Exercise 1: Klein's lemma. Let $f : \mathbb{R} \to \mathbb{R}$ be convex, and \mathcal{H}_N the set of $N \times N$ Hermitian matrices. Prove that

$$
\left\{ \begin{array}{ll} \mathcal{H}_N & \to \mathbb{R} \\ H & \mapsto \text{Tr}\, f(H) \end{array} \right.
$$

is convex.

Exercise 2: concentration of the Stieltjes transform. Let H be a random matrix from the $N \times N$ Gaussian orthogonal ensemble with the usual normalization, i.e. it has density $e^{-\frac{N}{4}\text{Tr}(H^2)}$ with respect to the Lebesgue measure on the symmetric matrices. Prove that there exists $c > 0$ such that for any $t > 0$, $N \ge 1$ and $z = E + i\eta \in \mathbb{H}$ we have

$$
\mathbb{P}\left(\left|\frac{1}{N}\text{Tr}\frac{1}{H-z} - \mathbb{E}\frac{1}{N}\text{Tr}\frac{1}{H-z}\right| \ge t\right) \le c^{-1}e^{-cN^2t^2\eta^4}
$$

.

Same question when H is symmetric with independent entries (in the upper triangle) bounded with $N^{-1/2}$.

Exercise 3: concentration of individual eigenvalues. Let H be a $N \times N$ symmetric random matrix such that $|H_{ij}| \leq 1$ for any i, j. The goal of this exercise is to prove that for any $1 \leq k \leq N$ and $t > 0$ we have

$$
\mathbb{P}(|\lambda_k - M(\lambda_k)| \ge t) \le 4e^{-\frac{t^2}{32k^2}},
$$

where $\lambda_N \leq \cdots \leq \lambda_1$, the λ_k 's are the eigenvalues of H and $M(\lambda_k)$ is the median of λ_k .

(i) Let S be the set of $N \times N$ symmetric random matrices, $t > 0$, $\mathcal{A} = \{A \in \mathcal{S} : \lambda_k(A) \leq M(\lambda_k)\}\$ and $\mathcal{B} = \{B \in \mathcal{S} : \lambda_k(B) \geq M(\lambda_k) + t\}.$ Let $(v_\ell)_{1 \leq \ell \leq N}$ be the normalized eigenvectors of some $B \in \mathcal{B}$ $(Bv_{\ell} = \lambda_k(B)v_{\ell})$ and

$$
\alpha_{ii} = \sum_{\ell=1}^k |v_\ell(i)|^2, \quad \alpha_{ij} = 2\left(\sum_{\ell=1}^k |v_\ell(i)|^2\right)^{1/2} \cdot \left(\sum_{\ell=1}^k |v_\ell(j)|^2\right)^{1/2}.
$$

Prove that $\sum_{1 \leq i \leq j \leq k} \alpha_{ij}^2 \leq 2k^2$.

- (ii) Prove that for any $A \in \mathcal{A}$ we have $\sum_{1 \leq i \leq j \leq N: A_{ij} \neq B_{ij}} \alpha_{ij} \geq \frac{t}{2}$.
- (iii) Conclude.

(iv) Open problem: is there an absolute c such that $\mathbb{P}(|\lambda_k - M(\lambda_k)| \ge t) \le c^{-1}e^{-ct^2}$ for any $k, t > 0$?

Exercise 4: eigenvalues distribution for the Ginibre ensemble. Read and reproduce the proof of Theorem 15.1.1 in Log gases and random matrices, to obtain the following fact. If G has density $e^{-\text{Tr}(GG*)}$ with respect to the Lebesgue density on $N \times N$ matrices (complex entires, no symmetry assumption), then writing $X = UTU^*$ for its Schur form, the matrix T has density $(\lambda_i = T_{ii})$ and T_{ij} , $i < j$ are the only non-zero entries) w.r.t the Lebesgue measure proportional to

$$
\prod_{i < j} |\lambda_i - \lambda_j|^2 e^{-\sum_{i=1}^N |\lambda_i|^2} e^{-\sum_{i < j} |T_{ij}|^2}.
$$

In particular the spectrum has density $\prod_{i < j} |\lambda_i - \lambda_j|^2 e^{-\sum_{i=1}^N |\lambda_i|^2}$.

Problem 1: the Selberg integral. The purpose of this problem is to calculate the partition function of Gaussian β -ensembles $(\beta \geq 0)$, i.e. proving that

$$
Z_N^{(\beta)} := \int_{\mathbb{R}^N} |\Delta(\lambda_1, \dots, \lambda_N)|^{\beta} e^{-N\frac{\beta}{4} \sum_{i=1}^N \lambda_i^2} d\lambda_1 \dots d\lambda_N = (2\pi)^{N/2} \left(\frac{\beta N}{2}\right)^{-\frac{N(N-1)\beta}{4} - \frac{N}{2}} \prod_{j=1}^N \frac{\Gamma(1+j\frac{\beta}{2})}{\Gamma(1+\frac{\beta}{2})},\tag{1}
$$

where $\Delta(\lambda_1,\ldots,\lambda_N) = \prod_{1 \leq i < j \leq N} (\lambda_i - \lambda_j)$ is the Vandermonde determinant. Here, $\Gamma(z) = \int e^{-t}t^{z-1}$ for $\Re(z) > 0$. First, we will prove the Selberg integral formula: for any $\gamma_1, \gamma_2 > -1$ and $\gamma \ge 0$,

$$
S_N(\gamma_1, \gamma_2, \gamma) := \int_{[0,1]^N} \left(\prod_{i=1}^N t_i \right)^{\gamma_1} \left(\prod_{i=1}^N (1 - t_i) \right)^{\gamma_2} |\Delta(t_1, \dots, t_N)|^{2\gamma} dt_1 \dots dt_N
$$

=
$$
\prod_{j=0}^{N-1} \frac{\Gamma(1 + \gamma_1 + j\gamma)\Gamma(1 + \gamma_2 + j\gamma)\Gamma(1 + (j+1)\gamma)}{\Gamma(2 + \gamma_1 + \gamma_2 + (N + j - 1)\gamma)\Gamma(1 + \gamma)}.
$$
 (2)

(i) Prove the Euler integral formula:

$$
\int_{[0,1]} t^{\gamma_1} (1-t)^{\gamma_2} dt = \frac{\Gamma(1+\gamma_1)\Gamma(1+\gamma_2)}{\Gamma(2+\gamma_1+\gamma_2)},
$$

by writing $\Gamma(1+\gamma_1)\Gamma(1+\gamma_2)$ as a double integral and making an appropriate change of variables. (ii) In question (ii) to (vii), assume $\gamma \in \mathbb{N}$. Prove that

$$
S_N(\gamma_1, \gamma_2, \gamma) = \sum_{0 \le n_1, \dots, n_N \le 2\gamma(N-1)} c_{n_1, \dots, n_N} \prod_{j=1}^N \frac{\Gamma(1 + \gamma_1 + n_j)\Gamma(1 + \gamma_2)}{\Gamma(2 + \gamma_1 + \gamma_2 + n_j)}
$$

for some coefficients $c_{n_1,...,n_N}$ independent of γ_1 and γ_2 .

(iii) Prove that if $c_{n_1,...,n_N} \neq 0$ then $\sum_{i=1}^N n_i = N(N-1)\gamma$. Assuming additionally that $n_1 \leq \cdots \leq n_N$, prove that for any $j \in [\![1,N]\!]$ we have

$$
(j-1)\gamma \le n_j \le (N+j-2)\gamma.
$$

For the first inequality, you can first consider $j = N$ and then observe that $\Delta(t_1, \ldots, t_j)$ divides $\Delta(t_1,\ldots,t_N)$. For the second inequality, you can write $\Delta(t_1,\ldots,t_j)$ in terms of $\Delta(t_1^{-1},\ldots,t_j^{-1})$. (iv) Prove that

$$
S_N(\gamma_1, \gamma_2, \gamma) = \frac{P(\gamma_1, \gamma_2)}{Q(\gamma_2)} \prod_{j=0}^{N-1} \frac{\Gamma(1+\gamma_1+j\gamma)\Gamma(1+\gamma_2+j\gamma)\Gamma(1+(j+1)\gamma)}{\Gamma(2+\gamma_1+\gamma_2+(N+j-1)\gamma)\Gamma(1+\gamma)}.
$$

where P and Q are polynomials with the same degree in γ_2 .

- (v) By symmetry in γ_1 and γ_2 , prove that P/Q is actually a constant $c(\gamma, N)$.
- (vi) By ordering $t_1 \leq \cdots \leq t_N$ and conditioning on t_N , prove that

$$
S_N(0,0,\gamma) = \frac{1}{\gamma(N-1)+1} S_{N-1}(0,2\gamma,\gamma)
$$

- (vii) Conclude that (2) holds for any $\gamma \in \mathbb{N}$.
- (viii) Prove that (2) holds for any $\gamma > 0$. You can assume the following theorem by Carlson.

If f is analytic on $\Re(z) \geq 0$, vanishes on N and $f(z) = O(e^{\mu z})$ with $\mu < \pi$, then $f = 0$ on $\Re(z) \geq 0$.

(ix) Prove (1). Hint: $e^{-c\lambda^2} = \lim_{L \to \infty} (1 - \lambda/L)^{cL^2} (1 + \lambda/L)^{cL^2}$.

Problem 2. Loop equations and linear statistics for the Gaussian Unitary Ensemble. Consider the probability distribution of eigenvalues from the Gaussian Unitary Ensemble:

$$
\mu(\mathrm{d}\lambda) = \frac{1}{Z_N} \prod_{1 \le k < \ell \le N} |\lambda_k - \lambda_\ell|^2 e^{-\frac{N}{2} \sum_{k=1}^N \lambda_k^2} \mathrm{d}\lambda_1 \dots \mathrm{d}\lambda_N
$$

on the simplex $\lambda_1 < \cdots < \lambda_N$. For a smooth $f : \mathbb{R} \to \mathbb{R}$ supported on $(-2+\kappa, 2-\kappa)$ $(\kappa > 0)$ we consider the general linear statistics $S_N(f) = \sum_{k=1}^N f(\lambda_k) - N \int f(s) \varrho(s) ds$, where $\varrho(s) = (2\pi)^{-1} \sqrt{(4-s^2)_+}$. We want to prove the weak convergence of $S_N(f)$ to a Gaussian random variable for large N, with no need of any normalization.

We are interested in the Fourier transform $Z(u) = \mathbb{E}_{\mu}(e^{iuS_N(f)})$. We will need a complex modification of the GUE, namely $d\mu^{u}(\lambda) = \frac{e^{iuS_N(f)}}{Z(u)}$ $\frac{d^{uS_N(t)}}{Z(u)} d\mu(\lambda)$, assuming that $Z(u) \neq 0$. Let $s_N(z) = \frac{1}{N} \sum_k \frac{1}{z - \lambda_k}$ and $m_{N,u}(z) = \mathbb{E}^{\mu^u}(s_N(z))$. The Stieltjes transform of the semicircle distribution is $m(z) = \int \frac{\varrho(s)}{z-s} ds =$ $\frac{z-\sqrt{z^2-4}}{2}$, where the square root is chosen so that m is holomorphic on $[-2,2]^c$ and $m(z) \to 0$ as $|z| \to \infty$.

(i) We denote the first correlation function $\varrho_1^{(N,u)}$, i.e. this is the unique continuous function such that $\mathbb{E}^{\mu^u}[\sum_i f(\lambda_i)] = N \int f \varrho_1^{(N,u)}$ for any continuous, bounded f. Prove that

$$
(m_{N,u}(z)-m(z))^2-\sqrt{z^2-4}\,(m_{N,u}(z)-m(z))+\frac{\mathrm{i}u}{N}\int_{\mathbb{R}}\frac{f'(s)}{z-s}\varrho_1^{(N,u)}(s)\mathrm{d} s=-\mathrm{var}_{\mu^u}\,(s_N(z)).
$$

This is called the (first) loop equation. To derive it, you may first prove that

$$
m_{N,u}(z)^{2} + \int_{\mathbb{R}} \frac{-s + iuN^{-1}f'(s)}{z - s} \varrho_1^{(N,u)}(s) ds = -\text{var}_{\mu^{u}}(s_{N}(z)).
$$

Hint: integrate by parts or change variables $\lambda_k = y_k + \varepsilon (\Re(\ell) \Im \mathfrak{m}) \frac{1}{z-y_k}$ and note $\partial_{\varepsilon=0} \log Z(u) = 0$. (ii) Remember the rigidity for Wigner matrices, in particular for GUE: for any $\xi, D > 0$ there exists

 $C > 0$ such that uniformly in $N \geq 1$ and $k \in [1, N]$ we have $\mu\left(|\lambda_k - \gamma_k| > N^{-\frac{2}{3} + \xi}(\hat{k})^{-\frac{1}{3}} \right) \leq$ CN^{-D} , where $\int_{-\infty}^{\gamma_k} \varrho(s \, ds) = \frac{k}{N}$ and $\hat{k} = \min(k, N + 1 - k)$. Assume $Z(u) \neq 0$. Prove that

$$
|\mu^u| \left(|\lambda_k - \gamma_k| > N^{-\frac{2}{3} + \xi} (\hat{k})^{-\frac{1}{3}} \right) \le C \frac{N^{-D}}{|Z(u)|}
$$

where $|\mu^u|$ is the total variation of the complex measure μ^u . Conclude that uniformly in $z = E + i\eta$, $-2 + \kappa < E < 2 - \kappa$, $0 < |\eta| < 1$, we have

,

.

$$
|\text{var}_{\mu^u}(s_N(z))| = \mathcal{O}\left(\frac{N^{-2+2\xi}}{\eta^2 |Z(u)|^2}\right)
$$

(iii) Prove that uniformly in $-2 + \kappa < E < 2 - \kappa$, $N^{-1+\xi} \leq \eta \leq 1$, we have

$$
m_{N,u}(z) - m(z) = \frac{1}{\sqrt{z^2 - 4}} \frac{iu}{N} \int_{\mathbb{R}} \frac{f'(s)}{z - s} \varrho(s) \mathrm{d}s + \mathcal{O}\left(\frac{N^{-2 + 3\xi}}{\eta^2 |Z(u)|^2}\right).
$$

(iv) Let $\chi : \mathbb{R} \to \mathbb{R}^+$ be a smooth function such that $\chi(y) = 1$ for $|y| < 1/2$ and $\chi(y) = 0$ for $|y| > 1$. Prove that for any $\lambda \in \mathbb{R}$, we have

$$
f(\lambda) = -\frac{1}{2\pi} \iint_{\mathbb{R}^2} \frac{iy f''(x)\chi(y) + i(f(x) + iy f'(x))\chi'(y)}{x + iy - \lambda} dxdy,
$$

where the right hand side converges absolutely. For this, you can reproduce the proof of Cauchy's integral formula based on Green's theorem, considering the quasi-analytic extension $(f(x) +$ $\mathrm{i} y f'(x) \chi(y)$.

(v) Note that $\partial_u \log Z(u) = \mathbb{E}_{\mu^u}(\mathrm{i}S_N(f))$. Conclude that bulk linear statistics converge to a Gaussian random variable.