

Random Matrix Theory: Lecture 2

Carlos G. Pacheco

CINVESTAV

Wishart ensemble

Wishart ensemble is $n \times n$ matrix given by $M_n = X_{m,n}^\top X_{m,n}$.
it also looks like

$$M_n v = \sum_{i=1}^n \langle q_i, v \rangle q_i.$$

$$q_i = \begin{pmatrix} X_n(i, 1) \\ \vdots \\ X_n(i, n) \end{pmatrix}. = i \text{ row (in column form)}$$

Also $\gamma_n(i)$ is the i column of X_n .

Again, we are interested in the empirical spectral distribution ν_n .

Now the correct scaling is $\frac{1}{n} M_n$, i.e. $X_n(i, j) \sim N(0, n^{1/4})$.

$$S_n(z) := \int_{-\infty}^{\infty} \frac{1}{x-z} \mu_n(dx), \quad z \in \mathbb{C}.$$

for $z \notin \mathbb{R}$, the eigenvalues of $[M_n - zI]^{-1}$ are $\frac{1}{\lambda_1 - z}, \dots, \frac{1}{\lambda_n - z}$, then

$$S_n(z) = \frac{1}{n} \text{trace} \left([M_n - zI]^{-1} \right).$$

Suppose that A is a $n \times n$ symmetric matrix and B be its upper-left $(n-1) \times (n-1)$ minor submatrix

$$A = \begin{bmatrix} B & v \\ v^\top & a \end{bmatrix},$$

Then,

$$\text{trace}A^{-1} - \text{trace}B^{-1} = \frac{1 + v^\top B^{-2}v}{a - v^\top B^{-1}v}.$$

Now, suppose that C is the $n \times n$ matrix

$$C = \begin{bmatrix} B & 0 \\ 0 & z \end{bmatrix},$$

where $z \in \mathbb{C}$ and 0 represents a column or row with only zeros.

Then,

$$\text{trace}C^{-1} = \text{trace}B^{-1} + \frac{1}{z}.$$

A set of matrices

Given M_n , define $A_n(k)$

$$A_n(k) = \begin{bmatrix} M_n(k) & 0 \\ 0 & 0 \end{bmatrix}.$$

$M_n(k)$ is the $k \times k$ minor of M_n .

For $t \in [0, 1]$ we define $s_n(z, t)$ such that when $t = \frac{k}{n} = t_n(k)$

$$s_n(z, t) := \frac{1}{n} \text{trace}([A_n(k) - zI]^{-1}).$$

$$\frac{s_n(z, t_n(k+1)) - s_n(z, t_n(k))}{t_n(k+1) - t_n(k)} \\ = \text{trace}([A_n(k+1) - zI]^{-1}) - \text{trace}([A_n(k) - zI]^{-1}).$$

Using the tricks

$$= \text{trace}([M_n(k+1) - zI]^{-1}) - \text{trace}([M_n(k) - zI]^{-1}) + \frac{1}{z}$$

and

$$\text{trace}([M_n(k+1) - zI]^{-1}) - \text{trace}([M_n(k) - zI]^{-1}) \\ = \frac{1 + v_n^\top(k) (M_n(k) - zI)^{-2} v_n(k)}{\alpha_n(k) - v_n^\top(k) (M_n(k) - zI)^{-1} v_n(k)}.$$

Elements of the equation

for $j = 1, 2$ the limit of

$$\begin{aligned} T_k^{(j)}(n) &= v_n^\top(k) (M_n(k) - zI)^{-j} v_n(k), \\ &= \gamma_n^\top(k+1) [\gamma_n(1), \dots, \gamma_n(k)] (M_n(k) - zI)^{-j} \begin{bmatrix} \gamma_n^\top(1) \\ \vdots \\ \gamma_n^\top(k) \end{bmatrix} \gamma_n(k+1). \end{aligned}$$

$\gamma_n(k)$ is the k column of X_n .

The all the variables in $\gamma_n(k+1)$ are independent from the rest of $T_k^{(j)}(n)$. Define $\mathcal{F}_{k,n}$ to be the information inside $T_k^{(j)}(n)$ without considering $\gamma_n(k+1)$.

Then

$$\begin{aligned} E[T_k^{(j)}(n) | \mathcal{F}_{k,n}] &= \frac{1}{n} \text{Trace} \left((M_n(k) - zI)^{-j} M_n(k) \right) \\ &= \frac{1}{n} \sum_{i=1}^k \frac{\lambda_i(k)}{(\lambda_i(k) - z)^j}, \end{aligned}$$

By Helly's theorem there exists a subsequence k_n such that

$$\frac{1}{n} \sum_{i=1}^{k_n} \frac{1}{(\lambda_i(k_n) - z)} \rightarrow s(z, t)$$

Limits of the elements 2

by adding and subtracting z we obtain that

$$\lim_n E[T_{k_n}^{(1)}(n) | \mathcal{F}_{k_n, n}] = 1 + zs(z, t)$$

and

$$\lim_n E[T_{k_n}^{(2)}(n) | \mathcal{F}_{k_n, n}] = z \frac{\partial s(z, t)}{\partial z} + s(z, t).$$

Moreover, also

$$\lim_n \text{Var}[T_{k_n}^{(j)}(n) | \mathcal{F}_{k_n, n}] = 0.$$

The final equation

We end up with

$$s_t(z, t) = -\frac{s_z(z, t)}{1 + s(z, t)},$$

and the initial condition is clearly $s(z, 0) = -\frac{1}{z}$.

To find $S(z) = s(z, 1)$, by the method of characteristics

$$S(z) = -\frac{1}{2} + \frac{\sqrt{(z-2)^2 - 4}}{2z}.$$

This Stieltjes transform comes from the density

$$f_{MP}(x) = \frac{\sqrt{(4-x)x}}{2\pi x}, \quad x \in [0, 4].$$