## Random Matrix Theory, homework 3, due December 22nd.

Exercise 1. Eigenvalues dynamics for covariance matrices. Let M = M(t) be a  $N \times N$  random matrix with  $(M_{i,j}(t))_{t\geq 0}$  a real-valued, standard Brownian motion, and denote  $H(t) = M(t)^{\mathrm{T}} M(t)$ . Assume that M(0) is such that H(0) has N distinct eigenvalues  $\lambda_1(0) > \cdots > \lambda_N(0) > 0$ . Prove that the process  $(\lambda(t))_{t\geq 0}$  satisfies

$$d\lambda_i(t) = 2\sqrt{\lambda_i(t)}dB_i(t) + Ndt + \sum_{k \neq i} \frac{\lambda_i(t) + \lambda_k(t)}{\lambda_i(t) - \lambda_k(t)}dt, \quad 1 \le i \le N,$$

where the  $B_i$ 's are independent, standard Brownian motions. As a first step you can provide a formal derivation of the above result. As a second step you can prove that almost surely, a process satisfying this SDE has non-colliding particles.

**Exercise 2.** Non-collision for arbitrary driving noise. Consider the stochastic differential equation

$$dx_i(t) = \frac{db_i(t)}{\sqrt{2N}} + \frac{1}{2N} \sum_{j \neq i} \frac{1}{x_i(t) - x_j(t)} dt, \quad 1 \le |i| \le N,$$

where  $0 < x_1(0) < \cdots < x_N(0), x_{-i}(0) = -x_i(0), (i \ge 1), (b_i)_{1 \le i \le N}$  is a collection of continuous martingales, and  $b_{-i}(t) = -b_i(t), (i \ge 1)$ . Note that the drift above includes a repulsive term between  $x_i$  and  $x_{-i}$ , equal to  $\frac{1}{2N} \frac{1}{2x_i}$ .

Assume that  $\frac{d\langle b_i \rangle_t}{dt} \leq 1$  for any  $i \geq 1, t \geq 0$ . Prove existence and strong uniqueness for the above stochastic differential equation.

## Exercise 3. Andreiev's identity, Kostlan's theorem.

- (i) On a measured space  $(E, \mathcal{E}, \mu)$ , prove that for any functions  $(\phi_i, \psi_i)_{i=1}^N \in L_2(\mu)^{2N}$ ,  $\frac{1}{N!} \int_{E^N} \det(\phi_i(\lambda_j)) \, \det(\psi_i(\lambda_j)) \, \mu(\mathrm{d}\lambda_1) \dots \mu(\mathrm{d}\lambda_N) = \det(f_{i,j}) \quad \text{where } f_{i,j} = \int_E \phi_i(\lambda) \psi_j(\lambda) \mu(\mathrm{d}\lambda).$
- (ii) Let  $E = \mathbb{C}$ , *m* the Lebesgue measure on *E*,  $g \in L_2(\mu)$ , with  $\mu(d\lambda) = \frac{N}{\pi} e^{-N|\lambda|^2} m(d\lambda)$ , and  $\{\lambda_1, \ldots, \lambda_N\}$  be the eigenvalues from the complex Ginibre ensemble (with the standard normalization for a limiting circular law on  $\mathbb{D}$ ). Prove that

$$\mathbb{E}\left(\prod_{k=1}^{N} g(\lambda_k)\right) = N^{\frac{N(N-1)}{2}} \det(f_{i,j})_{i,j=1}^{N} \quad \text{where } f_{i,j} = \frac{1}{(j-1)!} \int \lambda^{i-1} \bar{\lambda}^{j-1} g(\lambda) \mu(\mathrm{d}\lambda).$$

(iii) Prove that the set  $N\{|\lambda_1|^2, \ldots, |\lambda_N|^2\}$  is distributed as  $\{\gamma_1, \ldots, \gamma_N\}$ , a set of (unordered) independent Gamma variables of parameters  $1, 2, \ldots, N$ .

**Exercise 4. Moment matching.** Let X be a real-valued random variable with mean zero, variance 1. To simplify, we assume that X takes values in [-10, 10]. Prove that for any  $\gamma < 10^{-10}$  there exists a random variable  $Y_{\gamma}$ , valued in [-100, 100], such that

$$Z_{\gamma} := \sqrt{1 - \gamma} Y_{\gamma} + \sqrt{\gamma} G$$

and X have the same moments of order 1, 2, 3, and

$$|\mathbb{E}[Z_{\gamma}^4 - X^4]| \le 10^{10} \gamma.$$

Here G is a standard Gaussian random variable, independent of  $Y_{\gamma}$ .