Lecures on Statistical Field Theory Random Matrices and Statistical Physics – Homework 1

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The Wigner semi-circular law via replicas

We consider an $N \times N$ random matrix **J** drawn from the Gaussian Orthogonal Ensemble (GOE), i.e. **J** is a real symmetric matrix such that, for $k \leq l$, the entries J_{kl} are independent Gaussian entries of zero mean and variances $\mathbb{E}(J_{kk}^2) = 2/N$ and $\mathbb{E}(J_{kl}^2) = 1/N$ for k < l. We denote by $\lambda_1, \lambda_2, \dots, \lambda_N$ the N real eigenvalues of **J** and the goal is to compute the average density of eigenvalues

$$\rho(x) = \mathbb{E}\left[\frac{1}{N}\sum_{i=1}^{N}\delta(x-\lambda_i)\right],\qquad(1)$$

in the limit $N \to \infty$. In the following, ϵ denotes a strictly positive real number $\epsilon > 0$. You will find, at the end, some results on Gaussian integrals, which may be useful in the following computations.

1 Preamble

1.) Using the identity, valid for any real x in the sense of distribution,

$$\lim_{\epsilon \to 0^+} \frac{1}{x - i\epsilon} = \operatorname{PV}\frac{1}{x} + i\,\pi\delta(x) , \qquad (2)$$

where PV denotes the Cauchy principal value, show that

$$\rho(x) = -\frac{1}{N\pi} \lim_{\epsilon \to 0^+} \operatorname{Im} \left(\mathbb{E} \left[\operatorname{Tr}(\mathbf{J} - x_{\epsilon} \mathbb{I}_N)^{-1} \right] \right) , \ x_{\epsilon} = x - i\epsilon ,$$
(3)

where \mathbb{I}_N denotes the identity matrix of size N.

2.) Consider the N-fold Gaussian integral

$$Z_J(x_{\epsilon}) = \int_{-\infty}^{\infty} \mathrm{d}\phi_1 \cdots \int_{-\infty}^{\infty} \mathrm{d}\phi_N \exp\left[-\frac{i}{2}x_{\epsilon}\sum_{k=1}^N \phi_k^2 + \frac{i}{2}\sum_{k,l} J_{kl} \phi_k \phi_l\right]$$
(4)

$$= \int_{\mathbb{R}^N} \mathrm{d}\vec{\phi} \, \exp\left[-\frac{1}{2}\vec{\phi}^T (ix_\epsilon \mathbb{I}_N - i\,\mathbf{J})\vec{\phi}\right].$$
(5)

Justify that $Z_J(x_{\epsilon})$ is well defined and show, using results for Gaussian integrals, that

$$\operatorname{Tr}(\mathbf{J} - x_{\epsilon} \mathbb{I}_N)^{-1} = 2 \frac{\partial}{\partial x} \ln Z_J(x_{\epsilon}) .$$
(6)

2 Annealed computation

3.) To compute $\rho(x)$, we will first make the approximation $\mathbb{E}[\ln Z_J(x_{\epsilon})] \approx \ln(\mathbb{E}[Z_J(x_{\epsilon})])$. When do you expect this approximation to be valid?

4.) Show that the averaged "partition function" $\mathbb{E}[Z_J(x_{\epsilon})]$ can be written as

$$\mathbb{E}[Z_J(x_{\epsilon})] = \int_{-\infty}^{\infty} \mathrm{d}\phi_1 \cdots \int_{-\infty}^{\infty} \mathrm{d}\phi_N \exp\left[-\frac{i}{2} x_{\epsilon} \sum_{k=1}^N \phi_k^2 - \frac{1}{4N} \left(\sum_{k=1}^N \phi_k^2\right)^2\right].$$
(7)

5.) By using the standard Gaussian identity (18) for m = 1 to disentangle the ϕ_k variables show that $\mathbb{E}[Z_J(x_{\epsilon})]$ can be written as

$$\mathbb{E}[Z_J(x_{\epsilon})] = \sqrt{\frac{N}{\pi}} \int_{-\infty}^{\infty} \mathrm{d}q \,\mathrm{e}^{-N\,\varphi_x(q)} \,, \, \varphi_x(q) = q^2 - \frac{1}{2} \ln\left(\frac{2\pi}{i(x_{\epsilon} - 2q)}\right) \,. \tag{8}$$

6.) Explain why, to compute the density $\rho(x)$ in the large N limit, we only need to retain terms in (8) which are of order $\mathcal{O}(e^N)$.

7.) At this order of accuracy, the integral over q in (8) can be evaluated, for large N, by the Laplace (or saddle-point) method. To find this saddle point, we need to solve $\partial_q \varphi_x(q) = 0$. Show that this equation has actually two roots $q = q_+$ or $q = q_-$ given by

$$q_{\pm} = \frac{1}{4} \left(x_{\epsilon} \pm \sqrt{x_{\epsilon}^2 - 4} \right) . \tag{9}$$

8.) It can be shown that the correct saddle point is given by $q = q^+$. Deduce then that in the limit $N \to \infty$ one has

$$\lim_{N \to \infty} \rho(x) = \frac{2}{\pi} \lim_{\epsilon \to 0^+} \operatorname{Im} \frac{\partial}{\partial x} \left[\varphi_x(q_+) \right] \,. \tag{10}$$

9.) Obtain finally that, when $N \to \infty$, $\rho(x) = \sqrt{4 - x^2}/(2\pi)$ if $|x| \le 2$ while $\rho(x) = 0$ if x > 2.

3 Quenched computation (More complicated)

To compute the average $\mathbb{E}[\ln Z_J(x_{\epsilon})]$, going beyond the approximation $\mathbb{E}[\ln Z_J(x_{\epsilon})] \approx \ln (\mathbb{E}[Z_J(x_{\epsilon})])$, we will use the replica trick which we find convenient here to write as

$$\mathbb{E}[\ln Z_J(x_{\epsilon})] = \lim_{n \to 0} \frac{1}{n} \ln \left(\mathbb{E}[Z_J(x_{\epsilon})^n] \right) .$$
(11)

10.) Use this replica trick (11) to write $\rho(x)$ in terms of $\mathbb{E}[Z_J(x_{\epsilon})^n]$.

11.) By performing the averages over the random variables J_{kl} , show that

$$\mathbb{E}[Z_J(x_{\epsilon})^n] = \int_{-\infty}^{\infty} \prod_{k=1}^N \prod_{a=1}^n \mathrm{d}\phi_k^a \exp\left[-\frac{i}{2} x_{\epsilon} \sum_{k=1}^N \sum_{a=1}^n (\phi_k^a)^2 - \frac{1}{4N} \sum_{a,b=1}^n \left(\sum_{k=1}^N \phi_k^a \phi_k^b\right)^2\right].$$
 (12)

12.) By using n(n+1)/2 times the Gaussian identity (18) show that

$$\mathbb{E}[Z_J(x_{\epsilon})^n] = A_{N,n} \int_{-\infty}^{\infty} \prod_{k=1}^N \prod_{a=1}^n \mathrm{d}\phi_k^a \prod_{1=a \le b \le n} \mathrm{d}Q_{ab} \\ \times \exp\left[-N \sum_{a,b=1}^n (Q_{ab})^2 - \frac{i}{2} x_{\epsilon} \sum_{k=1}^N \sum_{a=1}^n (\phi_k^a)^2 + i \sum_{a,b=1}^n Q_{ab} \sum_{k=1}^N \phi_k^a \phi_k^b\right]$$
(13)

where $A_{N,n} = 2^{n(n-1)/4} \left(\frac{N}{\pi}\right)^{n(n+1)/4}$, with the convention, in (13), that $Q_{ab} = Q_{ba}$ when a > b.

13.) Observe that the *n*-dimensional vectors $\{\phi_k^1, \phi_k^2, \dots, \phi_k^n\}$, for different values of $k = 1, \dots, N$, are now independent on this expression (13), and show that

$$\mathbb{E}[Z_J(x_\epsilon)^n] = A_{N,n} \prod_{1 \le a \le b \le n} \int_{-\infty}^{\infty} \mathrm{d}Q_{ab} \exp\left[-N\Phi_x(\mathbf{Q})\right]$$
(14)

where **Q** is the (symmetric) overlap matrix, of size $n \times n$, with matrix elements Q_{ab} for $a \leq b$ (and thus $Q_{ab} = Q_{ba}$ if a > b), while $\Phi_x(\mathbf{Q})$ is given by

$$\Phi_x(\mathbf{Q}) = \operatorname{Tr}[\mathbf{Q}^2] - \frac{1}{2} \operatorname{Tr}\left[\ln\left(\frac{2\pi}{i} \left(x_\epsilon \,\mathbb{I}_n - 2\mathbf{Q}\right)^{-1}\right)\right] , \qquad (15)$$

where \mathbb{I}_n denotes the identity matrix of size n.

14.) In the large N limit, this multiple integral (14) can be evaluated by the Laplace (or saddle-point) method. Show that the saddle point equations read

$$2Q_{ab} - (x_{\epsilon}\mathbb{I}_n - 2\mathbf{Q})_{ab}^{-1} = 0 , \quad \forall \ a, b = 1, \cdots, n .$$
(16)

<u>**Hint:</u></u> we recall that, for an invertible matrix Y** depending on a parameter z, one has $\partial_z(\det \mathbf{Y}) = (\det \mathbf{Y}) \operatorname{Tr}(\mathbf{Y}^{-1}\partial_z \mathbf{Y}).$ </u>

15.) Show that these equations (16) admit the (replica symmetric) solution $Q_{ab} = q \,\delta_{ab}$ with $q = q_{\pm}$ as obtained before (9).

16.) Using the ansatz $Q_{ab} = q_+ \delta_{ab}$, assuming that this is the correct saddle-point solution to (16), compute the density $\rho(\lambda)$ in the large N limit and show that it gives back the Wigner semi-circle.

Some Gaussian integrals

We will find useful the following Gaussian integrals which hold for an invertible symmetric squared matrix **A** of size $m \times m$ (not necessarily real, provided the corresponding integrals are well defined) and a complex vector $\vec{b} \in \mathbb{C}^m$:

$$\int_{\mathbb{R}^m} \mathrm{d}\vec{x} \,\mathrm{e}^{-\frac{1}{2}\vec{x}^T \mathbf{A}\vec{x}} = \frac{(2\pi)^{m/2}}{\sqrt{\det \mathbf{A}}} \,, \tag{17}$$

$$\int_{\mathbb{R}^m} \mathrm{d}\vec{x} \,\mathrm{e}^{-\frac{1}{2}\vec{x}^T \mathbf{A}\vec{x} + \vec{b}^T \vec{x}} = \frac{(2\pi)^{m/2}}{\sqrt{\det \mathbf{A}}} \mathrm{e}^{\frac{1}{2}\vec{b}^T \mathbf{A}^{-1}\vec{b}} \tag{18}$$

$$\int_{\mathbb{R}^m} \mathrm{d}\vec{x} \, x_k x_l \, \mathrm{e}^{-\frac{1}{2}\vec{x}^T \mathbf{A}\vec{x}} = \frac{(2\pi)^{m/2}}{\sqrt{\det \mathbf{A}}} (\mathbf{A})_{kl}^{-1} \,. \tag{19}$$