Invariant $\beta$-ensembles and the Gauss-Wigner crossover

Romain Allez$^{1,2}$ and Jean-Philippe Bouchaud$^2$ and Alice Guionnet$^3$

$^1$ Université Paris-Dauphine, Ceremade, 75016 Paris, France,
$^2$ Capital Fund Management, 6–8 boulevard Haussmann, 75009 Paris, France and
$^3$ U.M.P.A. ENS de Lyon 46, allée d’Italie 69364 Lyon Cedex 07 - France.

(Dated: May 1, 2014)

We define a new diffusive matrix model converging towards the $\beta$-Dyson Brownian motion for all $\beta \in [0, 2]$ that provides an explicit construction of $\beta$-ensembles of random matrices that is invariant under the orthogonal/unitary group. For small values of $\beta$, our process allows one to interpolate smoothly between the Gaussian distribution and the Wigner semi-circle. The interpolating limit distributions form a one parameter family that can be explicitly computed. This also allows us to compute the finite-size corrections to the semi-circle.

Since Wigner’s initial intuition that the statistical properties of the eigenvalues of random matrices should provide a good description of the excited states of complex nuclei, Random Matrix Theory has become one of the prominent field of research, at the boundary between atomic physics, solid state physics, statistical mechanics, statistics, probability theory and number theory [1–2]. It is well known that the joint distribution of the eigenvalues of a large Gaussian random matrix can be expressed as the Boltzmann-Gibbs equilibrium weight of a one-dimensional repulsive Coulomb gas confined in an harmonic well. However, the effective “inverse temperature” $\beta$ of the system cannot take arbitrary values but is quantized (in units of the repulsive Coulomb potential). Depending on the symmetry of the random matrix, only three values are allowed $\beta = 1$ for symmetric real matrices, $\beta = 2$ for Hermitian matrices and $\beta = 4$ for the symplectic ensemble. This is known as Dyson’s “threefold way”. The existence of matrix ensembles that would lead to other, possibly continuous, values of $\beta$, is a very natural question, and the quest for such ensembles probably goes back to Dyson himself. Ten years ago, Dumitriu and Edelman [4] have proposed an explicit construction of tri-diagonal matrices with non-identically distributed elements whose joint law of the eigenvalues is the one of $\beta$-ensembles for general $\beta$. Another construction is proposed in [1] p. 426–427 (see also [5]) and uses a bordering procedure to construct recursively a sequence of matrices with eigenvalues distributed as $\beta$-ensembles. This construction gives not just the eigenvalue probability density of one matrix of the sequence but also the joint eigenvalue probability density of all matrices. This has lead to a renewed interest for those ensembles, that have connections with many problems, both in physics and in mathematics, see e.g. [6]. The aim of the paper is to provide another construction of $\beta$-ensembles that is, at least to our eyes, natural and transparent, and respects by construction the orthogonal/unitary symmetry [7]. Another motivation for our work comes from the recent development of free probability theory. “Freeness” for random matrices is the natural extension of independence for classical random variables. Very intuitively, two real symmetric matrices $A, B$ are mutually free in the large $N$ limit if the eigenbasis of $B$ can be thought of as a random rotation of the eigenbasis of $A$ (see e.g. [8] for an accessible introduction to freeness and for more rigorous statements). “Free convolution” then allows one to compute the eigenvalue distribution of the sum $A + B$ from the eigenvalue distribution of $A$ and $B$, much in the same way as convolution allows one to compute the distribution of the sum of two independent random variables. In this context, the Wigner semi-circle distribution appears as the limiting distribution for the sum of a large number of free random matrices, exactly as the Gaussian is the limiting distribution for the sum of a large number of iid (independent and identically distributed) random variables. A natural question, from this perspective, is whether one can build a natural framework that interpolates between these two limits.

Let us first recall Dyson’s Brownian motion construction of the GOE [9] (for the sake of simplicity, we will only consider here extensions of the $\beta = 1$ ensemble, but similar considerations hold for $\beta = 2$ Hermitian matrices see [10] for full details). One introduces a fictitious time $t$ for the evolution of an $N \times N$ real symmetric matrix $M(t)$. The evolution of the symmetric matrix is governed by the following stochastic differential equation (SDE):

$$\text{d}M(t) = -\frac{1}{2}M(t)\text{d}t + \text{d}H(t)$$

where $\text{d}H(t)$ is a symmetric Brownian increment (i.e. a symmetric matrix whose entries above the diagonal are independent Brownian increments with variance $\langle \text{d}H_{ij}^2(t) \rangle = \frac{\sigma^2}{2}(1 + \delta_{ij})\text{d}t$). Standard second order perturbation theory allows one to write the equation for the eigenvalues $\lambda_i$ of the matrix $M(t)$:

$$\text{d}\lambda_i = -\frac{1}{2}\lambda_i \text{d}t + \frac{\sigma^2}{2} \sum_{j \neq i} \frac{\text{d}t}{\lambda_i - \lambda_j} + \sigma \text{d}b_i,$$

where $b_i(t)$ are independent standard Brownian motions. This defines Dyson’s Coulomb gas model, i.e. “charged” particles on a line, with positions $\lambda_i$, interacting via a logarithmic potential, subject to some thermal noise and
confined by a harmonic potential. One can deduce from the above equation the Fokker-Planck equation for the joint density \( P(\{\lambda_i\}, t) \), for which the stationary joint probability density function (pdf) is readily found to be:

\[
P^*(\{\lambda_i\}) = Z \prod_{i<j} |\lambda_i - \lambda_j|^{\beta} \exp \left[ -\frac{1}{2\sigma^2} \sum_i \lambda_i^2 \right],
\]

with \( \beta = 1 \), independently of \( \sigma^2 \) and where \( Z \) is a normalization factor. The above expression is the well known joint distribution of the eigenvalues of an \( N \times N \) random GOE matrix. The Wigner distribution can be recovered either by a careful analysis of the mean marginal univariate distribution \( \rho(\lambda) = \int \ldots \int d\lambda_2 \ldots d\lambda_N P^*(\lambda = \lambda_1, \lambda_2, \ldots, \lambda_N) \) in the large \( N \) limit [12], or by using the above SDE [2] to derive a dynamical equation for the Stieltjes transform \( G(z, t) \) of \( \rho(\lambda, t) \):

\[
G(z, t) = \frac{1}{N} \sum_{i=1}^{N} \frac{1}{\lambda_i(t) - z}, \quad z \in \mathbb{C}.
\]

With this scaling, the spectrum is spread out in a region of width of order \( \sqrt{N} \) and therefore \( z \sim \sqrt{N} \) and \( G \sim 1/\sqrt{N} \). Applying Itô’s formula to \( G(z, t) \) and using [2], we obtain the following Burgers equation for \( G \) [13]:

\[
2 \frac{\partial G}{\partial t} = \frac{\sigma^2 N}{2} \frac{\partial^2 G}{\partial z^2} + \frac{\partial G}{\partial z} + (2 - \alpha) \frac{\sigma^2}{2} \frac{\partial^2 G}{\partial z^2}
\]

where \( \alpha \) is introduced for later convenience, with \( \alpha = 1 \) for now. Note that we have neglected a noise term of mean zero and variance of order \( 1/N\sqrt{N} \). It converges to the solution of Eq. (5) only up to fluctuations of order \( (G^2) - \langle G^2 \rangle \sim 1/N^3 \) (in agreement with e.g. [2] Theorem 9.2). The contribution of these fluctuations is thus \( 1/N \) smaller than the diffusion term in Eq. (5), and can be neglected.

For large \( N \), the last (diffusion) term of Eq. (5) is of order \( 1/N \) smaller than the others. To leading order, the stationary solution (where the derivative with respect to time is set to 0) can be easily integrated with respect to \( z \) as:

\[
\frac{1}{2} \frac{\alpha \sigma^2 N G^2(\infty)(z) + z G(\infty)(z)} = -1,
\]

where the integration constant comes from the boundary condition \( G(z) \sim -1/z \) when \( z \to \infty \). It is then easy to solve this equation to find the Stieltjes transform that indeed corresponds to the Wigner semi-circle density:

\[
G(\infty)(z) = \frac{1}{\alpha \sigma^2 N} \left[ \sqrt{z^2 - 2\alpha \sigma^2 N} - z \right],
\]

\[
\longrightarrow \rho(\lambda) = \frac{1}{\pi \alpha \sigma^2 N} \sqrt{2\alpha \sigma^2 N - \lambda^2} \mathbb{1}_{|\lambda| \leq \sqrt{2\alpha \sigma^2 N}}.
\]

Now let us turn to the central idea of the present paper. In Dyson’s construction, the extra Gaussian slice \( dM(t) \) that is added to \( H(t) \) is chosen to be independent of \( M(t) \) itself. The eigenbasis of \( dH(t) \) is a random rotation, taken uniformly over the orthogonal group. As mentioned above, this corresponds to free addition of matrices, and Eq. (5) can indeed be derived (for \( N = \infty \)) using free convolution [5]. If instead we choose to add a random matrix \( dY(t) \) that is always diagonal in the same basis as that of \( M(t) \), the process becomes trivial. The diagonal elements of \( M(t) \) are all sums of iid random variables, and the eigenvalue distribution converges towards the Gaussian. The construction we propose is to alternate randomly the addition of a “free” slice and of a “commuting” slice. More precisely, our model is defined as follows: we divide time into small intervals of length \( 1/n \) and for each interval \([k/n; (k+1)/n]\), we choose independently Bernoulli random variables \( \epsilon^k \), \( k \in \mathbb{N} \) such that \( \mathbb{P}(\epsilon^k = 1) = p = 1 - \mathbb{P}(\epsilon^k = 0) \). Then, setting \( \epsilon^t = \epsilon^n_{[nt]} \), our diffusive matrix process simply evolves as:

\[
dM_n(t) = -\frac{1}{2} M_n(t) dt + \epsilon^t dH(t) + (1 - \epsilon^t) dY(t)
\]

where \( dH(t) \) is a symmetric Brownian increment as above and where \( dY(t) \) is a symmetric matrix that is co-diagonalizable with \( M_n(t) \) (i.e. the two matrix have the same eigenvectors) but with a spectrum given by \( N \) independent Brownian increments of variance \( \sigma^2 dt \). It is clear that the eigenvalues of the matrix \( M_n(t) \) will cross at some points but only in intervals \([k/n; (k+1)/n]\) for which \( \epsilon^k = 0 \) (in the other intervals where they follow Dyson Brownian motion with parameter \( \beta = 1 \), it is well known that the repulsion is too strong and that collisions are avoided). In such a case, the eigenvalues are re-numbered at time \( t = (k + 1)/n \) in increasing order.

Now, using again standard perturbation theory, it is easy to derive the evolution of the eigenvalues of \( M_n(t) \) denoted as \( \lambda_i^n(t) \leq \cdots \leq \lambda_N^n(t) \):

\[
d\lambda_i^n = -\frac{1}{2} \lambda_i^n dt + \frac{\sigma^2}{2} \sum_{j \neq i} \frac{dt}{\lambda_i^n - \lambda_j^n} + \sigma db_i
\]

where the \( b_i \) are independent Brownian motions also independent of the \( \epsilon_i^k \), \( k \in \mathbb{N} \).

A mathematically rigorous derivation provided in [10] allows one to show that the scaling limits \( \lambda_i(t) \), when \( n \to \infty \), of the eigenvalues \( \lambda_i^n(t) \) obey the following modified Dyson SDE:

\[
d\lambda_i = -\frac{1}{2} \lambda_i dt + p \frac{\sigma^2}{2} \sum_{j \neq i} \frac{dt}{\lambda_i - \lambda_j} + \sigma db_i,
\]

with the additional ordering constraint \( \lambda_1(t) \leq \cdots \leq \lambda_N(t) \) for all \( t \). One of the difficulty of the proof comes from the fact that when \( p < 1 \), there is a positive probability for eigenvalues to collide in finite time (the ordering constraint is therefore useful at those points to re-start). The idea is then to show that collisions are in a sense
sufficiently rare for the above SDE to make sense (see [10] for further details). Using the SDE (10), one can derive as above the stationary distribution for the joint distribution of eigenvalues, which is still given by Eq. [9] but with now $\beta = \alpha = p \leq 1$. A very similar construction can be achieved in the GUE case, leading to $\beta = 2p$. As announced, our dynamical procedure, that alternates standard and free addition of random matrices, can lead to any $\beta$-ensemble with $\beta \leq 2$. The corresponding matrices $M(t)$ are furthermore invariant under the orthogonal (or unitary) group. This is intuitively clear, since both alternatives (adding a free slice or adding a commuting slice) respect this invariance, and lead to a Haar probability measure for the eigenvectors (i.e. uniform over the orthogonal/unitary group). We have also proved that a collision leads to a complete randomization of the eigenvectors within the two-dimensional subspace corresponding to the colliding eigenvalues, see again [10].

It is well known that the eigenvalue density corresponding to the measure $P^*$ given by [3] is the Wigner semi-circle for any $\beta > 0$. In fact, using [3] with now $\alpha = \beta = p$, one immediately finds that the eigenvalue density is a semi-circle with edges at $\pm \sqrt{2\beta N}$. We simulated numerically the matrix $M_n(t)$ with $N = 200$ for a very small step $1/n$ and until a large value of $t$ so as to reach the stationary distribution for the eigenvalues. Then we started recording the spectrum and the nearest neighbor spacings (NNS) every 100 steps so as to sample the ensemble. We verified that the spectral density of $M_n(t = \infty)$ is indeed in very good agreement with the Wigner semi-circle distribution for $\beta = 1/2$. Our sample histogram for the NNS distribution is displayed in Fig. 1. We also added the corresponding Wigner surmise (which is expected to provide a good approximate description of the NNSD).

**FIG. 1.** Empirical NNSD $P(s)$ for the matrix $M_n(t = \infty)$ for $\beta = p = 1/2$ with the Wigner surmise (red curve) corresponding to $\beta = 1/2$, which behaves as $s^3$ when $s \to 0$.

From the point of view of a cross-over between the standard Gaussian central limit theorem for random variables and the Wigner central limit theorem for random matrices, we see that as soon as the probability $p$ for a non-commuting slice is positive, the asymptotic density is the Wigner semi-circle, with a width of order $\sqrt{pN}$. A continuous cross-over therefore takes place for $p = 2c/N$ with $c$ strictly positive and independent of $N$. When $c = 0$, $\rho(\lambda)$ is a Gaussian of rms $\sigma$, which indeed corresponds to the solution of Eq. (5) for $\alpha = 0$. Setting for simplicity $\sigma = 1$, the SDE for the system $(\lambda_i(t))$ becomes

$$d\lambda_i = -\frac{1}{2}\lambda_i \, dt + \frac{c}{N} \sum_{j \neq i} \frac{dt}{\lambda_i - \lambda_j} + db_i,$$  

(11)

with the additional ordering constraint $\lambda_1(t) \leq \cdots \leq \lambda_N(t)$ and the stationary joint pdf is still given by [3] but with now a vanishing repulsion coefficient $\beta = 2c/N$. In order to elicit the cross-over, we study Eq. (11) with $\alpha = 2c/N$. The stationary differential equation corresponding to (5) (note this time that all terms are of the same and the second derivative term is not negligible) can be integrated with respect to $z$ again as:

$$cG^2 + zG + \frac{dG}{dz} = -1,$$  

(12)

where the integration constant comes from the boundary condition $G \sim -1/z$ for $z \to \infty$. Note that (12) can be recovered directly from the saddle point equation: under the measure $P^*$ with $\beta = 2c/N$, the energy of a configuration of the $\lambda_i$’s can be expressed in term of the continuous state density $\rho$, neglecting terms $\ll 1$, as:

$$\mathcal{E}[\rho] = \frac{1}{2} \int \lambda^2 \rho(\lambda) d\lambda - c \int \int \ln(|\lambda - \lambda'|) \rho(\lambda) \rho(\lambda') d\lambda d\lambda'.$$

The probability density $P^*$ therefore rewrites in term of $\rho$ as:

$$P^*[\rho] = Z \exp \left( -N \left[ \mathcal{E}[\rho] + \int \rho \ln(\rho) \right] \right) \delta(\int \rho - 1),$$

where the entropy term, which is negligible when $\beta = p$ is of order 1, is now of the same order as the energy term (see [14] for a detailed discussion on the origin of the entropy term). We now need to minimize the quantity $\mathcal{E}[\rho] + \int \rho \ln(\rho)$ with respect to $\rho$. It is easy to see that the unique minimizer $\rho_c$ satisfies:

$$\int \frac{\lambda \rho_c(\lambda)}{\lambda - z} d\lambda - 2c \int \int \frac{\rho_c(\lambda) \rho_c(\lambda')}{(\lambda - z)(\lambda - \lambda')} d\lambda d\lambda' + \int \rho_c(\lambda) d\lambda + \nu = 0,$$

where $\nu$ is an integration constant. It is now straightforward to derive (12) from this last equation by identifying each term and choosing the constant $\nu$ so as to have the correct boundary condition for the Stieltjès transform of a probability measure. As expected physically, the diffusion term in (12) corresponds exactly to the entropy contribution to the saddle-point.

Now, to solve (12), we first set $G(z) := u'(z)/cu(z)$ to obtain a second order equation on $u$:

$$u''(z) + z u'(z) + cu(z) = 0.$$  

(13)
It follows from the asymptotic behavior of $G(z)$ that, for $|z| \to \infty$,

$$u(z) \sim \frac{A_1}{z^c}.$$  \hfill (14)

Eq. (13) can in turn be transformed with the change of function $u(z) := e^{-z^2/4}y(z)$ into a Schrodinger equation (sometimes called \textit{Weber differential equation}) on $y(z)$:

$$y''(z) + \left[c - \frac{1}{2} - \frac{1}{4}z^2\right]y(z) = 0.$$  \hfill (15)

The solutions of (15) are known (see [21] page 1067) to write as $y(z) = A_2D_{c-1}(z) + A_3D_{-c}(iz)$ where $D_{c-1}, D_{-c}$ are parabolic cylinder functions and where $A_2$ and $A_3$ are two constants. The general solution for $u$ therefore is $u(z) = e^{-z^2/4}(A_2D_{c-1}(z) + A_3D_{-c}(iz))$ and the correct asymptotic behavior of $u$ is fulfilled for $A_2 = 0$. Now, one can recover the spectral density $\rho_c(\lambda)$ associated to $G$ by the classical inversion formula:

$$\rho_c(\lambda) = \lim_{\epsilon \to 0} \frac{1}{\pi} \Im \mathcal{I}(G(\lambda - i\epsilon))$$

$$= \frac{1}{c\pi |y(\lambda)|^2} \left(y_1'y_1 - y_2'y_2\right)(\lambda),$$  \hfill (16)

$$= \frac{1}{c\pi |y(\lambda)|^2} \left(y_1'y_1 - y_2'y_2\right)(\lambda),$$  \hfill (17)

where $y_1$ and $y_2$ are respectively the real and imaginary part of $y$. Using (15), it is straightforward to see that the derivative of $(y_1'y_1 - y_2'y_2)$ is identically zero and therefore the function is constant. This gives up a normalizing constant the exact expression for $\rho_c(\lambda)$. It happens that this constant can be computed explicitly [15]. Eq. (12) was also studied in detail by Kerov [16] and Askey & Wimp [17] (see also [18]). The final result for $\rho_c(\lambda)$ reads, for all $c > 0$:

$$\rho_c(\lambda) = \frac{1}{\sqrt{2\pi\Gamma(1+c)}} \frac{1}{|D_{-c}(i\lambda)|^2};$$  \hfill (18)

$$D_{-c}(z) = \frac{e^{-z^2/4}}{\Gamma(c)} \int_0^{\infty} dx e^{-x^2} x^{2-2c-1}. \quad \hfill (19)$$

This expression was again checked with numerical simulations with very good agreement. The integral representation for $D_{-c}(z)$ does not hold for $c = 0$, but the function $D_{-c}(iu)$ is still well defined for $c \in (-1; 0]$ (see [17] Theorem 8.2.2). It is easy to check that $\rho_0(u) = e^{-u^2/2}/\sqrt{2\pi}$ when $c = 0$, as expected. When $c \to \infty$, the Wigner semi-circle law is recovered

$$\rho_c(u) \approx \frac{1}{2\pi c^{2}} \sqrt{4c - u^2}.$$  \hfill (20)

At least when $c$ is a non-negative integer, the integral form of $D_{-c}(i\lambda)$ can be computed analytically. This enables to find the tails of $\rho_c(u)$ for some values of $c$ and large $u$. The asymptotic behavior reads:

$$\rho_c(u) \sim u^{2c} e^{-u^2/2} \quad (|u| \to \infty).$$  \hfill (21)

Let us return to (5) for $\beta = \alpha \in (0, 2)$. Interestingly, our method allows us to compute the correction to the Wigner semi-circle for the spectral density for large but finite $N$ due to the last diffusion term, which is usually neglected. Indeed one can solve as above the stationary equation of (5) keeping every term. This leads to the following \textit{corrected spectral density}, valid for large but finite $N$:

$$\rho(\lambda) = \frac{\sqrt{\alpha}}{\sqrt{2\pi}\Gamma(1+c)} \frac{1}{|D_{-c}(i\sqrt{\alpha}\lambda)|^2},$$

(22)

where $\alpha = 2/(2 - \beta)$ and $c = \beta N/(2 - \beta)$. Numerical verification for this is provided in Fig. 3.

The above discussion can also be formally extended to $-1 \leq c < 0$, corresponding to a weakly attracting Coulomb gas (also mentioned in [7]; see also [19] for an application). The system evolves, away from collisions, as in [11] but with $c \in (-1; 0]$ and with infinitesimal “up-down pushes” to separate particles when they collide. This kind of system has not been rigorously defined in the literature before but it should be possible, at least for a non trivial range of negative values of $c,

![FIG. 2. Density $\rho_c(u)$ for $c = 0, 1, 2, 3, 4$, showing the progressive deformation of the Gaussian towards Wigner’s semicircle.](image)

![FIG. 3. Numerical simulations of the state density of $M_{t}(t = \infty)$ for $p = \beta = 1/2, N = 50$. Black curve: sample density. Red curve: Finite size correction (22), which coincides almost perfectly with the sample density. Blue curve: $N \to \infty$ Wigner semi-circle density.](image)
since the attraction between particles is the same as for Bessel processes of dimension $\delta \in [0;1]$. (see [24] page 3)). We conjecture that the stationary density for large system is again given by the above Askey-Wimp-Kerov distributions $\rho_c$, for the parameter range $c \in (-1;0]$. For $c = -1$, the stationary density $\rho_{-1}$ is a Dirac mass at 0. Beyond this level, the attraction is too strong and the gas completely collapses on itself.

As a conclusion, we have provided here the first explicit construction of invariant $\beta$-ensembles of random matrices, for arbitrary $\beta \leq 2$. The stationary distribution for the eigenvectors is the Haar probability measure on the orthogonal group if $0 < \beta \leq 1$, respectively unitary group if $1 < \beta \leq 2$. We have found a natural scaling limit that allows one to interpolate smoothly between the Gaussian distribution, relevant for sums of independent random variables, and the Wigner semi-circle distribution, relevant for sums of free random matrices. The interpolating limit distributions form a a one parameter family that can be explicitly computed. Let us mention three interesting open problems. First, our alternate Bernoulli process of commuting and free matrix ‘slices’ can probably be done differently, for example by introducing an Ornstein-Uhlenbeck process on the orthogonal group that mean-reverts towards the identity matrix. By using these matrices $O$ to construct the ‘slices’ as $d\mathbf{M}(t) = O^T d\Delta O$, one may be able to generate other interesting ensembles by tuning the parameters of the Ornstein-Uhlenbeck process, again interpolating between commuting and free addition. Second, it would be interesting to know how the eigenvalue spacing distribution smoothly interpolates between the Poisson distribution and Wigner’s surmise. Finally, the statistics of the largest eigenvalue is also very interesting (and now well known for $\beta > 0$, see [22,25]): one should be able to interpolate smoothly, as a function of $c$, between the well-known Gumbel distribution of extreme value statistics and the Tracy-Widom($\beta$) distributions. Whether this can be mapped into a generalized KPZ/Directed polymer problem remains to be seen.

Acknowledgements

We are grateful to S. N. Majumdar for many important suggestions. We thank C. Garban, R. Rhodes, V. Vargas, F. Benaych-Georges, L. Dumaz, R. Chicheportiche, G. Schehr and P. P. Vivo for useful comments and discussions. A. G. and J.-P. B. thank the organizers of the NYU-Abu-Dhabi conference in January 2011, where this work was initiated (on way to buying dates).

[1] G. Akemann, J. Baik, Ph. Di Francesco, The Oxford Handbook of Random Matrix Theory, Oxford University Press (2011).
[2] Z. Bai, J. Silverstein, Spectral Analysis of Large dimensional random matrices, Second Edition, Springer (2010).
[3] G.W. Anderson, A. Guionnet and O. Zeitouni, An Introduction to Random Matrices, Cambridge Studies in Advanced Mathematics, Cambridge University Press (2009).
[4] I. Dumitriu, A. Edelman, Matrix Models for Beta Ensembles, Journal of Mathematical Physics 43, 5830–5847 (2002).
[5] P. J. Forrester, Log gases and Random Matrices, Princeton University Press (2010).
[6] P. J. Forrester, E. M. Rains, Interpretations of some parameter dependent generalizations of classical matrix ensembles, Prob. Th. Rel. Fields 131 (2005) 1-61.
[7] for an attempt in the context of $2 \times 2$ matrices, see: P. Vivo and S.N. Majumdar, Tunable correlations in a $2 \times 2$ hermitian random matrix model Physica A 387, 4839 (2008).
[8] A. Tulino, S. Verdù, Random Matrix Theory and Wireless Communications, Foundations and Trends in Communication and Information Theory, 1, 1-182 (2004).
[9] F.J. Dyson, A Brownian-Motion Model for the Eigenvalues of a Random Matrix, J. Math. Phys. 3, 1191 (1962)
[10] R. Allez and A. Guionnet, A diffusive matrix model for invariant $\beta$-ensembles, arXiv:1206.1460 (2012).
[11] E. Cépa and D. Lépingle, Diffusing particles with electrostatic repulsion, Probability Theory and Related Fields, (1997).
[12] M. L. Mehta, Random Matrices, Elsevier (Amsterdam) 2004.
[13] L. C. G. Rogers and Z. Shi, Interacting Brownian particles and the Wigner law, Probab. Theory Relat. Fields 95, 555-570 (1993)
[14] D. Dean, S. Majumdar, Extreme value statistics of eigenvalues of Gaussian random matrices, Phys. Rev. E 77, 041108 (2008)
[15] S. N. Majumdar, private communication, and in preparation.
[16] S. Kerov, Interlacing measures, in Kirillov’s seminar on representation theory, volume 181 Amer. Math. Soc. Transl. 181, 35-83. Providence, RI, 1998.
[17] R. Askey and J. Wimp. Associated Laguerre and Hermite polynomials. Proc. Roy. Soc. Edinburgh A, 9615-37, (1984)
[18] S. T. Belinschi, M. Bozejko, F. Lehner, and R. Speicher, The classical normal distribution is II-infinite divisibility, Advances in Mathematics, 226, 3677-3698 (2011).
[19] Y. V. Fyodorov, P. Le Doussal, A. Rosso, Freezing transition in decaying Burgers turbulence, Europhys. Lett. 90 (2010) 60004.
[20] W. Werner, H. Wu, On conformally Invariant CLE explorations, arxiv 1112.1211.
[21] I. S. Gradshteyn and I. M. Ryzhik, Table of integrals, Series, and products, corrected and enlarged edition, Academic Press.
[22] L. Dumaz and B. Virág, The right tail exponent of the Tracy-Widom-beta distribution, [arXiv:1102.4818] Annales de l'IHP.
[23] G. Borot, C. Nadal, Right tail expansion of Tracy-Widom beta laws, [arXiv:1111.2761]
[24] G. Borot, B. Eynard, S. N. Majumdar et C. Nadal, Large deviations of the maximal eigenvalue of random matrices, J. Stat. Mech. (2011) P11024
[25] P.J. Forrester, *Spectral density asymptotics for Gaussian and Laguerre $\beta$-ensembles in the exponentially small region*. arXiv:1111.1350.