \sqrt{n} - 2) \sim F_2$$

F_2 , called the Tracy-Widom distribution, is determined by differential equation $\sigma''(t) = t\sigma(t) + 2\sigma^3(t)$, $\sigma(t) \sim Ai(t)$ as $t \rightarrow \infty$