

Moments and Large Deviations of Linear Statistics of β -Ensembles

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XI Brunel-Bielefeld Workshop
12 December 2015

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Motivations

Gaussian β -Ensemble, $\beta > 0$,

$$\begin{aligned} P(x_1, \dots, x_N) &\propto \exp\left(-\beta N \sum_{j=1}^N x_j^2\right) \prod_{1 \leq j < k \leq N} |x_k - x_j|^\beta \\ &= \exp(-\beta H(x_1, \dots, x_N)), \end{aligned}$$

$$H(x_1, \dots, x_N) = -N \sum_{j=1}^N x_j^2 + \sum_{1 \leq j < k \leq N} \log |x_k - x_j|$$

The j.p.d.f. of the eigenvalues can be interpreted as the Boltzmann factor of a one-component plasma of Coulomb particles confined on a line

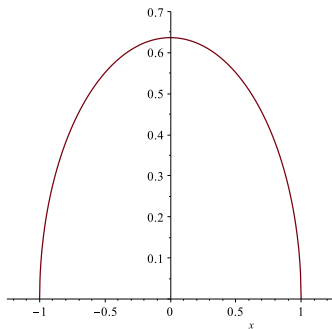
Motivations

In the limit as $N \rightarrow \infty$ the eigenvalue distribution is the unique Borel measure $d\mu$ that minimize the energy functional

$$I[\mu] = - \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} \log |t - u| d\mu(t) d\mu(u) + \int_{-\infty}^{\infty} t^2 d\mu(t)$$

Semicircle law:

$$d\mu(t) = \frac{2}{\pi} \sqrt{1 - t^2} dt$$



Motivations

For “well-behaved” test functions f

$$\frac{1}{N} \sum_{j=1}^N f(x_j) \rightarrow \frac{2}{\pi} \int_{-1}^1 f(t) \sqrt{1-t^2} dt \quad \text{a.s.}$$

Theorem (Johansson '98)

The linear statistics

$$\sum_{i=1}^N f(x_i) - \frac{2N}{\pi} \int_{-1}^1 f(t) \sqrt{1-t^2} dt$$

converges in distribution to a normal random variable with variance σ^2 which depends only on f .

- There is no $1/\sqrt{N}$ normalization!
- This behaviour of the global fluctuations of the spectra is a general phenomenon in random matrix ensembles!

Motivations

Question:

What can we say to the large N corrections to CLT's for linear statistics of Random Matrices?

Applications to:

- Quantum Transport — weak localization corrections.
- QCD and Statistical Mechanics — Free energy asymptotic expansions.
- Combinatorics.
- Quantum Chaos.
- Fluid Dynamics.
- Log-correlated Gaussian processes

Outline

- ① $\text{Tr } X^k$ for finite N in β ensembles. (FM, Alexi Reynolds, arXiv:1510.02390)
- ② Large Deviations for the 2d-OCP. (F. Cunden, A. Maltsev, FM, *Phys. Rev. E*, **91** (2015), 060105(R))

$\text{Tr } X^k$ for finite N in β -Ensemble

- Classical β -ensemble,

$$P(x_1, \dots, x_N) = K_{\mathcal{E}} \prod_{j=1}^N w(x_j) \prod_{1 \leq j < k \leq N} |x_k - x_j|^{\beta}, \quad \beta > 0.$$

- Classical weights,

$$w(x) = \begin{cases} e^{-x^2/2}, & x \in \mathbb{R}, & \text{Gaussian.} \\ x^{\gamma} e^{-x/2}, & x \in \mathbb{R}_+, & \text{Laguerre.} \\ x^{\gamma_1} (1-x)^{\gamma_2}, & x \in [0, 1], & \text{Jacobi.} \end{cases}$$

- What can we say about

$$M_k = \mathbb{E} \left\{ \text{Tr } X^k \right\}_{\mathcal{E}}, \quad k \in \mathbb{N}, ?$$

- Recall that for the Gaussian β -ensemble

$$\frac{1}{N} \sum_{j=1}^N f(x_j) \rightarrow \frac{2}{\pi} \int_{-1}^1 f(t) \sqrt{1-t^2} dt \quad \text{a.s.}$$

- This means that

$$\begin{aligned} \lim_{N \rightarrow \infty} \frac{1}{N} \mathbb{E} \left\{ \text{Tr } X^{2k} \right\}_{\text{G}\beta\text{E}} &= \frac{2}{\pi} \int_{-1}^1 t^{2k} \sqrt{1-t^2} dt \\ &= \left(\frac{1}{2} \right)^{2k} \underbrace{\frac{1}{k+1} \binom{2k}{k}}_{\text{Catalan numbers}} \end{aligned}$$

- We look at the finite N version of these moments

Previous results

- For the GUE (Brézin et al. 1978)

$$\frac{1}{N} \mathbb{E} \left\{ \text{Tr } X^{2k} \right\}_{\text{GUE}} = \sum_{g=0}^{\lfloor \frac{k}{2} \rfloor} a_g(k) N^{-2g}$$

- For the $G\beta E$ first few terms in the asymptotic expansion as $N \rightarrow \infty$ (Forrester and Witte 2014)
- Maple Algorithm (Dumitriu and Edelman 2006)
- Bourger turbulence. Fyodorov, Le Doussal, Rosso (2010)
- Applications to log-correlated Gaussian processes. Fyodorov and Le Doussal, arxiv:1511.04258

Main Result 1

FM and A. Reynolds (2015)

The finite N moments of the classical ($G\beta E$, $J\beta E$ and $L\beta E$) β -Ensemble are

$$\mathbb{E} \left\{ \text{Tr } X^k \right\}_{\mathcal{E}} = \frac{1}{\alpha^k k!} \sum_{\lambda \vdash k} j_{\lambda} \kappa_{(k)}^{\lambda}(\alpha) \mathbb{E} \left\{ C_{\lambda}^{(\alpha)} \right\}_{\mathcal{E}} \quad \alpha = 2/\beta$$

$$\lambda = (\lambda_1, \dots, \lambda_r), \quad \lambda_1 + \dots + \lambda_r = k$$

- The quantities $\mathbb{E} \left\{ C_{\lambda}^{(\alpha)} \right\}_{\mathcal{E}}$ are averages of Jack polynomials and can be expressed in terms of special values of multivariate classical OP's (Forrester and Baker 1998)
- The coefficients $\kappa_{(k)}^{\lambda}(\alpha)$ are

$$\kappa_{(k)}^{\lambda}(\alpha) = (-1)^{|\lambda| - \lambda_1} \frac{\alpha^k k (\lambda_1 - 1)!}{j_{\lambda} (|\lambda| - 1)!} \prod_{i=1}^{\ell(\lambda) - 1} \binom{\frac{i}{\alpha}}{\lambda_i + 1} \lambda_{i+1}!$$

$\text{Tr } X^k$ for finite N in β -Ensemble

For example, for the $L\beta E$ we have

$$\mathbb{E} \left\{ \text{Tr } X^k \right\}_{L\beta E} = \frac{1}{\alpha^k k!} \sum_{\lambda \vdash k} j_{\lambda} \kappa_{(k)}^{\lambda}(\alpha) L_{\lambda}^{\alpha, \gamma}(\mathbf{0})$$

$$L_{\lambda}^{\alpha, \gamma}(\mathbf{0}) = \left(\gamma + \frac{N-1}{\alpha} + 1 \right)_{\lambda}^{(\alpha)} C_{\lambda}^{(\alpha)}(1^N)$$

Note that the $\kappa_{(k)}^{\lambda}(\alpha)$ do not depend on the measure of the matrix ensembles.

The coefficients $\kappa_{(k)}^{\lambda}(\alpha)$ can be interpreted as inverse of Jack Characters

We can also compute averages of secular coefficients of characteristic polynomials

$\text{Tr } X^k$ and Jack characters

- Jack polynomials are a basis in the ring of symmetric functions.
- p_λ are the power sum symmetric functions (traces)

$$p_\lambda = p_{\lambda_1} \cdots p_{\lambda_r}, \quad p_j = x_1^j + \cdots + x_N^j.$$

- It turns out that

$$C_\lambda^{(\alpha)} = \frac{\alpha^k k!}{j_\lambda} \sum_{\mu \vdash k} \theta_\mu^\lambda(\alpha) p_\mu$$

- Jack Characters are **transition matrices** (Stanley 1989, Vassileva 2014).

$\text{Tr } X^k$ and Jack characters

- For example when $\alpha = 1$ $C_\lambda^{(1)} = s_\lambda$ (Schur polynomials)

$$s_\lambda = \frac{1}{m!} \sum_{\mu \vdash m} g_\mu \chi_\mu^\lambda p_\mu \quad \text{and} \quad p_\mu = \sum_{\lambda \vdash m} \chi_\mu^\lambda s_\lambda,$$

where χ_μ^λ are the characters of the symmetric group.

- Since $\text{Tr } X^k = p_k(x_1, \dots, x_N)$ and the $\kappa_{(k)}^\lambda(\alpha)$ are constant,

$$p_k = \frac{1}{\alpha^k k!} \sum_{\lambda \vdash k} j_\lambda \kappa_{(k)}^\lambda(\alpha) C_\lambda^{(\alpha)}$$

- The $\kappa_{(k)}^\lambda(\alpha)$ are the inverse of the Jack characters.

Large Deviations for the 2d-OCP

- The Boltzmann-Gibbs canonical measure of the 2d One-Component-Plasma (OCP) model is

$$\mathbb{P}_\beta(\{\vec{r}_k\}) = \frac{1}{\mathcal{Z}_{N,\beta}} e^{-\beta H(\vec{r}_1, \dots, \vec{r}_N)}$$

$$H(\{\vec{r}_k\}) = -\frac{q^2}{2} \sum_{i \neq j} \log\left(\frac{r_{ij}}{L}\right) + \frac{q^2 N}{2} \sum_k \left(\frac{r_k}{L}\right)^2,$$

$$\vec{r}_k = (x_k, y_k), \quad r_k = |\vec{r}_k|, \quad r_{ij} = |\vec{r}_j - \vec{r}_i|$$

- Model for vortex systems in superfluids, superconductor (Ginzburg-Landau model), Bose-Einstein condensates.
- $\beta = 2$ is the Ginibre Ensemble of complex random matrices

Mean radial displacement

$$\Delta_N = \frac{1}{N} \sum_{k=1}^N r_k.$$

Define

$$\mathcal{P}_{N,\beta}(x) = \langle \delta(x - \Delta_N) \rangle$$

F. Cunden, A. Maltsev, FM (2015)

$$\mathcal{P}_{N,\beta}(x) \asymp \begin{cases} \left(\frac{2x}{3}\right)^{\beta N^2/2} e^{-(2/3)\beta N^2(x^2-4/9)} & \text{for } x \ll 1 \\ \exp\left[-\beta N^2\left(x - \frac{2}{3}\right)^2\right] & \text{for } x \simeq \frac{2}{3} \\ \left(\frac{2x}{3}\right)^{\beta N^2/2} e^{-(1/2)\beta N^2(x^2-4/9)} & \text{for } x \gg 1. \end{cases}$$

Fourth order phase transition at $x = \frac{2}{3}$

Large Deviations for the 2d-OCP

- Consider the moment-generating function

$$\widehat{\mathcal{P}}_{N,\beta}(s) = \langle \exp(-\beta s N^2 \Delta_N) \rangle$$

- The average is taken with respect to the Boltzmann-Gibbs measure

$$\frac{1}{\mathcal{Z}_{N,\beta}} e^{-\beta H(\vec{r}_1, \dots, \vec{r}_N)}$$

- From the general theory of large deviations for log-gases

$$J(s) = - \lim_{N \rightarrow \infty} \frac{1}{\beta N^2} \log \widehat{\mathcal{P}}_{N,\beta}(s)$$

exists and is everywhere differentiable

Large Deviations for the 2d-OCF

- The probability law of Δ_N satisfies a large deviation principle with
 - Speed βN^2
 - Rate function

$$\Psi(x) = - \lim_{N \rightarrow \infty} \frac{1}{\beta N^2} \log \mathcal{P}_{N,\beta}(x) = \inf_s [J(s) - sx]$$

- We can compute $J(s)$:

$$J(s) = \frac{1}{2} \operatorname{arsinh} \left(\frac{s}{2} \right) - \frac{s^2}{4} + \frac{s}{48} \left[(s^2 + 10) \sqrt{s^2 + 4} - |s|^3 \right].$$

- Differentiating at $s = 0$

$$\kappa_1(\Delta_N) = \frac{2}{3}; \quad \kappa_2(\Delta_N) = \frac{1}{2\beta N^2}; \quad \kappa_3(\Delta_N) = \frac{1}{2\beta^2 N^4}.$$

Large Deviations for the 2d-OCP

- How do we compute $J(s)$?
- Minimize mean-field energy functional

$$I_s[\mu] = -\frac{1}{2} \iint_{\vec{r} \neq \vec{r}'} d\mu(\vec{r}) d\mu(\vec{r}') \log |\vec{r} - \vec{r}'| + \int d\mu(\vec{r}) V_s(|\vec{r}|)$$

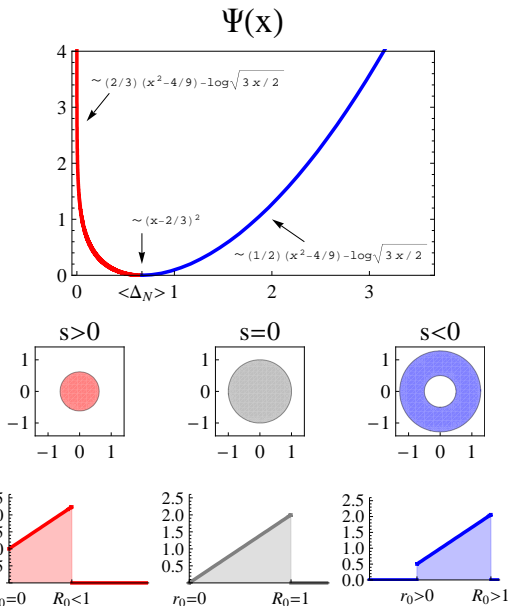
- Recall that

$$\mathcal{P}_{N,\beta}(s) = \langle \exp(-\beta s N^2 \Delta_N) \rangle,$$

where $V_s(|\vec{r}|) = \frac{1}{2} r^2 + sr$.

- $J(s)$ as excess free energy

$$\begin{aligned} J(s) &= - \lim_{N \rightarrow \infty} \frac{1}{\beta N^2} \log \left[\frac{\int e^{-\beta H(\vec{r}_1, \dots, \vec{r}_N; s)} }{\int e^{-\beta H(\vec{r}_1, \dots, \vec{r}_N; 0)} } \right] \\ &= I_s[\mu_s^*] - I_0[\mu_0^*], \end{aligned}$$



Top: rate function $\Psi(x)$. Centre: support of the equilibrium distribution. Bottom: radial distribution $\int_0^{2\pi} d\mu_s(|\vec{r}|, \phi)$.

Conclusions

- Explicit formulae for the averages of $\text{Tr } X^k$ and of the secular coefficients for finite N in β -ensembles in terms of Jack characters.
- Large deviations in 2d-OCP for β -ensembles unveil a 4th order phase transition.