

*Isaac Newton Institute for Mathematical Sciences*  
*RMAw02 - Recent Perspectives in Random Matrix Theory and Number Theory*

Gaussian ensembles of random matrices III  
*Y. Fyodorov (Brunel)*  
*30 March 2004*

Isaac Newton Institute for Mathematical Sciences  
20 Clarkson Road, Cambridge CB3 0EH, UK

Tel: +44 1223 335999      Fax: +44 1223 330508  
E-mail: [webseminars@newton.cam.ac.uk](mailto:webseminars@newton.cam.ac.uk)  
<http://www.newton.cam.ac.uk/webseminars>

## Characterization of spectral sequences

Let  $-\infty < \lambda_1, \dots, \lambda_N < \infty$  be the positions of  $N$  points on the real axis characterized by joint p.d.f

$$P(\lambda_1, \dots, \lambda_N)$$

of having, regardless of labelling one point  $[\lambda_1, \lambda_1 + d\lambda_1]$ , another in  $[\lambda_2, \lambda_2 + d\lambda_2]$  another in  $[\lambda_N, \lambda_N + d\lambda_N]$

The statistical properties of the sequence  $\lambda_i$ 's are characterized by  $n$ -point correlation functions

$$R_n(\lambda_1, \dots, \lambda_n) = \frac{N!}{(N-n)!} \int P(\lambda_1, \dots, \lambda_N) \underbrace{d\lambda_{n+1} \dots d\lambda_N}_{N-n}$$

These are related to statistics of the number  $N_B$  of points  $\lambda_i$  within any set  $B$  on the real axis.

Denote  $\chi_B(x) = \begin{cases} 1 & x \in B \\ 0 & x \notin B \end{cases}$  and  $\rho_N(\lambda) = \sum_{i=1}^N \delta(\lambda - \lambda_i)$   
 characteristic function      density of points

$$\text{Then } N_B = \int \chi_B(\lambda) \rho_N(\lambda) d\lambda \equiv N_B \{ \lambda_i \}$$

The expectation value  $N_B$  can be found as

$$\begin{aligned} N_B &\equiv \int d\lambda_1 \dots d\lambda_N P(\lambda_1, \dots, \lambda_N) N_B \{ \lambda_i \} \\ &= \int d\lambda \chi_B(\lambda) \int d\lambda_1 \dots d\lambda_N P(\lambda_1, \dots, \lambda_N) \sum_{i=1}^N \delta(\lambda - \lambda_i) \\ &= \int d\lambda \chi_B(\lambda) \cdot \underbrace{N \int d\lambda_2 \dots d\lambda_N P(\lambda_1, \lambda_2, \lambda_3, \dots, \lambda_N)}_{R_1(\lambda)} \end{aligned}$$

$$\text{Thus } \overline{N_B} = \int_B d\lambda R_1(\lambda) \Rightarrow \overline{\rho_N(\lambda)} \equiv R_1(\lambda) \quad \text{Mean density}$$

$$\text{Similarly } N_B^2 \int d\lambda d\lambda' \chi_B(\lambda) \chi_B(\lambda') \sum_{i,j=1}^N \delta(\lambda - \lambda_i) \delta(\lambda' - \lambda_j) \\ \int d\lambda d\lambda' \chi_B(\lambda) \chi_B(\lambda') \left\{ \delta(\lambda - \lambda') \sum_{i=1}^N \delta(\lambda - \lambda_i) + \sum_{i \neq j} \delta(\lambda - \lambda_i) \delta(\lambda' - \lambda_j) \right\} \\ N_B + \int d\lambda d\lambda' \chi_B(\lambda) \chi_B(\lambda') \sum_{i \neq j} \delta(\lambda - \lambda_i) \delta(\lambda' - \lambda_j)$$

Averaging over  $\mathcal{P}(\lambda_1, \dots, \lambda_N)$  immediately gives

$$N_B^2 = \overline{N_B^2} + \int d\lambda d\lambda' \chi_B(\lambda) \chi_B(\lambda') \cdot \underbrace{N(N-1) \int d\lambda_2 \dots d\lambda_N \mathcal{P}(\lambda, \lambda', \lambda_2, \dots, \lambda_N)}_{R_2(\lambda, \lambda')}$$

Thus  $\overline{N_B^2} = \overline{N_B} + \int_{B \times B} R_2(\lambda, \lambda') d\lambda d\lambda'$ , and the variance

of the number of points  $n_B$  is given by

$$\Sigma_2(B) = \overline{N_B^2} - \overline{N_B}^2 = N_B \int_{B \times B} Y_2(\lambda, \lambda') d\lambda d\lambda'$$

where we defined the so-called cluster function

$$Y_2(\lambda, \lambda') = R_1(\lambda) R_1(\lambda') - R_2(\lambda, \lambda')$$

Finally one more frequently used characteristic of the spectrum is called the hole probability

the probability that no points exist in the interval  $(L/2, L/2)$  It is given by

$$A(L) = \int d\lambda_1 \dots d\lambda_N \mathcal{P}(\lambda_1, \dots, \lambda_N) \prod_{k=1}^N [1 - \chi_L(\lambda_k)]$$

$$\sum_{j=0}^N (-1)^j \int d\lambda_1 \dots d\lambda_N \mathcal{P}(\lambda_1, \dots, \lambda_N) \underbrace{h_j[\chi_L(\lambda_1), \dots, \chi_L(\lambda_N)]}_{\text{symmetric functions}}$$

Symmetric functions:  $h_0(x_1, \dots, x_N) = 1$ ,

$$h_1(x_1, \dots, x_N) = \sum_{i=1}^N x_i, \quad h_2(x_1, \dots, x_N) = \sum_{i < j} x_i x_j,$$

$$h_N(x_1, \dots, x_N) = x_1 x_2 \dots x_N.$$

We see that  $h_j(\dots)$  contains  $\binom{N}{j} = \frac{N!}{j!(N-j)!}$  terms, and

$$\int \int d\lambda_1 \dots d\lambda_N \mathcal{P}(\lambda_1, \dots, \lambda_N) h_j[\gamma_1(\lambda_1), \dots, \gamma_N(\lambda_N)] \\ = \frac{N!}{j!(N-j)!} \cdot \frac{(N-j)!}{N!} \int_{-L/2}^{L/2} d\lambda_1 \dots \int_{-L/2}^{L/2} d\lambda_j R_j(\lambda_1, \dots, \lambda_j)$$

We see that the hole probability  $A(L)$  is expressed in terms of  $n$ -point correlation functions as

$$A(L) = \sum_{j=0}^N \frac{(-1)^j}{j!} \int_{-L/2}^{L/2} \dots \int_{-L/2}^{L/2} d\lambda_1 \dots d\lambda_j R_j(\lambda_1, \dots, \lambda_j)$$

Example: uncorrelated (Poissonian) sequences

$$\mathcal{P}(\lambda_1, \dots, \lambda_N) = p(\lambda_1) p(\lambda_2) \dots p(\lambda_N), \quad p(\lambda) \geq 0$$

$$\int p(\lambda) d\lambda = 1, \quad \bar{N}_B = N \int_B p(\lambda) d\lambda, \quad \Sigma_2(B) = \frac{N-1}{N} \bar{N}_B$$

$$A(L) = \left[ 1 - \int_{-L/2}^{L/2} p(\lambda) d\lambda \right]^N. \quad \text{In the limit } N \rightarrow \infty$$

typical spacing is  $\Delta = \left[ \overline{p(\lambda)} \right]^{-1} = \frac{1}{N p(\lambda)}$ . Let  $\frac{L}{\Delta} = S < \infty$

Then for  $B = (-L/2, L/2)$  we have:  $\bar{N}_B = S = \Sigma_2(B)$ , and

$$A(L) = e^{-S}.$$

# Method of orthogonal polynomials

van der Monde determinant:

$$\det \begin{pmatrix} 1 & \dots & 1 \\ \lambda_1 & \dots & \lambda_N \\ \lambda_1^2 & \dots & \lambda_N^2 \\ \vdots & \vdots & \vdots \\ \lambda_1^{N-1} & \dots & \lambda_N^{N-1} \end{pmatrix} = (-1)^{\frac{N(N-1)}{2}} \prod_{i < j} (\lambda_i - \lambda_j)$$

any set of polynomials of degree  $k$ :  $\pi_k(\lambda) = a_k \lambda^k + \dots$

we have

$$\prod_{i < j} (\lambda_i - \lambda_j) = \frac{(-1)^{\frac{N(N-1)}{2}}}{a_0 a_1 \dots a_{N-1}} \det \begin{pmatrix} \pi_0(\lambda_1) & \dots & \pi_0(\lambda_N) \\ \pi_1(\lambda_1) & \dots & \pi_1(\lambda_N) \\ \pi_2(\lambda_1) & \dots & \pi_2(\lambda_N) \\ \vdots & \vdots & \vdots \\ \pi_{N-1}(\lambda_1) & \dots & \pi_{N-1}(\lambda_N) \end{pmatrix} \propto \det (\pi_{i-1}(\lambda_j))_{i,j=1}^N$$

can rewrite the joint probability function as

$$\mathcal{P}(\lambda_1, \dots, \lambda_N) \propto \left[ \prod_{i < j} (\lambda_i - \lambda_j) \exp \left\{ -\frac{1}{2} \sum_{i=1}^N Q(\lambda_i) \right\} \right]^2$$

$$\propto \left[ \det \left( e^{-\frac{1}{2} Q(\lambda_j)} \pi_{i-1}(\lambda_j) \right)_{i,j=1}^N \right]^2$$

$$= \det \begin{pmatrix} e^{-\frac{1}{2} Q(\lambda_1)} \pi_0(\lambda_1) & \dots & e^{-\frac{1}{2} Q(\lambda_N)} \pi_0(\lambda_N) \\ \vdots & \ddots & \vdots \\ e^{-\frac{1}{2} Q(\lambda_1)} \pi_{N-1}(\lambda_1) & \dots & e^{-\frac{1}{2} Q(\lambda_N)} \pi_{N-1}(\lambda_N) \end{pmatrix} \cdot \det \begin{pmatrix} e^{-\frac{1}{2} Q(\lambda_1)} \pi_0(\lambda_1) & \dots & e^{-\frac{1}{2} Q(\lambda_1)} \pi_{N-1}(\lambda_1) \\ \vdots & \ddots & \vdots \\ e^{-\frac{1}{2} Q(\lambda_N)} \pi_0(\lambda_N) & \dots & e^{-\frac{1}{2} Q(\lambda_N)} \pi_{N-1}(\lambda_N) \end{pmatrix}$$

$$= \det \left( \sum_{j=1}^N e^{-\frac{1}{2} Q(\lambda_i)} \pi_{j-1}(\lambda_i) \cdot e^{-\frac{1}{2} Q(\lambda_k)} \pi_{j-1}(\lambda_k) \right)_{i,k=1}^N$$

$K_N(\lambda_i, \lambda_k)$  - "kernel"

Thus,

$$P(\lambda_1, \dots, \lambda_N) \propto \det(K_N(\lambda_i, \lambda_k))_{i,k=1}^N$$

$$K_N(\lambda, \lambda') = \sum_{j=0}^{N-1} \varphi_j(\lambda) \varphi_j(\lambda'), \quad \varphi_{i-1}(\lambda) = e^{-\frac{1}{2}Q(\lambda)} \frac{1}{\pi_{i-1}(\lambda)}$$

Suppose now that the polynomials  $\pi(x)$  form an orthonormalized system w.r.t the weight  $e^{-Q(x)}$

$$\int dx e^{-Q(x)} \pi_i(x) \pi_j(x) = \delta_{ij} \quad i, j \geq 1$$

Then the kernel  $K_N(x, y)$  satisfies the reproducing property

$$\int dy K_N(x, y) K_N(y, z) = \sum_{j,k=0}^{N-1} e^{-\frac{1}{2}(Q(x)+Q(z))} \frac{1}{\pi_j(x) \pi_k(z)} \underbrace{\int dy e^{-\frac{1}{2}Q(y)} \pi_j(y) \pi_k(y)}_{\delta_{jk}}$$

$$\sum_j e^{-\frac{1}{2}(Q(x)+Q(z))} \frac{1}{\pi_j(x) \pi_j(z)} K_N(x, z) \quad (A)$$

and also

$$q_N = \int dx K_N(x, x) = \sum_{j=0}^{N-1} \int e^{-Q(x)} \frac{1}{\pi_j(x) \pi_j(x)} dx \quad N \quad (B)$$

Theorem (Gaudin Mehta)

Let  $J_n(x) = (J_{ij})_{i,j=1}^n$  be  $n \times n$  matrix with entries  $J_{ij} = f(x_i, x_j)$  and the function  $f(x, y)$  satisfies for some measure  $d\mu(x)$

$$(A) \int f(x, y) f(y, z) d\mu(y) = f(x, z), \quad (B) \int f(x, x) d\mu(x) = q$$

$$\text{Then } \int \det J_n(x) d\mu(x_n) = [q^{(n-1)}] \det(J_{n-1})$$

where the matrix  $J_{n-1}$  has the same functional form with

$$x = (x_1, \dots, x_n) \text{ replaced by } x = (x_1, \dots, x_{n-1})$$