

Random matrix theory: lecture 9

1

Asymptotic analysis: first approach

Observation: the capacity of a  $n \times n$  MIMO channel whose channel matrix  $H$  has i.i.d.  $N_c(0,1)$  entries is given by

$$C = \mathbb{E} \left( \log \det \left( \mathbf{I} + \frac{P}{n} H H^* \right) \right)$$

Notice that  $\left( \frac{1}{n} H H^* \right)_{jj} = \frac{1}{n} \sum_{\ell=1}^n |h_{j\ell}|^2 \xrightarrow[n \rightarrow \infty]{} \mathbb{E}(|h_{jj}|^2) = 1$   
for each  $j$  fixed, by the law of large numbers

Similarly  $\left( \frac{1}{n} H H^* \right)_{jk} = \frac{1}{n} \sum_{\ell=1}^n h_{j\ell} \overline{h_{k\ell}} \xrightarrow[n \rightarrow \infty]{} \mathbb{E}(h_{j\ell} \overline{h_{k\ell}}) = 0 \quad \forall j \neq k$

Therefore,  $\frac{1}{n} H H^* \underset{n \rightarrow \infty}{\sim} \mathbf{I}$ , so

$$C \underset{n \rightarrow \infty}{\sim} \mathbb{E} \left( \log \det(\mathbf{I} + P \mathbf{I}) \right) = n \log(1+P) \quad ???$$

Why would we need random matrix analysis, then?

The problem is that the determinant (and likewise, the eigenvalues)

is not a continuous function of the entries, as the

size of the matrix goes to infinity. Therefore, even though

$\frac{1}{n} H H^*$  converges entrywise to  $\mathbf{I}$  as  $n \rightarrow \infty$ ,

$\det(\mathbf{I} + \frac{P}{n} H H^*)$  diverges from  $\det(\mathbf{I} + P \mathbf{I})$ .

Here is another simple <sup>(counter-)</sup> example: let  $A = (1 - \frac{1}{n}) \mathbf{I}$  of size  $n \times n$

Then  $A \underset{n \rightarrow \infty}{\sim} \mathbf{I}$ , but  $\det A = \left(1 - \frac{1}{n}\right)^n \underset{n \rightarrow \infty}{\sim} \frac{1}{e} \neq \det \mathbf{I} = 1$ .

## General formulation of the problem

Let  $(A^{(n)})_{n \geq 1}$  be a sequence of random matrices of increasing size ( $A^{(n)}$  is of size  $n \times n$ ),  $\lambda_1^{(n)} \dots \lambda_n^{(n)}$  denote the (ordered or un-ordered) eigenvalues of  $A^{(n)}$  and let  $\lambda^{(n)}$  be one eigenvalue picked uniformly at random among  $\lambda_1^{(n)} \dots \lambda_n^{(n)}$ . Let also  $p^{(n)}(\lambda)$  denote the distribution of  $\lambda^{(n)}$ . We are interested in the limiting behaviour of  $p^{(n)}(\lambda)$  as  $n \rightarrow \infty$ .

First remarks: for most random matrix ensembles, a rescaling of the matrices in  $n$  is needed in order to obtain the convergence of  $p^{(n)}(\lambda)$ .

### a) (Square) complex Wishart Ensemble

Let  $H^{(n)}$  have i.i.d.  $N_{\mathbb{C}}(0,1)$  entries and  $A^{(n)} = H^{(n)}(H^{(n)})^*$ .  
( $n \times n$  matrix)

What is the first moment of  $p^{(n)}(\lambda)$ ?

$$\begin{aligned} \int_0^{\infty} \lambda p^{(n)}(\lambda) d\lambda &= \mathbb{E}(\lambda^{(n)}) = \mathbb{E}\left(\frac{1}{n}(\lambda_1^{(n)} + \dots + \lambda_n^{(n)})\right) \\ &= \mathbb{E}\left(\frac{1}{n} \text{Tr}(A^{(n)})\right) = \frac{1}{n} \sum_{j,k=1}^n \underbrace{\mathbb{E}(|h_{jk}^{(n)}|^2)}_{=1} = \frac{1}{n} \cdot n^2 = n \end{aligned}$$

This heuristics indicates that in order to obtain the convergence of at least the first moment of  $p^{(n)}(\lambda)$ , we need to scale down the matrix  $A^{(n)}$  by a factor  $\frac{1}{n}$  (note that this heuristics coincides by the way with the "wrong" analysis of page 1, saying that  $\frac{1}{n} H^{(n)} (H^{(n)})^*$  converges entrywise to  $I$ ).

In the case where  $H$  is a <sup>rectangular</sup>  $n \times m$  matrix, where  $\frac{m}{n} = c$  is a fixed parameter, the same heuristics shows that we should scale down the matrix  $A^{(n)}$  by a factor  $\frac{1}{n}$ .

## b) GUE

Let  $A^{(n)} = H^{(n)}$  have independent entries  $h_{jk}^{(n)}$ ,  $j \leq k$ , with  
( $n \times n$  matrix)  
 $h_{jj}^{(n)} \sim N_{\mathbb{R}}(0, 1)$ ,  $h_{jk}^{(n)} \sim N_{\mathbb{C}}(0, 1)$  and  $h_{kj}^{(n)} = \overline{h_{jk}^{(n)}}$

First moment of  $p^{(n)}(\lambda)$ :

$$\begin{aligned} \int_{\mathbb{R}} \lambda p^{(n)}(\lambda) d\lambda &= \mathbb{E}(\lambda^{(n)}) = \mathbb{E}\left(\frac{1}{n} \sum_{j=1}^n \lambda_j^{(n)}\right) = \mathbb{E}\left(\frac{1}{n} \text{Tr} A^{(n)}\right) \\ &= \frac{1}{n} \sum_{j=1}^n \underbrace{\mathbb{E}(h_{jj}^{(n)})}_0 = 0 \end{aligned}$$

So far, it therefore seems that no rescaling is needed, but...

Second moment of  $p^{(n)}(\lambda)$ :

$$\begin{aligned} \int_{\mathbb{R}} \lambda^2 p^{(n)}(\lambda) d\lambda &= \mathbb{E}(\lambda^{(n)2}) = \mathbb{E}\left(\frac{1}{n} \sum_{j=1}^n (\lambda_j^{(n)})^2\right) \\ &= \mathbb{E}\left(\frac{1}{n} \text{Tr}\left(\underbrace{(A^{(n)})^2}_{= A^{(n)}(A^{(n)})^*}\right)\right) = \frac{1}{n} \sum_{j,k=1}^n \underbrace{\mathbb{E}(|h_{jk}^{(n)}|^2)}_{=1} = n \end{aligned}$$

According to this heuristics, we therefore need to scale down the matrix  $A^{(n)}$  by a factor  $\frac{1}{\sqrt{n}}$ .

c) CUE

Let  $A^{(n)}$  be distributed according to the Haar dist. on  $U(n)$ .  
( $n \times n$  matrix)

In this case, we have:

$$\begin{aligned} \int_0^{2\pi} |e^{i\theta}|^2 p^{(n)}(\theta) d\theta &= \mathbb{E}(|\lambda^{(n)}|^2) = \mathbb{E}\left(\frac{1}{n} \sum_{j=1}^n |\lambda_j^{(n)}|^2\right) \\ &= \mathbb{E}\left(\frac{1}{n} \text{Tr}(A^{(n)}(A^{(n)})^*)\right) = \mathbb{E}\left(\frac{1}{n} \text{Tr} I\right) = 1 \end{aligned}$$

Therefore, no rescaling is needed. Moreover,  $p^{(n)}(\theta) = \frac{1}{2\pi}$

is the uniform distribution on  $[0, 2\pi)$  for all  $n$ ,

so the asymptotic analysis is particularly simple!

(\*) NB: it is clear that this integral is equal to 1, since  $|e^{i\theta}| = 1$  &  $p^{(n)}(\theta)$  is a p.d.f.!

Aside: asymptotic analysis of the second order

marginal distribution in the CUE

$$\text{We have: } p^{(n)}(\theta, \varphi) = \frac{1}{4\pi^2} \cdot \frac{n}{n-1} \left( 1 - \left( \frac{\sin\left(\frac{n(\theta-\varphi)}{2}\right)}{n \sin\left(\frac{\theta-\varphi}{2}\right)} \right)^2 \right)$$

Since there are  $n$  eigenvalues on the unit circle of (fixed) length  $2\pi$ , the natural scaling for the spacings of eigenvalues is of order  $\frac{1}{n}$ . Let us

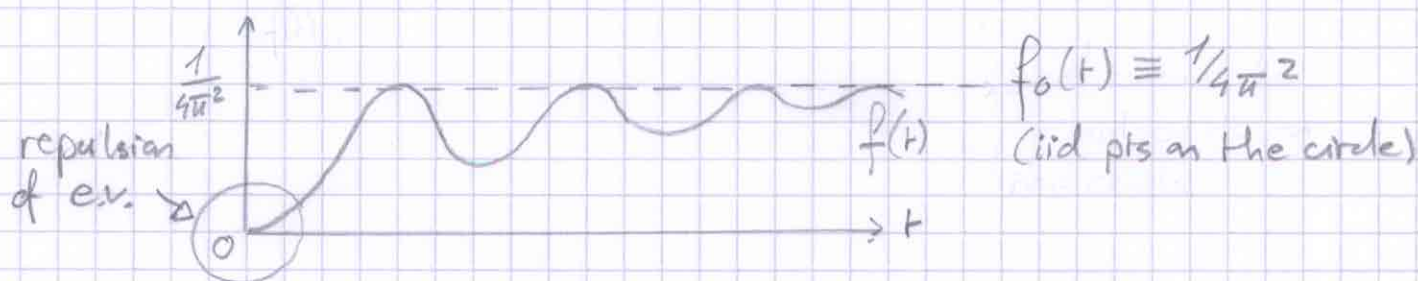
therefore define  $\Theta = \varphi + \frac{t}{n}$ :

$$p^{(n)}\left(\varphi + \frac{t}{n}, \varphi\right) = \frac{1}{4\pi^2} \cdot \frac{n}{n-1} \left( 1 - \frac{(\sin(t/2))^2}{(n \sin(t/2n))^2} \right)$$

$$\underset{n \rightarrow \infty}{\sim} \frac{1}{4\pi^2} \left( 1 - \frac{(\sin(t/2))^2}{(t/2)^2} \right) := f(t)$$

"two-point correlation function"

Since  $n \sin(x/n) \rightarrow x$  as  $n \rightarrow \infty$ .



A similar behaviour near zero can be obtained for the probability that two neighbouring ev. are separated by a distance  $t/n$  (but the analysis is more difficult).

Back to the first order marginal, in the GUE:

We had computed, for  $A^{(n)} = H^{(n)}$ :

$$p^{(n)}(\lambda) = \sum_{\ell=0}^{n-1} H_{\ell}(\lambda)^2 e^{-\lambda^2/2} \quad \text{with } H_{\ell} = \text{Hermite polynomials}$$

Considering now the rescaled matrices  $\tilde{A}^{(n)} = \frac{1}{\sqrt{n}} H^{(n)}$ , we obtain:

$$\tilde{p}^{(n)}(\lambda) = \sqrt{n} p^{(n)}(\sqrt{n}\lambda) = \frac{1}{\sqrt{n}} \sum_{\ell=0}^{n-1} H_{\ell}(\sqrt{n}\lambda)^2 e^{-n\lambda^2/2}$$

Analyzing this expression further requires a good knowledge

of Hermite polynomials! By the Christoffel-Darboux formula,

we obtain:

$$\tilde{p}^{(n)}(\lambda) = \left( \sqrt{n} H_n(\sqrt{n}\lambda)^2 - \sqrt{n+1} H_{n+1}(\sqrt{n}\lambda) H_{n-1}(\sqrt{n}\lambda) \right) e^{-n\lambda^2/2}$$

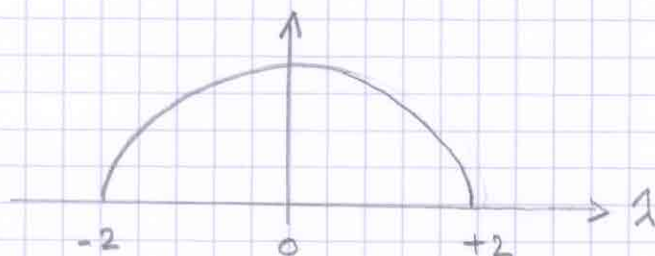
} Next, the Plancherel-Rotach formula gives:

$$\left. \begin{array}{l} \\ \end{array} \right\} (2n)^{1/4} H_n(\sqrt{n}\lambda) e^{-n\lambda^2/4} = \text{some } f(\lambda) + o\left(\frac{1}{n}\right)$$

... we "therefore" have:

$$\lim_{n \rightarrow \infty} \tilde{p}^{(n)}(\lambda) = \frac{1}{2\pi} \sqrt{4 - \lambda^2} \cdot \mathbb{1}_{|\lambda| \leq 2}$$

Wigner's semi-circle law



## Remarks

- The above "proof" is quite technical.
- It generalizes to other random matrix ensembles, but requires different formulas for different orthogonal polynomials
- It relies heavily on the fact that both  $p^{(n)}(\lambda_1, \dots, \lambda_n)$  and the marginal  $p^{(n)}(\lambda)$  are computable at finite  $n$ .

We will see that much more powerful techniques allow to strengthen the result in two main directions:

- the result can be shown for more general random matrix ensembles (for which  $p^{(n)}(\lambda)$  is unknown at finite  $n$ )
- the convergence of  $p^{(n)}(\lambda)$  to a limit as  $n \rightarrow \infty$  is a convergence "in expectation" <sup>(\*)</sup>; it can be shown that an "almost sure" convergence also holds for random matrices.

$$\begin{aligned}
 (*) \quad \mathbb{E} \left( \frac{1}{n} \sum_{j=1}^n f(\lambda_j^{(n)}) \right) &= \mathbb{E} \left( f(\lambda^{(n)}) \right) \\
 &= \int_{\mathbb{R}} f(\lambda) p^{(n)}(\lambda) d\lambda \xrightarrow{n \rightarrow \infty} \int_{\mathbb{R}} f(\lambda) p(\lambda) d\lambda
 \end{aligned}$$