Large deviations of the maximal eigenvalue of random matrices

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Received 7 October 2011
Accepted 8 November 2011
Published 30 November 2011

Online at stacks.iop.org/JSTAT/2011/P11024
doi:10.1088/1742-5468/2011/11/P11024

Abstract. We present detailed computations of the nondecaying terms (three dominant orders) of the free energy in a one-cut matrix model with a hard edge \( a_{\beta} \) in \( \beta \) ensembles, with any polynomial potential. \( \beta > 0 \) is not restricted to the standard values \( \beta = 1 \) (Hermitian matrices), \( \beta = 1/2 \) (symmetric matrices) and \( \beta = 2 \) (quaternionic self-dual matrices). This model allows us to study the statistics of the maximum eigenvalue of random matrices. We compute the large deviation function to the left of the expected maximum. We specialize our results to the Gaussian \( \beta \) ensembles and check them numerically. Our method is based on general results and procedures already developed in the literature to solve the Pastur equations (also called 'loop equations'). It allows us to compute the left tail of the analog of Tracy–Widom laws for any \( \beta \), including the constant term.

Keywords: matrix models
## Large deviations of the maximal eigenvalue of random matrices

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doi:10.1088/1742-5468/2011/11/P11024
1. Introduction

Extreme value statistics, in particular the statistics of the maximum of a set of random variables, play a very important role as they have a large variety of applications in various fields such as finance, hydrology (rainfall, flood, etc) and engineering. On the other hand, random matrix theory has raised a great deal of interest in both the physics and
mathematics communities. The distribution of the maximal eigenvalue of a random matrix is thus of great importance.

For usual Gaussian ensembles of random matrices (Gaussian orthogonal ensemble, Gaussian unitary ensemble and Gaussian symplectic ensemble), the distribution of the maximal eigenvalue close to its mean value is known to converge to the Tracy–Widom distribution \[24,25\] for large \(N\) (where the matrix is of size \(N \times N\)). This describes small typical fluctuations around the mean value. We are interested here in the large deviation of the distribution, which describes large and rare events (well spaced from the mean value). We actually consider as a starting point a distribution on \(\mathbb{R}^N\) that generalizes the distribution of the eigenvalues of a random matrix for a general potential \(V(x)\) (instead of a quadratic potential for Gaussian ensembles) and a general \(\beta\) (Dyson index).

We consider the following joint distribution:

\[
P(\lambda_1, \ldots, \lambda_N) = \frac{1}{Z(\infty)} |\Delta(\lambda_i)|^{2\beta} \prod_{i=1}^{N} e^{-(N\beta/4)V(\lambda_i)} \quad \text{with} \quad \Delta(\lambda_i) = \prod_{1 \leq i < j \leq N} (\lambda_i - \lambda_j) \quad (1.1)
\]

where \(Z(\infty) = \int_{\mathbb{R}^N} d\lambda_1 \cdots d\lambda_N |\Delta(\lambda_i)|^{2\beta} \prod_{i=1}^{N} e^{-(N\beta/4)V(\lambda_i)}\) is the (unconstrained) partition function.

We define

\[
Z(a) = \int_{[-\infty, a]^N} d\lambda_1 \cdots d\lambda_N |\Delta(\lambda_i)|^{2\beta} \prod_{i=1}^{N} e^{-(N\beta/4)V(\lambda_i)}, \quad (1.2)
\]

\(Z(a)\) is the partition function for the weight given by equation (1.1), multiplied by \(\prod_{i=1}^{N} \Theta(a - \lambda_i)\). The probability that the maximal eigenvalue \(\lambda_{\text{max}}\) is smaller than or equal to \(a\) (i.e. the probability that all eigenvalues \(\lambda_i\) are smaller than or equal to \(a\)) is

\[
P(a) = \int_{[-\infty, a]^N} d\lambda_1 \cdots d\lambda_N P(\lambda_1, \ldots, \lambda_N) = \frac{Z(a)}{Z(\infty)}, \quad (1.3)
\]

The probability that the largest eigenvalue lies between \(a\) and \(a + da\) is

\[
\partial_a P(a) da = \frac{dZ(a)}{Z(\infty)}. \quad (1.4)
\]

For \(\beta = 1/2, 1, 2\), the probability distribution on \(\mathbb{R}^N\) (the space of eigenvalues) given in equations (1.1) derives from an invariant probability distribution on a set of matrices, whereas no such derivation is known for other values of \(\beta\). We still call \(Z(a)\) a ‘matrix model’ and talk about ‘\(\beta\) ensembles’. This is only a convention of language to refer to a statistical weight of the form of equation (1.2). In the case of a quadratic potential \(V(x) = x^2/2\), one recovers the usual Gaussian ensembles: Gaussian orthogonal ensemble (\(\beta = 1/2\)), Gaussian unitary ensemble (\(\beta = 1\)) and Gaussian symplectic ensemble (\(\beta = 2\)). And, in general, one defines the Gaussian \(\beta\) ensemble. Note that \(\beta\) differs from the usual Dyson index \(\tilde{\beta}\) by a factor 2: \(\tilde{\beta} = 2\beta\). The normalization we choose in the definition of \(G\beta E\) is also the one used in [23]. Let us mention Dumitriu and Edelman [12] have found a tridiagonal matrix model, whose distribution of eigenvalues reproduces the \(G\beta E\)
ensemble. As far as our methods are concerned, these sparse matrix models are very different in nature from those we consider.

We want to compute the large \( N \) expansion of \( Z(a) \). For a one-cut solution, we assume\(^3\) that \( \ln Z(a) \) has an expansion in powers of \( 1/N \). It takes the form

\[
\ln Z(a) = D_{N,\beta} + \sum_{g,k \geq 0} \left( \frac{N \sqrt{\beta}}{t} \right)^{2-2g-k} \left( \sqrt{\beta} - \frac{1}{\sqrt{\beta}} \right)^k F^{g,k}(a).
\] (1.5)

In the following, we will write \( F(a) = \sum_{g,k \geq 0} (N \sqrt{\beta}/t)^{2-2g-k} (\sqrt{\beta} - (1/\sqrt{\beta}))^k F^{g,k}(a) \). The coefficients \( F^{g,k} \) are independent of \( N \) and \( \beta \). They may depend on \( a \), on \( t \) and on the potential \( V(x) \). They are given by geometric quantities associated with a plane curve, which were introduced in \([10]\). In this case, the plane curve is related to the equilibrium density of eigenvalues. \( D_{N,\beta} \) is a normalization constant, it depends only on \( N \) and \( \beta \), and not on \( a, t \) and \( V \). We tackle the question of this constant in section 3.8.

For instance, limiting ourselves to the nonnegative powers of \( N \) (we will call them ‘nondecaying terms’ or ‘stable’ terms), we have

\[
\ln Z(a) = D_{N,\beta} + \frac{N^2}{t^2} \beta F^{0,0}(a) + \frac{N}{t} (\beta - 1) F^{0,1}(a) + F^{1,0}(a) + (\beta + \beta^{-1} - 2) F^{0,2}(a) \quad (1.6)
\]

where \( F^{0,0} \) is called the prepotential, \( F^{0,1} \) is the entropy, \( F^{1,0} \) is related to a Laplacian determinant and \( F^{0,2} \) is the Polyakov anomaly. All the higher \( F^{g,k} \) are also given by geometric quantities (their interpretation is less known), and there exists an algorithm to compute them recursively.

For the statistics of the largest eigenvalue, \( F^{0,0}(a) \) is the function of large deviation, and higher \( F^{g,k} \) are its subleading corrections. This expansion of \( F(a) \) as a series in \( 1/N \) for large \( N \) is actually valid only on the left of the mean value. On the right of the mean value, \( \ln(Z(a)/Z(\infty)) \) is expected to be exponentially small, i.e. much smaller than any power of \( 1/N \). Note that, for the Gaussian case (quadratic potential \( V(x) = x^2/2 \)), the leading term \( F^{0,0}(a) \) of the left-tail large deviation had already been computed a few years ago \([13,14]\) using a Coulomb gas technique \([22]\). Here, we explain how to compute recursively all the \( F^{g,k} \) for a general potential and a general \( \beta \). We also give an explicit expression for the nondecaying terms \( F^{0,0}, F^{0,1}, F^{1,0} \) and \( F^{0,2} \) for the Gaussian case (quadratic potential) and compare our results with Monte Carlo numerical simulations.

**Outline of the paper**

We first present (section 2) the topological expansion of matrix models and the method of loop equations. We describe in detail the one-cut solution of the loop equations in section 3. We specialize it for the Gaussian \( \beta \) ensembles in section 4, with comparison to numerics in section 4.2. We discuss in general the scaling limit at the edge of the spectrum and make the link with the Tracy–Widom law in section 5. The appendices give the proof of the results used in the text to settle the ‘constant problem’.

\(^3\) The proof of this assumption under some hypothesis is a work in progress \([8]\).
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2. Correlators and loop equations

2.1. Definitions

We shall consider the expectation value of the resolvent

\[ W_1(x) = \left\langle \sum_{i=1}^{N} \frac{1}{x - \lambda_i} \right\rangle \]  

(2.1)

and, more generally, the non-connected correlation of \( n \) resolvents:

\[ W_n(x_1, \ldots, x_n) = \left\langle \prod_{j=1}^{n} \sum_{i=1}^{N} \frac{1}{x_j - \lambda_i} \right\rangle. \]  

(2.2)

Their cumulants (the connected correlation functions) are called ‘correlators’ for short:

\[ W_n(x_1, \ldots, x_n) = \left\langle \prod_{j=1}^{n} \sum_{i=1}^{N} \frac{1}{x_j - \lambda_i} \right\rangle_C. \]  

(2.3)

For instance, \( W_2(x_1, x_2) = W_2(x_1, x_2) + W_1(x_1)W_1(x_2). \)

Let us write the potential

\[ V(x) = t_0 + \sum_{j \geq 1} t_j x^j \]  

(2.4)

and introduce the loop insertion operator

\[ \frac{\partial}{\partial V(x)} = -x^{-1} \frac{\partial}{\partial t_0} - \sum_{j \geq 1} x^{-(j+1)} \frac{1}{j} \frac{\partial}{\partial t_j}. \]  

(2.5)

Applying successively this operator to \( \ln Z(a) \) gives the \( n \)-point correlation functions:

\[ W_n(x_1, \ldots, x_n) = \left( \frac{t}{N \beta} \right)^n \frac{\partial}{\partial V(x_1)} \cdots \frac{\partial}{\partial V(x_n)} \ln Z(a). \]  

(2.6)

The correlators satisfy loop equations. These are named Schwinger–Dyson equations (in the physics literature) or Pastur equations (in the maths literature). They follow from the invariance of an integral under a change of variable. Alternatively, they can be obtained by integration by parts. One must just pay attention to boundary terms at \( \lambda_i = a \). We derive them by two equivalent methods, giving complementary information.

2.2. Loop equation—first method

Consider an infinitesimal change of variables \( \lambda_i \to \lambda_i + \epsilon (\lambda_i - a)/(x - \lambda_i) \). This change of variable preserves the range of integration \( ] - \infty, a] \). We have, to first order in \( \epsilon \)

\[ \frac{\partial}{\partial V(x)} \ln Z(a). \]
\[ \prod_{i=1}^{N} d\lambda_i \rightarrow \prod_{i=1}^{N} d\lambda_i \left( 1 + \epsilon \sum_{i} \frac{x-a}{(x-\lambda_i)^2} + O(\epsilon^2) \right) \]

\[ |\Delta(\lambda_i)|^{2\beta} \rightarrow |\Delta(\lambda_i)|^{2\beta} \left[ 1 + \beta \epsilon \sum_{1 \leq j \leq N} \left( \frac{1}{x-\lambda_i} - \frac{1}{x-\lambda_j} \right) \frac{x-a}{\lambda_i - \lambda_j} + O(\epsilon^2) \right] \]

\[ \rightarrow |\Delta(\lambda_i)|^{2\beta} \left[ 1 + \beta \epsilon \sum_{1 \leq j \neq i \leq N} \frac{x-a}{(x-\lambda_i)(x-\lambda_j)} + O(\epsilon^2) \right] \quad (2.7) \]

\[ \prod_{i} e^{-N\beta/\ell V(\lambda_i)} \rightarrow \prod_{i} e^{-N\beta/\ell V(\lambda_i)} \left( 1 - \epsilon \frac{N\beta}{\ell} \sum_{i} V'(\lambda_i) \frac{(\lambda_i - a)}{x-\lambda_i} + O(\epsilon^2) \right). \quad (2.8) \]

Writing that \( \delta \epsilon \ln Z(a) = 0 \), we get to order 1 in \( \epsilon \)

\[ \left\langle \frac{N\beta}{\ell} \sum_{i=1}^{N} \frac{V'(\lambda_i) \left( \lambda_i - a \right)}{x-\lambda_i} \right\rangle = \left\langle \sum_{i=1}^{N} \frac{x-a}{(x-\lambda_i)^2} + \beta \sum_{1 \leq j \neq i \leq N} \frac{x-a}{(x-\lambda_i)(x-\lambda_j)} \right\rangle \]

\[ = (x-a) \left\langle \sum_{i=1}^{N} \frac{1-\beta}{(x-\lambda_i)^2} + \beta \sum_{i=1}^{N} \frac{1}{x-\lambda_i} \sum_{j=1}^{N} \frac{1}{(x-\lambda_j)} \right\rangle \]

\[ = (x-a) ((\beta - 1) W'_1(x) + \beta W_1(x) + \beta W_2(x, x)) \quad (2.9) \]

where we have used the notations of equation (2.3) above.

The left-hand side can also be written as

\[ \left\langle \sum_{i=1}^{N} \frac{V'(\lambda_i) \left( \lambda_i - a \right)}{x-\lambda_i} \right\rangle = (x-a) \left\langle \sum_{i=1}^{N} \frac{V'(\lambda_i)}{x-\lambda_i} \right\rangle - \left\langle \sum_{i=1}^{N} V'(\lambda_i) \right\rangle \]

\[ = (x-a) \left\langle \sum_{i=1}^{N} \frac{V'(\lambda_i) - V'(x) + V'(x)}{x-\lambda_i} \right\rangle - \left\langle \sum_{i=1}^{N} V'(\lambda_i) \right\rangle \]

\[ = (x-a)V'(x)W_1(x) - (x-a) \left\langle \sum_{i=1}^{N} \frac{V'(x) - V'(\lambda_i)}{x-\lambda_i} \right\rangle \]

\[ - \left\langle \sum_{i=1}^{N} V'(\lambda_i) \right\rangle. \quad (2.10) \]

Let us define

\[ P_1(x) = \left\langle \sum_{i=1}^{N} \frac{V'(x) - V'(\lambda_i)}{x-\lambda_i} \right\rangle = (V'(x)W_1(x))_+ \quad (2.11) \]

which is a polynomial of \( x \), of degree \( \deg V' - 1 \), and

\[ c = \left\langle \sum_{i=1}^{N} V'(\lambda_i) \right\rangle. \quad (2.12) \]
The loop equation can thus be written:

$$\beta W_1(x)^2 + \beta W_2(x, x) + (\beta - 1)W_1'(x) = \frac{N\beta}{t} \left(V'(x)W_1(x) - P_1(x) - \frac{c}{x - a}\right).$$  \hfill (2.13)

### 2.3. Loop equation—second method

We can also perform an infinitesimal change of variable changing the bound of integration:

$$\lambda_i \rightarrow \lambda_i + \epsilon \frac{1}{x - \lambda_i} + O(\epsilon^2)$$  \hfill (2.14)

We have to first order in $\epsilon$

$$\prod_{i=1}^N d\lambda_i \rightarrow \prod_{i=1}^N d\lambda_i \left(1 + \epsilon \sum_{j=1}^N \frac{1}{(x - \lambda_j)^2} + O(\epsilon^2)\right)$$

$$\Delta(\lambda)^{2\beta} \rightarrow \Delta(\lambda)^{2\beta} \left[1 + \epsilon \sum_{1 \leq i \neq j \leq N} \left(\frac{1}{x - \lambda_i} - \frac{1}{x - \lambda_j}\right) \frac{1}{\lambda_i - \lambda_j} + O(\epsilon^2)\right]$$

$$\rightarrow \Delta(\lambda)^{2\beta} \left[1 + \epsilon \sum_{1 \leq i \leq N} \frac{1}{x - \lambda_i} \frac{1}{x - \lambda_j} - \sum_{i=1}^N \frac{1}{(x - \lambda_i)^2} + O(\epsilon^2)\right]$$  \hfill (2.15)

$$\prod_{i=1}^N e^{-(N\beta/t)V(\lambda_i)} \rightarrow \prod_{i=1}^N e^{-(N\beta/t)V(\lambda_i)} \left[1 + \epsilon \sum_{j=1}^N \frac{\beta N V'(\lambda_j)}{t} + O(\epsilon^2)\right].$$  \hfill (2.16)

Under this change of variable, the integral changes by $\epsilon(\partial_a Z(a)/x - a) + O(\epsilon^2)$. Collecting the first order in $\epsilon$, we obtain

$$\beta W_2(x, x) + \beta(W_1(x))^2 + (\beta - 1)W_1'(x) - \frac{\beta N}{t}(V'(x)W_1(x) - P_1(x)) = \frac{\partial_a F(a)}{x - a}. \hfill (2.17)$$

This is again equation (2.13), with the alternative information that $(\beta N/t)c = -\partial_a F(a)$.

### 2.4. Higher-order loop equations

By application of $(t/N\beta)(\partial/\partial V(x_1)), \ldots, (t/N\beta)(\partial/\partial V(x_{n-1}))$ to equation (2.17), one can obtain loop equations for the $n$-point correlators. We have to introduce

$$P_n(x; x_1, \ldots, x_{n-1}) = \left\langle \text{Tr} \frac{V'(x) - V'(M)}{x - M} \prod_{i=1}^{n-1} \text{Tr} \frac{1}{x_i - M} \right\rangle_e.$$  \hfill (2.18)
Let us write $I = \{x_1, \ldots, x_{n-1}\}$ for the spectator variables. The result is

$$
\sum_{J \subseteq I} \beta W_{|J|+1}(x, J) W_{n-|J|}(x, I \setminus J) + \beta W_{n+1}(x, x, I) + (\beta - 1) \frac{dW_n(x, I)}{dx}
= \frac{\beta N}{t} (V'(x) W_n(x, I) - P_n(x; I)) + \sum_{x_i \in I} \frac{d}{dx_i} \left( \frac{W_{n-1}(x, I \setminus \{x_i\}) - W_{n-1}(I)}{x - x_i} \right)
+ \frac{\partial_a W_{n-1}(I)}{x - a}.
$$

(2.19)

2.5. Recovering $F$

The easiest way to find $F$ is to integrate its derivative with respect to the independent parameters $(a, t, t_j)$. These derivatives can be expressed in terms of the correlators.

- With respect to the hard edge $a$. By comparison of the loop equations obtained by the two change of variables, we have seen that

$$
\partial_a F = -\frac{\beta N}{t} \langle \text{Tr} V'(M) \rangle = \frac{\beta N}{t} \text{Res} \int dV'(x) W_1(x).
$$

(2.20)

This could also be obtained by comparing the $1/x$ term (when $x \to \infty$) in equation (2.17). Another method uses the fact that $P_1$ is a polynomial, thus regular at $x = a$. So, we have

$$
\partial_a F = \lim_{x \to a} (x - a) \left[ \beta W_2(x, x) + \beta W_1(x)^2 + (\beta - 1)W_1'(x) - \frac{N\beta}{t} V'(x)W_1(x) \right].
$$

(2.21)

We will see later another method based on a residue formula.

- With respect to the times $t$ and $t_j$. We have from the matrix integral

$$
\partial_t F = \frac{N\beta}{t^2} \langle V(M) \rangle
= -\frac{N\beta}{t^2} \text{Res} \int dV(x) W_1(x)
$$

(2.22)

$$
\partial_{t_j} F = -\frac{N\beta}{t} \left\langle \frac{M_j}{j} \right\rangle
= \frac{N\beta}{t} \text{Res} \int dx_j W_1(x).
$$

(2.23)

2.6. Topological and $\hbar$ expansion

2.6.1. Definition. We introduce two parameters:

$$
\nu = \frac{N\sqrt{\beta}}{t}, \quad \hbar = \frac{t}{N} \left( 1 - \frac{1}{\beta} \right).
$$

(2.24)
We shall consider the large $N$ expansion of the free energy $F(a)$ and of the correlators $W_n(x_1, \ldots, x_n)$ in the form
\[
F = \sum_{g,k \geq 0} \nu^{2-2g} h^k F^{g,k} \tag{2.25}
\]
\[
W_n(x_1, \ldots, x_n) = \beta^{-n/2} \sum_{g,k \geq 0} \nu^{2-2g-n} h^k W_n^{g,k}(x_1, \ldots, x_n). \tag{2.26}
\]
The prefactor in equation (2.26) was chosen such that
\[
\frac{\partial}{\partial V(x_{n+1})} W_n^{g,k}(x_1, \ldots, x_n) = W_{n+1}^{g,k}(x_1, \ldots, x_n, x_{n+1}). \tag{2.27}
\]
For the $\beta$ matrix ensembles, $h$ is of the order of $1/N$. Those definitions are such that $F^{g,k}$ and the functions $W_n^{g,k}$ are independent of $N$ and $\beta$. They depend on $a$, on $t$, and on the potential $V(x)$. For instance, the free energy is
\[
F(a) = \frac{N^2}{t^2} \beta F^{0,0} + \frac{N}{t}(\beta - 1) F^{0,1} + F^{1,0} + (\beta + \beta^{-1} - 2) F^{0,2} + \frac{t}{N}((1 - \beta^{-1}) F^{1,1} + (\beta - 3 + 3 \beta^{-1} - \beta^{-2}) F^{0,3}) + \frac{t^2}{N^2}(\beta^{-1} F^{2,0} + (1 - 2 \beta^{-1} + \beta^{-2}) F^{1,2} + (\beta - 4 + 6 \beta^{-1} - 4 \beta^{-2} + \beta^{-3}) F^{0,4}) + o(1/N^2). \tag{2.28}
\]

2.6.2. Expansion of the loop equations. All the same, we expand
\[
P_n(x, I) = \beta^{-n/2} \sum_{g,k \geq 0} \nu^{2-2g} h^k P_n^{g,k}(x, I).
\]
To the leading order, the loop equation (equation (2.17)) is
\[
(W_1^{0,0}(x))^2 = V'(x) W_1^{0,0}(x) - P_1^{0,0}(x) - \frac{\epsilon^{0,0}}{x - a}. \tag{2.29}
\]
This relation is sometimes called the master loop equation. Also, for higher $g$, $k$, we have
\[
(V'(x) - 2W_1^{0,0}(x)) W_1^{g,k}(x) = W_2^{g-1,k}(x, x) + \frac{d}{dx} W_1^{g,k-1}(x)
\]
\[
+ \sum_{0 \leq g' \leq g, 0 \leq k' \leq k} W_1^{g',k'}(x) W_1^{g-g',k-k'}(x) + P_1^{g,k}(x) + \frac{\epsilon^{g,k}}{x - a}. \tag{2.30}
\]
Also, for $n$-point correlators, the expansion of equation (2.19) is
\[
(V'(x) - 2W_1^{0,0}(x)) W_n^{g,k}(x, I) = W_{n+1}^{g-1,k}(x, x, I) + \frac{d}{dx} W_n^{g,k-1}(x, J)
\]
\[
+ \sum_{0 \leq g' \leq g, 0 \leq k' \leq k, J \subseteq I} W_n^{g',k'}(x, J) W_n^{g-g',k-k'}(x, I \setminus J)
\]
\[
+ \sum_{x_i \in I} \frac{d}{dx_i} \left( \frac{W_n^{g,k}(x, I \setminus \{x_i\}) - (x_i - a)/(x - a) W_n^{g,k}(I)}{x - x_i} \right)
\]
\[
+ P_n^{g,k}(x, I) + \frac{\epsilon_n^{g,k}(I)}{x - a}. \tag{2.31}
\]
In those expressions, $\sum'$ means that we exclude from the sum the terms where $W_{1,0}^0$ appear (we put them by hand in the LHS).

The general solution of those equations was found in [10, 11]. All $W_{n}^g$ are ‘geometric covariants’ and all $F_{n}^g$ are ‘geometric invariants’, associated with the plane curve $L$:

$$(x, y) \in L \subseteq \mathbb{C}^2 \iff y = \frac{V'(x)}{2} - W_{1,0}^0(x).$$

The function $y$ defined on $L$ is closely related to the density of eigenvalues in the thermodynamic limit ($N \to \infty$). The triplet $(x, y, L)$ is often called the spectral curve. A deep property of stable $F_{n}^g$ is their invariance under symplectic transformations, i.e. all transformations $(x, y) \to (x_1, y_1)$ such that $|dx \wedge dy| = |dx_1 \wedge dy_1|$. It is conjectured that stable $F_{n}^g$ for $k \neq 0$ are also invariant in this sense, but there is no proof at the moment. So, we often call $F_{n}^g$ the ‘symplectic invariants of the curve’, but we use the name ‘geometric quantities associated with the curve’ to refer to general $F_{n}^g$.

In practice, the first step is to compute the unstable correlators, that is the correlators $W_{n}^{g,k}$ such that $2 - 2g - k - n \geq 0$. The most important one is the resolvent to leading order, $W_{1,0}^0$. The other unstable correlators are $W_{2}^{0,0}$ and $W_{1}^{0,1}$. Then, all the remaining correlators follow. In a similar way, there are general formulae for unstable free energies ($F_{0}^0$, $F_{0}^1$, $F_{1}^0$, $F_{1}^1$) and a uniform algorithm to compute all the other $F_{n}^g$.

Notice that unstable terms come with a nonnegative power of $N$ in the topological expansion, whereas stable terms are coefficients in the $o(1)$. In a sense, this $o(1)$ is easier to compute than the nondecaying terms.

Let us also mention that all $W_{n}^{g,k}$ (and $W_{0}^{g,k} \equiv F_{n}^g$) have a combinatorial interpretation, as counting (possibly non-orientable) ribbon graphs of genus $g$, with $k$ Möbius strips and $n$ marked vertices [11].

### 2.6.3. First few loop equations.

- For $n = 1$:

\[
2y(x)W_{1}^{0,1}(x) = \frac{d}{dx} W_{1}^{0,0}(x) + P_{1}^{0,1}(x) + \frac{c_{1}^{0,1}}{x - a}
\]

(2.33)

\[
2y(x)W_{1}^{1,0}(x) = W_{2}^{0,0}(x, x) + P_{1}^{1,0}(x) + \frac{c_{1}^{1,0}}{x - a}
\]

(2.34)

\[
2y(x)W_{1}^{0,2}(x) = \frac{d}{dx} W_{1}^{0,1}(x) + (W_{1}^{0,1}(x))^2 + P_{1}^{0,2}(x) + \frac{c_{1}^{0,2}}{x - a}
\]

(2.35)

and so on.

- For $n = 2$:

\[
2y(x)W_{2}^{0,1}(x, x') = \frac{d}{dx} W_{2}^{0,0}(x, x') + W_{1}^{0,1}(x) \left(2W_{2}^{0,0}(x, x') + \frac{1}{(x - x')^2}\right) - \frac{d}{dx'} \left(\frac{x' - a}{x - a} \frac{W_{1}^{0,1}(x')}{x - x'}\right) + P_{2}^{0,1}(x; x') + \frac{c_{2}^{0,1}(x')}{x - a}
\]

(2.36)

We may identify the free energies $F_{n}^g$ to ‘zero-point correlators, $W_{0}^{g,k}$.

\[\text{doi:10.1088/1742-5468/2011/11/P11024}\]
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\[ 2g(x)W_2^{1,0}(x, x') = W_3^{0,0}(x, x, x') + W_1^{0,1}(x) \left( 2W_2^{0,0}(x, x') + \frac{1}{(x - x')^2} \right) \]

\[ - \frac{d}{dx'} \left( x' - a \frac{W_1^{1,0}(x')}{x - a} \right) + P_2^{1,0}(x; x') + \frac{c_2^{1,0}(x')}{x - a} \]

\[ 2g(x)W_2^{0,2}(x, x') = \frac{d}{dx} W_2^{0,1}(x, x') + 2W_1^{0,1}(x)W_2^{0,1}(x, x') \]

\[ + 2W_1^{0,2}(x) \left( W_2^{0,0}(x, x') + \frac{1}{(x - x')^2} \right) \]

\[ - \frac{d}{dx'} \left( x' - a \frac{W_1^{0,2}(x')}{x - a} \right) + P_2^{0,2}(x; x') + \frac{c_2^{0,2}(x')}{x - a} \]

and so on.

- For \( n = 3 \):

\[ 2g(x)W_3^{0,0}(x, x_1, x_2) = 2W_2^{0,0}(x, x_1)W_2^{0,0}(x, x_2) + \frac{W_2^{0,0}(x, x_2)}{(x - x_1)^2} + \frac{W_2^{0,0}(x, x_1)}{(x - x_2)^2} \]

\[ + \left( \frac{d}{dx_1} \frac{(x_1 - a)/(x - a)}{x - x_1} + \frac{d}{dx_2} \frac{(x_2 - a)/(x - a)}{x - x_2} \right) W_2^{0,0}(x_1, x_2) \]

\[ + P_3^{0,0}(x; x_1, x_2) + \frac{c_3^{0,0}(x_1, x_2)}{x - a} \]

and so on.

### 2.7. Density of eigenvalues

The expected density of eigenvalues is defined as

\[ \rho(x) = \left\langle \sum_{i=1}^{N} \delta(x - \lambda_i) \right\rangle. \tag{2.36} \]

The linear form \( f \to \int f \rho \), defined on an appropriate space of test functions, has a large \( N \) expansion (in the weak topology) of the type\(^5\)

\[ \rho(x) \overset{\text{weak}}{=} \frac{1}{\sqrt{\beta}} \sum_{g,k \geq 0} \left( \frac{N \sqrt{\beta}}{t} \right)^{1-2g-k} \left( \sqrt{\beta} - \frac{1}{\sqrt{\beta}} \right)^k \rho^{g,k}(x). \tag{2.37} \]

The one-point correlator, \( W_1(x) \), is the Stieltjes transform of the density. This means that, if \( \mathcal{D} \) contains the eigenvalue support,

\[ W_1(x) = \int_{\mathcal{D}} \frac{dx'}{x - x'}. \tag{2.38} \]

Conversely, one can see the density as the discontinuity of \( W_1(x) \):

\[ 2i \pi \rho(x) = W_1(x - i0^+) - W_1(x + i0^+). \tag{2.39} \]

\(^5\) In general, this is not true when \( \rho(x) \) is considered as a function. \( \rho(x) \) has a pointwise \( \rho^0 \), but exhibits fast oscillations beyond leading order, even in the one-cut case. We thank a referee for pointing out an erroneous statement in an earlier version of the text.
In the thermodynamic limit \((N \to \infty)\)
\[
\frac{1}{N} g(x) \sim g_0^0(x) = \frac{-1}{2i\pi} \left( W_1^{0,0}(x + i0^+) - W_1^{0,0}(x - i0^+) \right).
\] (2.40)

The origin of \(\mathcal{D}\) will become more precise when we present the analytical structure of \(W_1^{0,0}\) derived from the master loop equation.

When \(x \to \infty\), we know that \(W_1^{0,0} \sim t/x\) and this implies the normalization:
\[
\int_{\mathcal{D}} dx g_0^0(x) = t.
\] (2.41)

3. One-cut solution to the loop equations

The spectral curve is defined by the master loop equation, which we rewrite:
\[
W_1^{0,0}(x) = \frac{V'(x)}{2} - \sqrt{\frac{V'(x)^2}{4} - P_1^{0,0}(x) - \frac{c_0^0}{x - a}}.
\] (3.1)

If eigenvalues accumulate in the large \(N\) limit on one segment, of the form \([b, a]\), we know that \(W_1^{0,0}(x)\) must have only one cut \([b, a]\) in the complex plane, and thus \(W_1^{0,0}(x)\) must be of the form
\[
W_1^{0,0}(x) = \frac{V'(x)}{2} - \frac{M(x)}{2} \sqrt{\frac{x - b}{x - a}}
\] (3.2)

where \(M(x)\) is a polynomial to be determined.

In principle, knowing that \(W_1^{0,0}(x)\) has one cut and matching equation (3.2) with the large \(x\) behavior \(W_1^{0,0}(x) \sim t/x\) allow us to compute explicitly the endpoint \(b\), as well as the polynomials \(M(x)\) and \(P_1^{0,0}(x)\). The easiest way to do this computation is to use the Zhukovsky map.

The purpose of this paper is to illustrate our methods, so we assume that:

- the support of eigenvalues is of the form \([b, a]\), in particular it is connected;
- \([b, a] \subseteq \mathbb{R}\).

Then, one can prove that \(W_n^{g,k}(x_1, \ldots, x_n)\) has one cut on \([b, a]\) in each variable for \(x_i \in [b, a]\), and that it is holomorphic in each variable outside this cut.

We present detailed computations of the unstable correlators \(W_{1}^{0,0}, W_{2}^{0,0}, W_{1}^{0,1}\) and the unstable free energies \((F_{0,0}, F_{0,1}, F_{0,2}, F_{1,0})\), and the algorithm to compute the other \(W_n^{g,k}\) and \(F_n^{g,k}\). General formulae for \(W_1^{0,0}\) and the procedure to compute all \(W_n^{k,g}\) when the support of eigenvalues is a finite set of paths in the complex plane can be found in the literature [9, 10].

3.1. Spectral curve, \(W_1^{0,0}\) and density

The Zhukovsky map is defined as
\[
x(z) = \alpha + \gamma(z + 1/z)
\] (3.3)
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with $\alpha = (a + b)/2$ and $\gamma = (a - b)/4$. We also have

$$\gamma(z - 1/z) = \sqrt{(x - a)(x - b)}, \quad \sqrt{\frac{x - b}{x - a}} = \pm \frac{1 + z}{1 - z}. \quad (3.4)$$

It is then clear that $W_{1,0}^{0,0}(x(z))$ and $\rho(x(z))$ are rational functions of $z$ (for this reason, $z$ is called a uniformizing variable).

Consider any potential $V$ of finite degree $d$:

$$V(x) = t_0 + \sum_{j=1}^{d} \frac{t_j}{j} x^j. \quad (3.5)$$

We have

$$V'(x) \sim t_d x^{d-1}. \quad (3.6)$$

We assume that $V$ is such that the matrix integral converges. Let us decompose

$$V'(x(z)) = u_0 + \sum_{k=1}^{d-1} u_k (z^k + z^{-k}). \quad (3.7)$$

With the Zhukovsky variable, $w_1(z) = W_{1,0}^{0,0}(x(z))$ is the solution to the problem:

(i) $w_1$ is a meromorphic function defined on $\mathbb{C}$.

(ii) $w_1$ is meromorphic outside the unit disc and

$$w_1(z) \sim t/x(z) \quad z \to \infty$$

(iii) $w_1$ is regular at $z = -1$ (i.e. $x = b$) and has at most a simple pole at $z = 1$ (i.e. $x = a$).

(iv) $w_1$ satisfies the following equation (deduced from the loop equation for a one-cut solution):

$$\forall z \in \mathbb{C}, \quad w_1(x(z)) + w_1(x(1/z)) = V'(x(z)).$$

The solution is unique and it is easy to find the answer in the form

$$w_1(z) = \frac{r}{z - 1} + \sum_{k=1}^{d-1} u_k z^{-k} \quad (3.8)$$

The behavior when $z \to \infty$ determines $b$ through

$$u_0 = -r, \quad \gamma(r + u_1) = t. \quad (3.9)$$

Then, the density in the thermodynamic limit $\rho(x) = \rho^{0,0}(x)$ is

$$\rho(x(z)) = \frac{1}{2\pi i} (w_1(1/z) - w_1(z))$$

$$= \frac{1}{2\pi i} \left( r \frac{1 + z}{1 - z} + \sum_{k=1}^{d-1} u_k (z^k - z^{-k}) \right). \quad (3.10)$$
3.1.1. More on the density. It is sometimes convenient to use the variable $\theta$ such that $z = e^{i\theta}$, that is $x = \alpha + 2\gamma \cos \theta$. We have
\[
\rho(x(e^{i\theta})) = \frac{1}{2\pi} \left( r \cotan(\theta/2) + 2 \sum_{k=1}^{d-1} u_k \sin(k\theta) \right). \tag{3.11}
\]
Equation (3.7) is, in fact, a decomposition of $V'(x)$ on the basis of Chebyshev polynomials of the first kind $T_n = \cos(n\theta)$:
\[
V'(x) = u_0 + \sum_{k \geq 1} 2u_k T_k(\cos \theta). \tag{3.12}
\]
Likewise, it is useful to decompose $V$ on the basis $(T_n)_{n \in \mathbb{N}}$:
\[
V(x) = v_0 + \sum_{k \geq 1} 2v_k T_k(\cos \theta). \tag{3.13}
\]
Then, $V'$ is naturally decomposed on the basis of Chebyshev polynomials of the second kind $U_n = \sin(n+1)\theta/\sin \theta$:
\[
V'(x) = \frac{1}{\gamma} \sum_{k \geq 1} kv_k U_k(\cos \theta). \tag{3.14}
\]
We recall that $T_n$ are orthogonal polynomials for some scalar product $(\cdot | \cdot)_T$:
\[
(T_m | T_n)_T = \frac{1}{\pi} \int_0^{\pi} d\theta T_m(\cos \theta)T_n(\cos \theta) = \frac{1}{2}\delta_{n,m} \quad \text{if } n \neq 0 \tag{3.15}
= \frac{1}{2}\delta_{0,m} \quad \text{if } n = 0. \tag{3.16}
\]
Similarly, $U_n$ are orthogonal polynomials for a related scalar product $(\cdot | \cdot)_U$:
\[
(U_m | U_n)_U = \frac{1}{\pi} \int_0^{\pi} d\theta \sin^2(\theta) U_n(\theta)U_m(\theta) = \frac{1}{2}\delta_{n,m} \quad \text{if } n \neq 0 \tag{3.17}
= \frac{1}{2}\delta_{0,m} \quad \text{if } n = 0. \tag{3.18}
\]
The expansion of equation (3.14) is related to equation (3.7) through
\[
\forall k \geq 2 \quad \gamma(u_{k-1} - u_{k+1}) = \left( T_{k-1} - T_{k+1} \left| \sum_{n \geq 1} nv_n U_{n-1} \right. \right)_T = \left( 2\sin^2 \theta U_{k-1} \left| \sum_{n \geq 1} nv_n U_{n-1} \right. \right)_T = 2 \left( U_{k-1} \left| \sum_{n \geq 1} nv_n U_{n-1} \right. \right)_U = kv_k.
\]
One can check that this formula holds for $k = 1$ as well:
\[
\forall k \geq 1, \quad \gamma(u_{k-1} - u_{k+1}) = kv_k. \tag{3.19}
\]
So, we have a compact expression for the density $dx \rho(x)$:

$$
\begin{align*}
\text{d}x(z) \rho(x(z)) &= \frac{1}{2\pi} \frac{dz}{z} \gamma \left( z - \frac{1}{z} \right) \left( \frac{1+z}{1-z} + \sum_{k \geq 1} u_k (z^k + z^{-k}) \right) \\
&= \frac{1}{2\pi} \frac{dz}{z} \left( -2\gamma (r + u_1) + \sum_{k \geq 1} \gamma (u_{k-1} - u_{k+1}) (z^k + z^{-k}) \right) \\
&= \frac{d\theta}{\pi} \left( -t + \sum_{k \geq 1} k \nu_k T_k (\cos \theta) \right). 
\end{align*}
$$

(3.20)

The eigenvalue support $\mathcal{D} = [b, a]$ is mapped to the upper unit circle from $-1$ to $1$ (we call it $-\mathcal{C}_+$) and to $[\pi, 0]$ in the variable $\theta$. The $-1$ sign in equation (3.20) is consistent with this reverse orientation in the variable $\theta$, since the total weight of $\rho$ is $t$.

### 3.1.2. The moment polynomial.

The density can also be written

$$
\rho(x) = -\frac{M(x)}{2\pi} \sqrt{\frac{x-b}{a-x}}. 
$$

(3.21)

From equation (3.1), one knows that $\rho(x)$ behaves as $\sqrt{x-b}$ when $x \to b$. So, $M$ must be a polynomial, and it is called the *moment*. Since the support of eigenvalues stays connected, $b \in ]-\infty, a[ \subseteq \mathbb{R}$ is determined as the unique zero of the polynomial in brackets such that $\rho(x)$ remains nonnegative for $x \in ]b, a[$. Besides, the degree of $M$ is $d - 1$, so it admits $d - 1$ zeros $x_1, \ldots, x_{d-1}$. For generic $V$ (i.e. a non-critical potential), those zeros are distinct from $a$ and $b$. They must lie outside $]a, b[$, and are either real, or come in pairs of complex conjugate values. Let us call $s_j$ the point such that $x(s_j) = x_j$ and $|s_j| > 1$ (we also have $x(1/s_j) = x_j$). We have

$$
M(x) = t_d \prod_{j=1}^{d-1} (x - x_j). 
$$

(3.22)

Also, we may write in a factorized form

$$
\rho(x(e^{i\theta})) = \frac{t_d \gamma^{d-1}}{2\pi} \cotan(\theta/2) \prod_{j=1}^{d-1} \frac{(e^{i\theta} - s_j)(e^{i\theta} - 1/s_j)}{e^{i\theta}}. 
$$

(3.23)

For $p \in \mathbb{N}$, we may also introduce the following quantities:

$$
m_{a, b}^{(p)} = \sum_{k \geq 1} k^p u_k, \quad m_b^{(p)} = \sum_{k \geq 1} (-1)^k k^p u_k. 
$$

(3.24)

They allow us to relate the values of the $p$th derivative of $M(x)$ at points $a$ and $b$ to the coefficients $u_k$. By expanding equation (3.25) near $z \to 1$ ($x = a$), or $z \to -1$ ($x = b$):

$$
\begin{align*}
M(a) &= r \\
M'(a) &= m_{a}^{(1)}/\gamma \\
&\quad \vdots \\
M(b) &= 4m_{b}^{(1)} - r \\
M'(b) &= -(m_{b}^{(1)} + 2m_{b}^{(3)})/3\gamma \quad \vdots
\end{align*}
$$

doi:10.1088/1742-5468/2011/11/P11024
3.1.3. The spectral curve. The function $y$ defining the spectral curve is given by
\[
y(z) = i\rho(z) = \frac{1}{2}(w_1(1/z) - w(z)) \\
= \frac{1}{2} \left( \frac{z + 1}{z - 1} M(x(z)) \right) \\
= \frac{1}{2} \left( \frac{r \frac{z + 1}{z - 1} + \sum_{k \geq 1} u_k(z^k - z^{-k})}{z - 1} \right).
\]
(3.25)

3.2. The prepotential, $F^{0,0}$

3.2.1. General formula for $F^{0,0}$. The prepotential may be found by integrating equation (3.29), which might be useful for simple examples or numerics, but do give a hint for a general formula. For this purpose, one has from homogeneity (invariance of $F$ with respect to $(t, t_j) \mapsto (\lambda t, \lambda t_j)$ for $\lambda > 0$):
\[
F^{0,0} = \frac{1}{2} \text{Res}_{x \to \infty} \frac{dx}{V(x)} W^{0,0}_1(x) + \frac{t}{2} \partial_t F^{0,0}.
\]
(3.26)

We have introduced the chemical potential $\mu = \partial_t F^{0,0}$. We state [18] that it can be computed as
\[
\mu = 2t \ln \gamma + \text{Res}_{x \to \infty} \frac{dz}{z} V(x).
\]
(3.27)

With our notations, we find $\mu = 2t \ln \gamma - v_0$.

To compute the first term in $F^{0,0}$, we can move the contour from $\infty$ to $C$ surrounding the cut. We have
\[
\text{Res}_{x \to \infty} \frac{dx}{V(x)} W^{0,0}_1(x) = \text{Res}_{z \to 0} \frac{dz}{z} \left( v_0 + \sum_{k \geq 1} v_k(z^k + z^{-k}) \right) \left( r(1 + 1/z) + \sum_{k \geq 1} kv_kz^{-k} \right)
\]
\[= \frac{1}{2} tv_0 + \frac{1}{2} \sum_{k \geq 1} kv_k^2.
\]

Eventually, we collect the terms
\[
F^{0,0} = \frac{1}{2} \sum_{k \geq 1} kv_k^2 - tv_0 + t^2 \ln \gamma.
\]
(3.28)

3.2.2. Computation of $\partial_a F^{0,0}$. After equation (2.21), $\partial_a F^{0,0}$ appears as the coefficient of the simple pole $1/(x - a)$ in
\[
[(W_1^{0,0}(x))^2 - V'(x)W_1^{0,0}(x) - P_1^{0,0}(x)].
\]
Since $P_1^{0,0}$ is regular at $x = a$, and $W_1^{0,0}$ behaves as $(x - a)^{-1/2}$, we have when $x \to a$ (or $z \to 1$)
\[
(W_1^{0,0}(x))^2 \sim \frac{\partial_a F^{0,0}}{x - a} \\
\sim \frac{\partial_a F^{0,0}}{\gamma} \frac{1}{(z - 1)^2}.
\]

doi:10.1088/1742-5468/2011/11/P11024
With equation (3.8), we obtain

\[ \frac{\partial}{\partial a} F^{0,0} = \gamma r^2 \]

3.3. Finite correction to the one-point correlator, \( W^{0,1}_1 \)

The loop equation (equation (2.33)) is

\[ M(x) \sqrt{\frac{x-b}{x-a}} W^{0,1}_1(x) = \frac{d}{dx} W^{0,0}_1(x) + P^{0,1}_1(x) + \frac{e^{0,1}}{x-a}. \]  

(3.29)

Hence

\[ W^{0,1}_1(x) = \frac{1}{4(x-a)} - \frac{1}{4(x-b)} - \frac{M'(x)}{2M(x)} + \frac{Q(x)}{M(x) \sqrt{(x-a)(x-b)}} \]  

(3.30)

where \( Q(x) \) is a polynomial. In the Zhukovsky variable \( z \), this expression is clearly a rational function of \( z \). We define

\[ \omega^{0,1}_1(z) \equiv W^{0,1}_1(x(z))' x'(z). \]  

(3.31)

We see that \( \omega^{0,1}_1(z) \) has simple poles at \( z = 1 \) (with residue \( 1/2 \)), at \( z = -1 \) (with residue \( -1/2 \)) and at \( z = 1/s_j, z = s_j \) and \( z = 0 \). Since \( W^{0,1}_1(x) \) is holomorphic outside \([b, a[\), \( \omega^{0,1}_1(z) \) must be meromorphic outside the unit disc. Therefore, the polynomial \( Q(x) \) is determined such that the poles at \( s_j \) cancel. This implies that the poles at \( 1/s_j \) have residue \(-1\). Besides, \( W^{0,1}_1(x) \in O(1/x^2) \) when \( x \to \infty \) imply that \( \omega^{0,1}_1(z) \in O(1/z^2) \) when \( z \to \infty \). This determines the residue of the pole at \( z = 0 \). Eventually we get

\[ \omega^{0,1}_1(z) = \frac{1}{2(z-1)} - \frac{1}{2(z+1)} - \sum_{j=1}^{d-1} \left( \frac{1}{z-1/s_j} - \frac{1}{z} \right) \]  

(3.32)

3.4. The entropy, \( F^{0,1} \)

3.4.1. General formula for \( F^{0,1} \). The first correction \( F^{0,1} \) was computed by Dyson himself [17]. In our notation, it is the opposite of the entropy:

\[ F^{0,1} = \int_b^a dx \rho(x) \ln(\rho(x)/t). \]

Let us insert the expression for the density (equation (3.20)), and its factorized form (equation (3.23)):

\[ F^{0,1} = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} k v_k T_k(\cos \theta) \right) \ln \rho(e^{i\theta}) \]

\[ = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} k v_k T_k(\cos \theta) \right) \ln(\rho(x)/t) \]

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\[
\times \left| \frac{t_d \gamma^{d-1}}{2\pi t} \cotan(\theta/2) \prod_{j=1}^{d-1} \frac{(e^{i\theta} - s_j)(e^{i\theta} - 1/s_j)}{e^{i\theta} - e^{i\theta}} \right|
\]

\[
= t \ln \left( \frac{t_d \gamma^{d-1}}{2\pi t} \right) - \mathcal{I} + \sum_{j=1}^{d-1} [\mathcal{J}(s_j) + \mathcal{J}(1/s_j)]
\]

(3.33)

where

\[
\mathcal{I} = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right) \ln |\tan(\theta/2)|
\]

\[
\mathcal{J}(s) = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right) \ln |e^{i\theta} - s|.
\]

To compute \(\mathcal{I}\), we use the classical integral \(\int_0^\pi \ln |\tan(\theta/2)| = 0\) and do an integration by parts:

\[
\mathcal{I} = \int_0^\pi \frac{d\theta}{\pi} \sum_{k \geq 1} v_k \frac{\sin(k\theta)}{\sin \theta}
\]

\[
= \sum_{k \text{ odd}} v_k.
\]

We compute \(\mathcal{J}(s)\) for \(|s| > 1\):

\[
\mathcal{J}(s) = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right) \left( \ln |s| + \frac{1}{2} \ln(1 - e^{i\theta}/s) + \frac{1}{2} \ln(1 - e^{-i\theta}/\bar{s}) \right)
\]

\[
= t \ln |s| + \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right)
\]

\[
\times \left( -\sum_{m=1}^{\infty} \text{Re}(s^{-m}) \frac{T_m(\cos \theta)}{m} + \text{Im}(s^{-m}) \frac{\sin(m\theta)}{m} \right)
\]

\[
= t \ln |s| + \frac{1}{2} \sum_{k \geq 1} v_k \text{Re}(s^{-k}) - \sum_{k \geq 1} v_k \text{Im} \left( \sum_{m=1}^{\infty} A_{m,k} \frac{1}{s^m} \right).
\]

\(A_{m,k}\) is a real coefficient which vanish when \(m\) and \(k\) are of the same parity, and else is equal to

\[
A_{m,k} = \frac{1}{\pi} \left( \frac{1}{m + k} - \frac{1}{m - k} \right).
\]

We compute \(\mathcal{J}(1/s)\) for \(|s| > 1\) in the same way:

\[
\mathcal{J}(1/s) = \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right) \left( \frac{1}{2} \ln(1 - e^{-i\theta}/s) + \frac{1}{2} \ln(1 - e^{i\theta}/\bar{s}) \right)
\]

\[
= \int_0^\pi \frac{d\theta}{\pi} \left( t - \sum_{k \geq 1} kv_k T_k(\cos \theta) \right)
\]

doi:10.1088/1742-5468/2011/11/P11024
\[
\sum_{m=1}^{\infty} -\text{Re}(s^{-m}) \frac{T_m(\cos \theta)}{m} - \text{Im}(s^{-m}) \frac{\sin(m\theta)}{m}
\]

\[
= \frac{1}{2} \sum_{k \geq 1} v_k \text{Re}(s^{-k}) - \sum_{k \geq 1} v_k \text{Im} \left( \sum_{m=1}^{\infty} A_{m,k} \frac{1}{s^m} \right).
\]

In the sum \( J(s) + J(1/s) \), the term containing an imaginary part cancels. Moreover, we know that, if \( s_j \) is a zero of \( \rho \), then \( \overline{s_j} \) is also a zero of \( \rho \). So, \( \sum_j s_j^{-k} \) is already real and the absolute value in the log only remove the sign of \( s_j \). Collecting the terms

\[
F^{0,1} = t \ln \left( \frac{td}{2\pi t} \prod_{j=1}^{d-1} s_j \right) - \sum_{k \text{ odd}} v_k + \sum_{k \geq 1} v_k \left( \sum_{j=1}^{d-1} \frac{1}{s_j^k} \right)
\]

(3.34)

3.4.2. Computation of \( \partial_a F^{0,1} \). One may also compute \( \partial_a F^{0,1} \), which requires less algebra. We have when \( x \rightarrow a \) (or \( z \rightarrow 1 \)) by expansion of equation (2.21):

\[
[ (2W_1^{0,0}(x) - V'(x)) W_1^{0,1}(x) + \partial_x W_1^{0,0}(x) ] \sim \frac{\partial_a F^{0,1}}{x-a}
\]

\[
\sim \frac{\partial_a F^{0,1}}{\gamma} \frac{1}{(z-1)^2}.
\]

We may extract \( \partial_a F^{0,1} \) by computing a residue. Since \( x'(z) \sim 2\gamma(z-1) \) when \( z \rightarrow 1 \), we write

\[
\partial_a F^{0,1} = \frac{1}{2} \text{Res} \left( \frac{v}{z-1} \right) \left[ \left( 2w_1^{0,0}(z) - V'(x(z)) \right) w_1^{0,1}(z) + \partial_z w_1^{0,0}(z) \right] w_1^{0,1}(z) + \partial_z w_1^{0,0}(z) \right]
\]

\[
= - \text{Res} \left( \frac{v}{z-1} \right) \left( y(z) \omega_1^{0,1}(z) \right).
\]

(3.35)

With \( \omega_1^{0,1} \) given by equation (3.32), and \( y(z) = i\pi \rho(z) \) given by equation (3.25), we obtain

\[
\partial_a F^{0,1} = \text{Res} \left( \frac{v}{z-1} \right) \left( \frac{1}{2} \frac{1}{2} \frac{1}{2} - \sum_{j=1}^{d-1} \frac{1}{s_j-1} + O(z-1) \right)
\]

(3.36)

3.5. Leading order of the two-point correlator, \( W_2^{0,0} \)

By application of the loop insertion operator to equation (3.9), we find that \( w_2(z_1, z_2) = W_2^{0,0}(x(z_1), x(z_2)) \) is the unique solution to the following problem:

(i) \( w_2(z_1, z_2) \) is a meromorphic function in \( z_1 \) and \( z_2 \), which is symmetric in its two variables.

doi:10.1088/1742-5468/2011/11/P11024
depends on the endpoints.

(iii) For $z_2 \neq \pm 1$, $w_2(z_1, z_2)$ has at most simple poles when $z_1 \to \pm 1$ (i.e. when $x_1 = a$ or $b$).

(iv) $w_2$ satisfies the following equation (deduced from the loop equation for a one-cut solution):

$$
\forall z_1 \in \mathbb{C}, \forall z_2 \in \mathbb{C} \setminus \{ \pm 1 \}, \quad w_2(z_1, z_2) + w_2(1/z_1, z_2) = -\frac{1}{(x(z_1) - x(z_2))^2}.
$$

This problem is the same with or without hard edges and its solution is universal (it only depends on the endpoints $a$ and $b$, through the Zhukovsky map). It is given by

$$
w_2(z_1, z_2) = \frac{1}{x'(z_1)x'(z_2)} \frac{1}{(z_1 - z_2)^2} - \frac{1}{(x(z_1) - x(z_2))^2}
$$

or equivalently

$$
w_2(z_1, z_2) = \frac{1}{x'(z_1)x'(z_2) (z_1 z_2 - 1)^2}
$$

3.6. Unstable $W^{k,g}_n$

3.6.1. Analytical structure. Initially, $W^{g,k}_n(x(z_1), \ldots, x(z_n))$ was defined and meromorphic for $|z_i| > 1$ and $z_i \neq \pm 1$. We shall see that it can be analytically continued (this is true under the one-cut assumption) as a meromorphic function on the whole complex plane, i.e. as a rational function of its variables. We call this function $w^{g,k}_n(z_1, \ldots, z_n)$. Let us introduce

$$
\omega^{g,k}_n(z_1, \ldots, z_n) = x'(z_1) \cdots x'(z_n) w^{g,k}_n(z_1, \ldots, z_k).
$$

We shall now investigate the analytical properties of $\omega^{g,k}_n$. We already know the important property:

(ii) $\omega^{g,k}_n(z_1, \ldots, z_n)$ has no singularity for $|z_i| \geq 1$ and $z_i \neq \pm 1$. Let us rewrite the loop equation by separating ‘singular’ and ‘regular’ terms. We recall the notation

$$
w^{0,0}_1(x(z)) = \frac{V'(x(z))}{2} - y(z)
$$

and we introduce

$$
\mathcal{E}^{g,k}_n(z, I) = \left[ \omega^{-1,k}_{n+1}(z, z, I) + \sum_{J \subseteq I, 0 \leq g' \leq g, 0 \leq k' \leq k} \omega^{g',k'}_{n+1}(z, J) \omega^{g-g',k-k'}_{n-J}(z, I \setminus J) \right]
$$

and

$$
\partial_z \left( \omega^{g,k-1}_n(z, I) \right) - \frac{x''(z)}{x'(z)} \omega^{g,k-1}_n(z, I) + \sum_{z_i \in I} (x'(z))^2 \partial_z \left( \frac{\omega^{g,k-1}_n(z, I \setminus \{z_i\})/x'(z) - ((x(z_i) - a)/(x(z) - a)) (\omega^{g,k-1}_n(I)/x'(z_i))}{x(z) - x(z_i)} \right)
$$

$$
(3.40)
$$

\text{doi:10.1088/1742-5468/2011/11/P11024}
The loop equations may be rewritten:

\[ \omega^{g,k}(z, I) = S^{g,k}(z, I) + R^{g,k}(z, I). \]  

(3.43)

\( S^{g,k} \) has the same polar structure as \( \omega^{g,k} \), apart from the fact that it may have a pole in the physical sheet, at \( z = s_j \). To find \( \omega^{g,k} \), we should add \( R^{g,k} \) to \( S^{g,k} \), thus canceling this pole. The remainder \( R^{g,k}(z, I) \) has the following properties:

- \( R^{g,k}(z, I) \) is regular when \( z \to \pm 1 \).
- \( R^{g,k}(z, I) \) may have poles only at \( z = s_j, z = 1/s_j \) and they are at most simple.

By a recursion of \( 2g - 2 + n + k \) using the loop equations, It is then possible to complete the description of the analytical structure of \( \omega^{g,k} \):

(i) \( \omega^{g,k}(z_1, \ldots, z_n) \) is a rational function of all its variables (thus defined on \( \mathbb{C}^n \)).

(ii) \( \omega^{g,k}(z_1, \ldots, z_n) \) may have poles at \( z_l = \{ \pm 1 \}, z_l = 1/s_j (1 \leq j \leq d - 1), \, z_l = 1/z_i (i \neq l) \) and \( z_l = 0 \).

(iii) \( \omega^{g,k}(z_1, \ldots, z_n) \in O(1/z^2) \) when \( z_l \to \infty \).

It is clear that the loop equations, the analytical properties (i)–(iv) and the data of \( \omega_0, \omega_1, \omega_2 \) determines uniquely the family of rational functions \( \omega^{g,k} \).

(iv) We know an extra property of the \( \omega^{g,k} \) we are looking for: they are symmetric functions of all their variables. Although not obvious, the fact that this unique solution is symmetric can be proven from the loop equations.

(v) It is also possible to prove by recursion that the pole at \( z = 0 \) is at most simple (we assume \( d \geq 2 \)). It comes from the leading term of \( P^{g,k}_n(x, I) \) (in \( R^{g,k}_n(z, I) \)) when \( z \to 0 (x \to \infty) \). All the other terms in \( S^{g,k}_n(z, I) \) and \( R^{g,k}_n(z, I) \) are regular when \( z \to 0 \).

3.6.2. Algorithm. Let us recall that \( x(z) = x(1/z) \) and \( y(z) = -y(1/z) \). Since \( P_n^{g,k}(x(z), I) \) is a polynomial in \( x(z) \), we know that

\[ R_n^{g,k}(1/z, I) = z^2 R_n^{g,k}(z, I). \]  

(3.44)

Since this rational function in \( z \) has only simple poles at \( z = s_j \) and \( 1/s_j \), it must be of the form

\[ R_n^{g,k}(z, I) = \sum_{j=1}^{d-1} R_{n,j}^{g,k}(I) \left( \frac{1}{z - 1/s_j} - \frac{1}{z - s_j} \right). \]  

(3.45)

To cancel the pole at \( z = s_j \) in \( S^{g,k}_n \), we must have

\[ R_{n,j}^{g,k}(I) = \text{Res}_{z \to s_j} S^{g,k}_n(z, I). \]  

(3.46)
Since $S_{n}^{g,k}(z,J)$ has only a simple pole at $z = s_j$, we have

$$\mathcal{R}_{n,j}^{g,k}(I) = \frac{\mathcal{E}_{n}^{g,k}(s_j,I)}{2y'(s_j)x'(s_j)}.$$  
(3.47)

This is enough to compute $\omega_{n}^{g,k}$ with equation (3.43) by recursion on $2g + k + n$. We can also find a residue formula, based on the following principle. The divergent part of a rational function $f$ at a pole $\lambda$ is given by

$$[f(z_0)]_{\lambda} = \Res_{z=\lambda} \frac{dz}{z - z_0} f(z).$$  
(3.48)

$\omega_{n}^{g,k}(z_0,I)$ is the sum of its divergent part near $z_0 = \pm 1$, $z = 1/z_i$, $z = 1/s_j$ and $0$. After $(vi)$, $\omega_{n}^{g,k}(z_0,I)$ has only a simple pole at $z_0 = 0$ and it must be such that $\omega_{n}^{g,k}(z_0,I) \in O(1/z_0^2)$ when $z_0 \to 0$. Thus

$$\omega_{n}^{g,k}(z_0,I) = \Res_{z_0=0} dz \left( \frac{1}{z_0 - z} - \frac{1}{z_0} \right) S_{n}^{g,k}(z,I)$$

$$+ \sum_{j=1}^{d-1} \Res_{z_{s_j} = z_0} dz \left( \frac{1}{z_0 - 1/s_j} - \frac{1}{z_0} \right) S_{n}^{g,k}(z,I).$$  
(3.49)

To compute the residues at $z = \pm 1$, the term $-\partial_z(\omega_{n}^{g-1}(I)\cdots)$ do not contribute and we can replace $S_{n}^{g,k}$ by

$$\hat{\mathcal{E}}_{n}^{g,k}(z,I) = \left( \omega_{n+1}^{g-1}(z,z,I) + \sum_{J \subseteq I, 0 \leq g' \leq g, 0 \leq k' \leq k} \omega_{|J|+1}^{g',k'}(z,J) \omega_{n-|J|}^{g-g',k-k'}(z,I \setminus J) \right)$$

$$+ \partial_z(\omega_{n}^{g,k-1}(z,I)) - \frac{x'(z)}{x(z)} \omega_{n+1}^{g,k-1}(z,I)$$

$$+ \sum_{z_i \in I} \frac{x'(z)x'(z_i)}{(x(z) - x(z_i))^2} \omega_{n-1}^{g,k}(z,I \setminus \{z_i\})$$  
(3.50)

$$\hat{S}_{n}^{g,k}(z,I) = \frac{\mathcal{E}_{n}^{g,k}(z,I)}{2y(z)x'(z)}.$$  
(3.51)

3.6.3. Improvement for $\omega_{n}^{g,0}$. This algorithm can be further simplified in the case $k = 0$, i.e. the limit of a Hermitian matrix model. Notice first that $\omega_{n}^{g,0}$ is obtained by the former algorithm without computing any $\omega_{n}^{g,k}(z,I)$ with $k \neq 0$. Then, one can prove the following ‘mirror relation’ by a recursion using the loop equations:

$$\omega_{n}^{g,0}(1/z,I) = z^2 \omega_{n}^{g,0}(z,I)$$  
(3.52)

There is no similar mirror relation for $\omega_{n}^{g,k}(z,I)$ with $k \neq 0$ because of the term $\partial_z \omega_{n}^{g,k-1}(z,I)$ in the loop equations. As a consequence, $\omega_{n}^{g,0}$ can only have a pole at $z = \pm 1$. Indeed, no pole at $z = z_i$ (resp. $z = s_j$) implies no pole at $z = 1/z_i$ (resp. $z = 1/s_j$) by the mirror relation. All the same, $\omega_{n}^{g,0} \in O(1/z^2)$ when $z \to \infty$ imply that

doi:10.1088/1742-5468/2011/11/P11024
\( \omega_n^0 \) is regular at \( z = 0 \). Besides

\[
\text{Res}_{\xi \to \pm 1} \omega_n^0(\xi, I) = - \text{Res}_{\xi \to \pm 1} d\xi \omega_n^0(\xi, I) = 0. \tag{3.53}
\]

Therefore, we can compute \( \omega_n^0 \) in a simple way if the \( \omega_n^{g'} \)'s with \( 2g' + n' < 2g + n \) are known: \( \omega_n^0 \) is the divergent part in the Laurent expansion of \( S_n^{g,k}(z, I) \) at \( z \to \pm 1 \).

We can again replace \( S_n^0(z, I) \) by \( S_n^0(z, I) \) and write

\[
\omega_n^0(z, I) = \text{Res}_{z \to \mp 1} \frac{dz}{z_0 - z} \hat{S}_n^0(z, I). \tag{3.54}
\]

### 3.6.4. Standard form of the topological recursion.

Formula (3.54) is valid for one-cut solution of loop equations: we used the fact that \( \omega_n^0 \) is a rational function of its variables. It is false for other cases. However, we are going to transform slightly this formula into the form of the ‘residue formula of the topological recursion’ [19], which still holds in more general settings.

With the mirror relation, one can prove that \( S_n^0(1/z, I) = z^2 \) (or use directly equation (3.43) and equation (3.44)). Hence

\[
\omega_n^0(z, I) = \frac{1}{2} \text{Res}_{z \to \mp 1} dz \left( \frac{1}{z_0 - z} - \frac{1}{z_0 - 1/z} \right) \hat{S}_n^0(z, I)
\]

\[
= \text{Res}_{z \to \mp 1} \frac{dz}{2g(z)x'(z)} \left( \frac{1}{z_0 - z} - \frac{1}{z_0 - 1/z} \right)
\]

\[
\times \left[ \omega_{n+1}^{g-1,0}(z, z, I) + \sum_{J \subseteq I, 0 \leq h \leq g} \omega_{J|J+1}^{h,0}(z, I) \omega_{n-[J]}^{g-h,0}(z, I \setminus J) \right.
\]

\[
+ \sum_{z_i \in I} \frac{x'(z) x'(z_i)}{(x(z) - x(z_i))^2} \omega_{n-1}^{0,0}(z, I \setminus \{z_i\}) \right]. \tag{3.55}
\]

Let us define

\[
\omega^0_2(z_1, z_2) = \omega^0_2(z_1, z_2) + \frac{1}{2} \frac{dx(z_1) dx(z_2)}{(x(z_1) - x(z_2))^2} \tag{3.56}
\]

and if \((n, g) \neq (2, 0)\):

\[
\omega^g_2(z_1, z_2) = \omega^g_2(z_1, z_2) + \frac{dx(z_1) dx(z_2)}{(x(z_1) - x(z_2))^2} \tag{3.57}
\]

and if \((n, g) \neq (2, 0)\):

\[
\omega_n^g(z_1, \ldots, z_n) = \omega_n^g(z_1, \ldots, z_n) = \omega_n^g(z_1, \ldots, z_n). \tag{3.58}
\]

We also define the point \( \bar{z} \neq z \) such that \( x(\bar{z}) = x(z) \). In general, it is defined at least locally around the simple zeros of \( x' \) (the branch points at which we take the residues). Here, we have \( \bar{z} = 1/z \), which happens to be defined globally.
Using the mirror property, we can rewrite

$$
\omega_n^{0,0}(z_0, I) = \text{Res}_{z \to 1, -1} dz K(z_0, z) \left[ \omega_n^{g-1,0}(z, \bar{z}, I) + \sum_{J \leq I, 0 \leq h \leq g} \omega_{[J]}^{g,0}(z, J) \omega_{n-[J]}^{g-h,0}(z, I \setminus J) \right].
$$

(3.59)

This is the standard form of the topological recursion associated with a spectral curve \((\mathcal{L}, x, y)\), also called the ‘residue formula’. The residues are computed at the simple zeros of \(dx\) on \(\mathcal{L}\). \(\omega^{0,0}_1(z_1, z_2)\) is the heat kernel of \(\mathcal{L}\), also called the Bergman kernel. \(K(z, z_0)\) is the recursion kernel. Here

$$
K(z_0, z) = \frac{1}{4y(z)x'(1/z)} \left( \frac{1}{z_0 - z} - \frac{1}{z_0 - 1/z} \right)
= \frac{1}{2(y(z) - y(\bar{z}))x'(\bar{z})} \left( \frac{1}{z_0 - z} - \frac{1}{z_0 - \bar{z}} \right)
= \frac{\int_{z}^{\bar{z}} dz' \omega^{0,0}_2(z', z_0)^n}{2(y(z) - y(\bar{z}))x'(\bar{z})}.
$$

(3.60)

The last line is the expression of the recursion kernel in the general case.

The residue formula (equation (3.59)) can be seen as a definition of some quantities \(\omega_n^{g,0}\) (for \(2 - 2g - n < 0\)) from the data of the spectral curve \((\mathcal{L}, x, y)\). These quantities satisfy some loop equations. When this formalism is applied to a spectral curve coming from a matrix model, it gives the topological expansion of the correlators of the matrix model. As a matter of fact, the residue formula allows us to express all the solutions of loop equations (non-necessarily with one-cut). The quantities \(\omega_n^{g,0}\) defined by equation (3.59) enjoy interesting properties and can be computed with a diagrammatic technique (see the review [19]).

Let us mention that another residue formula and a diagrammatic technique also exists in general for all \(\omega_n^{g,k}\) (the case of hyperelliptic spectral curves was treated in [10]).

3.6.5. Example. We apply the algorithm described in section 3.6.2 to find the \(1/N\) corrections to the one-point correlator.

\begin{itemize}
\item \((n, g, k) = (1, 1, 0)\). \(E_1^{1,0}(z) = \omega_2^{0,0}(z, z)\) has a symmetry when \(z \mapsto 1/z\). So, killing the pole at \(z = s_j\) also kills the pole at \(z = 1/s_j\). Thus, \(\omega_1^{1,0}(z)\) is the singular part at \(z = \pm 1\) of \(S_1^{1,0}(z)\). We find

$$
\omega_1^{1,0}(z) = \frac{1}{16\gamma M(a)} \frac{1}{(z - 1)^2}
+ \frac{1}{4\gamma M(b)} \frac{1}{(z + 1)^4} - \frac{1}{4\gamma M(b)} \frac{1}{(z + 1)^3}
+ \frac{-M(b) + 4\gamma M'(b)}{16\gamma M(b)^2} \frac{1}{(z + 1)^2}.
$$
\end{itemize}
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- $(n,g,k) = (1,0,2)$. This is the first case for which there is no short cut:

$$\omega^0_1(z) = -\frac{3}{16}\gamma M(a) (z-1)^2 + \frac{5}{4}\gamma M(b) (z+1)^4 + \frac{1}{4}\gamma M(b) \left(-13 + \sum_j \frac{1}{s_j+1}\right)$$

$$\times \left[ \frac{1}{(z+1)^3} + \frac{1}{\gamma M(b)} \left(\frac{3}{16} + \sum_{j \neq i} \frac{1}{(s_j+1)(s_i+1)}\right) \right]$$

$$+ \frac{5M'(b)}{4M(b)^2} \left[ \frac{1}{(z+1)^2} + \frac{1}{\gamma M(b)} \left(\sum_j \frac{2}{s_j+1}\right) \right]$$

$$\times \left[ \left(\sum_j \frac{1}{(s_j+1)^2} - \frac{1}{s_j+1}\right) - \frac{\gamma M'(b)}{M(b)} \right] \left(\frac{1}{z+1} - \frac{1}{z}\right)$$

$$+ \sum_j \frac{-2}{2} \left[ - \frac{M''(x_j)}{\gamma (M'(x_j))^3} \frac{1}{(s_j+1)^3} + \frac{2}{\gamma M'(x_j)} \frac{s_j}{(s_j+1)^3} \right] \frac{1}{(z-1/s_j)^2}$$

$$+ \left[ \frac{1}{6} \frac{1}{\gamma M'(x_j)^3} \left(2M''(x_j)M'(x_j) - 3(M'(x_j))^2\right) \frac{s_j}{s_j+1} \right.$$

$$+ \frac{M''(x_j)}{\gamma (M'(x_j))^3} \frac{s_j}{(s_j+1)^3} \left(1 - \sum_{l \neq j} \frac{s_j}{s_l-s_j}\right)$$

$$\left. + \frac{1}{\gamma M'(x_j)} \frac{s_j^2}{(s_j+1)^3} \sum_{l \neq j} \frac{2s_j}{(s_l-s_j)^2} + \frac{s_j}{s_l-s_j} \frac{-s_j^3 + s_j^2 + 2s_j - 3}{(s_j+1)} \right]$$

$$\times \left[ \frac{2}{(s_ls_j-1)^2} + \frac{2(-2s_j+3)}{s_ls_j-1} \right]$$

$$\left. + \sum_{l \neq j, l' \neq j, l \neq l'} \frac{1}{(s_ls_j-1)(s_l's_j-1)} \right] \left(\frac{1}{z-1/s_j} - \frac{1}{z}\right).$$

3.6.6. The Bergman-tau function, $F_{1,0}$. Finding $F_{1,0}$ requires the integration of $\omega^{1,0}_1(x)$ with respect to the variations of the potential, and we have

$$\omega^{1,0}_1(z_0) = \text{Res}_{z \to -1, -1} \frac{dz}{z} \frac{\omega^{0,0}_2(z, z)}{2y(z)x(z)}. \quad (3.61)$$

doi:10.1088/1742-5468/2011/11/P11024
A closed formula (when there are no hard edges) was first given by Ambjørn et al [3] and it was generalized to hyperelliptic curves by Akemann [2]. As we have seen, $\omega_{10}^2(z, z)$ is closely related to the Bergman kernel of the spectral curve: the general method of integration of $\omega_{10}$ was studied in [20] and the solution was expressed in terms of the ‘Bergman tau-function’. It depends on the positions of the edges (hard or not) and is defined as the result of integration of a closed 1-form. When there exist some hard edges, the result is slightly modified and was studied by Chekhov [9]. We apply it in the one-cut case:

$$F^{1,0} = -\frac{1}{24} \ln \left[ M(a)^3 M(b) \left( \frac{a-b}{t} \right)^4 \right]$$

(3.62)

3.6.7. The Polyakov anomaly, $F^{0,2}$. This term was first considered by Wiegmann and Zabrodin (equation 5.20 of [27]). We adapt the expression given in [10] when there are some hard edges, in the one-cut case. By comparing with $\partial_a F^{0,2}$ computed for the Gaussian potential, we found that the result of [10] must be multiplied by 1/4 to be correct. Namely, the correct formula in our case is

$$F^{0,2} = F^{0,2}_{(t)} + \frac{1}{12} \ln \left[ \frac{M(a)^3}{M(b)} \left( \frac{a-b}{t} \right)^2 \right]$$

(3.63)

where

$$F^{0,2}_{(t)} = \frac{1}{16\pi^2} \int_{C_{\text{ext}}} \frac{d\rho(z_1)}{\rho(z_1)} \int_{C_+} dz_2 \left( \frac{1}{z_2 - z_1} - \frac{1}{z_2 - 1/z_1} \right) \ln(\rho(z_2)/t)$$

$C_{\text{ext}}$ is a contour outside the unit disc, such that ind$_{C_{\text{ext}}}(z) = 1$ whenever $|z| \leq 1$ and ind$_{C_{\text{ext}}}(s_j) = -1$ for all $1 \leq j \leq d - 1$.

Let us compute, for $|z_1| > 1$:

$$A(z_1) = \frac{1}{\pi} \int_{C_+} dz_2 \left( \frac{1}{z_2 - z_1} - \frac{1}{z_2 - 1/z_1} \right) \ln(\rho(z_2)/t)$$

$$= \int_0^\pi \frac{d\theta_2}{\pi} \left( -1 - \sum_{k=1}^{d-1} \frac{2}{k z_1^k} T_k(\cos \theta_2) \right) \ln(\rho(e^{i\theta_2})/t).$$

We have already computed this kind of integral in equation (3.33). We just have to substitute

$$t \longrightarrow -1, \quad k v_k \longrightarrow 2/z_1^k$$

in the result equation (3.34). Therefore

$$A(z_1) = -\ln \left( \frac{\pi d}{2\pi t} \prod_{j=1}^{d-1} \gamma|s_j| \right) - \sum_{k=1}^{d-1} \frac{2}{k z_1^k} + \sum_{k=1}^{d-1} \frac{2}{k z_1^k} \left( \sum_{l=1}^{d-1} \frac{1}{s_l^k} \right)$$

$^6$ Beware: the free energy is defined as $-\ln Z$ in Chekhov, while we use the opposite convention.
Then

$$F^{0,2}_{(l)} = -\frac{1}{81\pi} \int_{c_{\text{ext}}} \frac{d\rho(z)}{\rho(z)} A(z)$$

$$= \frac{1}{81\pi} \int_{c_{\text{ext}}} dz \left[ \frac{1}{z+1} - \frac{1}{z-1} + \sum_{j=1}^{d-1} \left( \frac{1}{z-s_j} + \frac{1}{z-1/s_j} - \frac{1}{z} \right) \right] A(z).$$

The contour integral can be computed for each term of $A(z)$. Individually, each term giving a meromorphic function to integrate. So, for each term, we can move the contour surrounding the poles outside the unit disc, that is $z = \infty$ and $z = s_j$ $(1 \leq s_j \leq d - 1)$. We obtain

$$F^{0,2}_{(l)} = -\frac{1}{4} \left[ -(d-1) \ln \left( \frac{t_d}{2\pi t} \prod_{j=1}^{d-1} \gamma |s_j| \right) + \sum_{j=1}^{d-1} \ln \left( \frac{t_d}{2\pi t} \prod_{l=1}^{d-1} \gamma |s_l| \right) \right]$$

$$+ \sum_{j=1}^{d-1} \sum_{k \text{ odd}} \frac{2}{k} \frac{1}{s_j^k} - \sum_{j=1}^{d-1} \frac{2}{k} \frac{1}{s_j^k} \left( \sum_{l=1}^{d-1} \frac{1}{s_l^k} \right)$$

$$= \frac{1}{4} \sum_{j=1}^{d-1} \left[ \ln \left( \frac{s_j + 1}{s_j - 1} \right) + 2 \sum_{l=1}^{d-1} \ln \left( 1 - \frac{1}{s_j s_l} \right) \right]$$

$$= \frac{1}{4} \sum_{j=1}^{d-1} \ln \left( \frac{s_j^2}{(s_j - 1)(s_j + 1)^3} \right) - \sum_{1 \leq j < l \leq d-1} \ln \left( 1 - \frac{1}{s_j s_l} \right).$$

We also write the second term of $F^{0,2}$ in terms of $s_j$'s:

$$\frac{1}{12} \ln \left( \frac{16\gamma^2 M(a)^3}{M(b)} \right) = \frac{1}{12} \ln \left( \frac{16t_d^2 \gamma^{2d}}{t^2} \right) + \frac{1}{12} \sum_{j=1}^{d-1} \ln \left( \frac{1}{s_j^2} \frac{(s_j - 1)^6}{(s_j + 1)^2} \right).$$

Therefore

$$F^{0,2} = \frac{1}{12} \left[ \ln \left( \frac{16t_d^2 \gamma^{2d}}{t^2} \right) + \sum_{j=1}^{d-1} \ln \left( \frac{s_j^{10} (s_j - 1)^{11}}{(s_j + 1)^{11}} \right) \right] - \sum_{1 \leq j < l \leq d-1} \ln \left( 1 - \frac{1}{s_j s_l} \right)$$

$$\tag{3.64}$$

### 3.7. Unstable $F^{g,k}$

#### 3.7.1. Computation of $\partial_a F^{g,k}$. We can repeat the argument which led to the residue representation for $\partial_a F^{0,1}$. It shows that the derivative term $\partial_a w^{g,k}_1(z)$ has a vanishing contribution to the residue and we have

$$\partial_a F^{g,k} = \frac{1}{2} \text{Res}_{z=1} \left( -2y(z) \omega^{g,k}_1(z) + \sum_{0 \leq s' \leq g, 0 \leq k' \leq k} \frac{\omega^{g,k'}_1(z) \omega^{g-g'-k'-k}_1(z)}{x'(z)} + \omega^{g-1,k}_2(z, z) \right).$$

Another method is to extract directly $c^{g,k}_0 = -\partial_a F^{g,k}$ from $R^{g,k}$, which is a by-product of the algorithm. Indeed, after equation (3.42) and recalling $y(z) \sim -r/(z-1)$

$$\text{doi:10.1088/1742-5468/2011/11/P11024}$$

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when \( z \to 1 \):

\[ R^{g,k}_1(z) = \frac{\partial_a F^{g,k}}{r} + O(z - 1). \tag{3.66} \]

Then, after equations (3.45) and (3.46):

\[ \partial_a F^{g,k} = -r \sum_{j=1}^{d-1} \frac{s_j + 1}{s_j - 1} \text{Res } d\xi \hat{S}_1^{g,k}(\xi) \tag{3.67} \]

With this second method, we obtain

\[ \partial_a F^{1,0} = \frac{r}{\gamma^2} \sum_{j=1}^{d-1} \frac{1}{M'(x_j)} \frac{s_j^4}{(s_j - 1)^4(s_j + 1)^4} \]

\[ \partial_a F^{0,2} = \sum_{j=1}^{d-1} \frac{r}{\gamma^2 M'(x_j)} \frac{s_j^2}{(s_j - 1)^2(s_j + 1)^2} \]

\[ \times \left( -2 \sum_{l \neq j} \frac{1}{(s_l s_j - 1)^2} - \frac{2}{s_l s_j - 1} \frac{s_j^2 - s_j + 1}{(s_j - 1)(s_j + 1)} \right. \]

\[ \left. - \sum_{l \neq j, l' \neq j / l \neq l'} \frac{1}{(s_l s_j - 1)(s_l s_j - 1)} + \frac{2s_j^3 - 3s_j^2 + 4s_j - 2}{(s_j - 1)^2(s_j + 1)^2} \right). \]

3.7.2 Algorithm for \( F^{g,k} \). We state the so-called integration formula [10] for the stable free energies:

\[ F^{g,k} = \frac{1}{2 - 2g - k} \text{Res } dz \phi(z) \omega_1^{g,k}(z) \tag{3.68} \]

where \( \phi \) is a primitive of \( y(z)x'(z) \):

\[ \phi(z) = \int_1^z ydx. \tag{3.69} \]

For \( k = 0 \) and \( g \geq 2 \), this formula simplifies into

\[ F^{g,0} = \frac{1}{2 - 2g} \text{Res } dz \phi(z) \omega_1^{g,k}(z). \tag{3.70} \]

This formula is not valid (or simply does not make sense) for unstable free energies. The correct formulae were equations (3.26) and (3.27) for \( F^{0,0} \), equation (3.33) for \( F^{0,1} \), equation (3.62) for \( F^{1,0} \) and equation (3.64) for \( F^{0,2} \).
3.8. Normalization of the geometric quantities

3.8.1. Position of the problem. Let us define the partition function of a $\beta$-matrix model by

$$Z_{N,\beta}(a, V) = \int_{-\infty}^{a} \ldots \int_{-\infty}^{a} d\lambda_1 \cdots d\lambda_N |\Delta(\lambda)|^{2\beta} \exp \left( -\frac{N\beta}{t} \sum_{i=1}^{N} V(\lambda_i) \right). \quad (3.71)$$

It is known that, when the leading order ($N \to \infty$) of $W_1(x)$ has only one cut in the complex plane, there exists an asymptotic expansion ($N \to \infty$) of the form

$$\ln Z_{N,\beta}(a, \beta, V) = D_{\beta} \ln N + \sum_{g,k \geq 0} \left( \frac{N\sqrt{\beta}}{t} \right)^{2-2g-k} \left( \sqrt{\beta} - \frac{1}{\sqrt{\beta}} \right)^k \left( f^{g,k}(a, V) + d^{g,k} \right) \quad (3.72)$$

$D_{\beta}$ depends only on $\beta$ and $d^{g,k}$ is a constant. They do not depend on $a$, $V$ and $N$.

To this matrix model, we associate a plane curve $S_{a,V}$, which does not depend on $N$ and $\beta$:

$$S_{a,V} : y = W_{1}^{0,0}(x) - \frac{V'(x)}{2}. \quad (3.73)$$

Out of the curve were constructed some geometric quantities $\mathcal{F}^{g,k}(S_{a,V})$. We can form a formal asymptotic series in $N$:

$$\mathcal{F}_{N,\beta}(S_{a,V}) = \sum_{g,k \geq 0} \left( \frac{N\sqrt{\beta}}{t} \right)^{2-2g-k} \left( \sqrt{\beta} - \frac{1}{\sqrt{\beta}} \right)^k \mathcal{F}^{g,k}(S_{a,V}). \quad (3.74)$$

The theorem completely proved in [10] (and only sketched here) is that there exist constants $\hat{d}^{g,k}$ such that

$$\forall g,k \geq 0 \quad f^{g,k}(a, V) = \mathcal{F}^{g,k}(S_{a,V}) + \hat{d}^{g,k}. \quad (3.75)$$

In other words, there exists a constant $C_{N,\beta}$ depending only on $N$ and $\beta$, such that

$$Z_{N,\beta}(a, V) = C_{N,\beta} \exp (\mathcal{F}_{N,\beta}(S_{a,V})) \quad (3.76)$$

and $\ln C_{N,\beta}$ has a topological expansion, modulo an extra $\ln(N)$ term.

If one wishes to test equation (3.76) with Monte Carlo computations of $Z$, or to compare the predictions for the statistics of the maximal eigenvalue to Tracy–Widom laws, it is important to know the normalization constant $C_{N,\beta}$.

3.8.2. Value of $C_{N,\beta}$. It can be computed explicitly by specializing to some $a$ and $V$ such that:

- one has a close formula for $F^{g,k}$ for any $g,k$.
- one has independently a close formula for the eigenvalue integral.
Comparing the two gives the constant $C_{N,\beta}$. Such expressions are available for the Gaussian potential ($V(x) = x^2/2$) in the limit $a \to -\infty$. In this case indeed the eigenvalue integral can be reduced to a Selberg integral on the one hand, and unstable $F^{g,k}$ vanish on the other hand. We do the computations in appendix A and the result is

$$C_{N,\beta} = e^{D_{N,\beta}}$$

is the normalization constant relating the series constructed out of the geometric quantities to the genuine eigenvalue integral. Its value is given in equation (3.77).

4. Example: Gaussian $\beta$ ensemble

We consider the Gaussian model, i.e. with a quadratic potential $V(x) = x^2/2$:

$$I_{N,\beta}(a) = \int_{[\infty,a]^N} d\lambda_1 \cdots d\lambda_N \left| \Delta(\lambda) \right|^{2\beta} e^{-\left(\beta N/2\right) \sum_{i=1}^N \lambda_i^2}.$$ (4.1)

We also write $I_{N,\beta}$ for the unbounded eigenvalue integral $I_{N,\beta}(\infty)$. Let us denote $F^{g,k}(a)$ the geometric quantities of this model and write

$$F(a) = \sum_{g,k} t^{2-2g} h^k F^{g,k}(a).$$ (4.2)

We are interested in

$$P(a) \equiv P[\lambda_{\max} \leq a] = \frac{I_{N,\beta}(a)}{I_{N,\beta}(\infty)} = \frac{C_{N,\beta}}{I_{N,\beta}} e^{F(a)}.$$ (4.3)

$$C_{N,\beta} = e^{D_{N,\beta}}$$ is the normalization constant relating the series constructed out of the geometric quantities to the genuine eigenvalue integral. Its value is given in equation (3.77).

Notice that $I_{N,\beta}(a)$ is a function of $a$, $t$, but we have a priori a relation of homogeneity:

$$I_{N,\beta}(a) = |a|^{\beta N^2+(1-\beta)N} I\left(t/|a|^2\right)$$ (4.4)

where

$$I(\tau) = \int_{[\infty,|\tau|]} d\lambda_1 \cdots d\lambda_N \left| \Delta(\lambda) \right|^{2\beta} e^{-\left(\beta N/2\right) \sum_{i=1}^N \lambda_i^2}.$$ (4.5)

Thus

$$F^{0,0}(a) = t^2 \ln |a| + \text{function of } t/|a|^2$$ (4.6)

$$F^{0,1}(a) = -t \ln |a| + \text{function of } t/|a|^2$$ (4.7)

else $F^{g,k}$ = function of $t/|a|^2$. (4.8)
4.1. Theoretical results for arbitrary $a$

4.1.1. Density of eigenvalues. The eigenvalue support is a single interval\(^7\) if $a \leq 2\sqrt{t}$. In this case, it is of the form $[b, a]$, where $b = \frac{1}{3}(a - 2\sqrt{a^2 + 12t})$. The parameters $\alpha = (a + b)/2$ and $\gamma = (a - b)/4$ are thus given by

$$
\begin{align*}
\alpha &= \frac{1}{3}(2a - \sqrt{a^2 + 12t}) \\
\gamma &= \frac{1}{6}(a + \sqrt{a^2 + 12t})
\end{align*}
$$

and we have

$$
\begin{align*}
v_0 &= \frac{\alpha^2}{2} + \gamma^2 \\
v_1 &= \alpha \gamma \\
v_2 &= \frac{\gamma^2}{2}
\end{align*}
$$

$$
\begin{align*}
\rho(x) &= -\frac{1}{2\pi} M(x) \sqrt{\frac{x - b}{a - x}} \quad \text{for } x \in [b, a] \\
M(x) &= x - 2\gamma.
\end{align*}
$$

When $a = 2\sqrt{t}$, the eigenvalue support is $[b, a] = [-2\sqrt{t}, 2\sqrt{t}]$ and we recover the Wigner semicircle law (with $t = 1/2$ and $a = \sqrt{2}$ usually):

$$
\rho(x) = \frac{1}{2\pi} \sqrt{4t - x^2}.
$$

For $a < 2\sqrt{t}$, $\rho(x)$ has another zero on the real line at $x = x_1 = 2\gamma$, which corresponds to $s = s_1$:

$$
s_1 = \frac{1}{2\gamma} \left( 2\gamma - \alpha + \sqrt{\alpha(\alpha - 4\gamma)} \right)
$$

defined by $|s_1| \geq 1$ and $x(s_1) = x_1$, where $x(z) = \alpha + \gamma(z + 1/z)$ is the Zhukovsky map. For $a \leq 2\sqrt{t}$, we have $s \in [1, \infty]$.\(^8\)

4.1.2. First terms in $F(a)$. The formulae for the nondecaying terms in the free energy yield

$$
\begin{align*}
F^{0,0}(a) &= \frac{\alpha^2 \gamma^2}{2} + \frac{\gamma^4}{4} - t \left( \gamma^2 + \frac{\alpha^2}{2} \right) + t^2 \ln \gamma \\
F^{0,1}(a) &= t \ln \left( \frac{\gamma^8}{2\pi t} \right) - \left( \alpha \gamma + \frac{\gamma^2}{2} \right) + \left( \frac{\alpha \gamma}{2} + \gamma^2 \right) \frac{1}{s} \\
F^{1,0}(a) &= -\frac{1}{24} \ln \left[ 16t^4 \gamma^4 \alpha^3 (\alpha - 4\gamma) \right] \\
F^{0,2}(a) &= \frac{1}{12} \ln \left( 16t^{-2} \gamma^4 \frac{s^{10}(s - 1)^3}{(s + 1)^{11}} \right).
\end{align*}
$$

Indeed, the dominant term (equation (4.13)) was obtained in [13, 14] for all $\beta$ by Coulomb gas techniques.

\(^7\) We will not analyze the case where $a > 2\sqrt{t}$, which cannot be reached naively as below.
Their derivatives with respect to \( a \) are maybe easier to handle:

\[
\begin{align*}
\frac{\partial}{\partial a} F^{0,0}(a) &= \gamma a^2 \quad (4.17) \\
\frac{\partial}{\partial a} F^{0,1}(a) &= -\frac{1}{2} \left( \alpha + \sqrt{\alpha(a-4\gamma)} \right) \quad (4.18) \\
\frac{\partial}{\partial a} F^{1,0}(a) &= -\frac{\gamma^2}{\alpha(a-4\gamma)^2} \quad (4.19) \\
\frac{\partial}{\partial a} F^{0,2}(a) &= \frac{\alpha s^2(2s^3 - 3s^2 + 4s - 2)}{\gamma^2 (s-1)^3(s+1)^4}. \quad (4.20)
\end{align*}
\]

For example, in terms of \( a \) and \( t \) only:

\[
\begin{align*}
F^{0,0}(2\sqrt{t}) - F^{0,0}(a) &= \frac{1}{3} ta^2 - \frac{1}{216} a^4 - \frac{a}{216}(a^2 + 30t)\sqrt{a^2 + 12t} \\
&\quad - t^2 \ln \left( \frac{a + \sqrt{a^2 + 12t}}{6\sqrt{t}} \right) \quad (4.21)
\end{align*}
\]

\[
\begin{align*}
F^{1,0}(a) &= \frac{7}{24} \ln 3 - \frac{\ln 2}{8} + \frac{\ln t}{6} - \frac{1}{48} \ln(a^2 + 12t) \\
&\quad - \frac{1}{24} \ln \left[ a(-a^6 + 6a^4t + 36a^2t^2 - 432t^3) ight] \\
&\quad + \left( -a^6 + 12a^4t - 54a^2t^2 + 216t^3 \right) \sqrt{a^2 + 12t}. \quad (4.22)
\end{align*}
\]

4.1.3. Example: \( a = 0 \). \( P(0) = (C_{N,\beta}/\mathcal{L}_{N,\beta}) e^{F(0)} \) is the probability that all the eigenvalues are negative. The first terms in \( F(0) \) are

\[
\begin{align*}
F^{0,0}(0) &= t^2 \left( -\frac{3}{4} - \frac{\ln 3}{2} + \frac{\ln t}{2} \right) \quad (4.23) \\
F^{0,1}(0) &= t \left[ \frac{1}{2} + \ln \left( 1 + \frac{2}{\sqrt{3}} \right) - \ln(2\pi) - \frac{\ln t}{2} \right] \quad (4.24) \\
F^{1,0}(0) &= -\frac{\ln 2}{3} + \frac{\ln 3}{8} \quad (4.25) \\
F^{0,2}(0) &= -\frac{3}{8} \ln 3 + \frac{1}{2} \ln \left( 1 + \frac{2}{\sqrt{3}} \right). \quad (4.26)
\end{align*}
\]

4.2. Numerical simulations: principles

To verify the analytical predictions derived in the previous sections, we simulated the joint distribution of eigenvalues in equation (1.2) in the case of the Gaussian ensemble:

\[
\begin{align*}
P(\lambda_1, \ldots, \lambda_N) &= B_{M,N} \prod_{i} e^{(-N\beta/\mathcal{L})(\lambda_i^2/2)} \prod_{i<j} |\lambda_i - \lambda_j|^{2\beta} \\
&= B_{M,N} e^{-\beta E[\{\lambda_i\}]} \quad (4.27)
\end{align*}
\]
where the effective energy \( E[\lambda] \) is given by

\[
E[\{\lambda_i\}] = \frac{N}{t} \sum_{i=1}^{N} \frac{\lambda_i^2}{2} - 2 \sum_{i<j} \ln |\lambda_i - \lambda_j|.
\]  

(4.28)

In the simulations, we fix \( t = 1/2 \). We sample this probability distribution using a Monte Carlo Metropolis algorithm (see [28]).

4.2.1. Standard Metropolis algorithm. We start with an initial configuration of the \( \lambda_i \)'s taken at random within the range \([-2\sqrt{t}, 2\sqrt{t}] = [-\sqrt{2}, \sqrt{2}] \) (the average density support). At each step, a small modification \( \{\lambda_i\} \rightarrow \{\lambda'_i\} \) is proposed in the configuration space. In our algorithm, the proposed move consists of picking at random an eigenvalue \( \lambda_j \) and proposing to modify it as \( \lambda_j \rightarrow \lambda_j + \varepsilon \), where \( \varepsilon \) is a real number drawn from a Gaussian distribution with mean zero and with a variance that is set to achieve an average rejection rate of 1/2 (the proposed move is indeed accepted or rejected as explained below).

The move is accepted with the standard probability

\[
p = \min\left( \frac{P(\{\lambda'_1, \ldots, \lambda'_N\})}{P(\{\lambda_1, \ldots, \lambda_N\}), 1} \right) = \min\left( e^{-\beta(E[\{\lambda'_i\}] - E[\{\lambda_i\})]}, 1 \right),
\]  

(4.29)

and rejected with probability \( 1 - p \). This dynamics enforces detailed balance and ensures that at long times the algorithm reaches thermal equilibrium (at inverse ‘temperature’ \( \beta \)) with the correct Boltzmann weight \( e^{-\beta E[\{\lambda_i\}]}. \)

At long times (from about \( 10^6 \) steps in our case), the Metropolis algorithm thus generates samples of \( \{\lambda_i\} \) drawn from the joint distribution in equation (4.27). We can then start to construct a histogram for the maximal eigenvalue (by keeping the value of \( \lambda_{\text{max}} \) about every 100 steps). This histogram describes the probability density function of \( \lambda_{\text{max}} \), which can be compared to our analytical predictions: \( P(\lambda_{\text{max}} = a) = \partial_a P(a) \) with \( P(a) \) given in equation (C.13) for large \( N \). More precisely, we plot here the rate function \( \Phi(a) = -\ln P(\lambda_{\text{max}} = a) \). We compare the numerical data \( \Phi_{\text{num}}(a) \) with analytical results \( \Phi(a) \).

However, as the distribution of the maximal eigenvalue is highly peaked around its average, a standard Metropolis algorithm does not allow us to explore in a ‘reasonable’ time a wide range of values of \( \lambda_{\text{max}} \). The probability to reach a value \( \lambda_{\text{max}} = a \) decreases rapidly with \( N \) as \( e^{-N^2\Phi_2(a) + O(N)} \), where \( \Phi_2(a) \) is positive (for \( a \) different from the mean value). Therefore, we modified the algorithm in order to explore the full distribution of \( \lambda_{\text{max}} \) and to compare it with our analytical predictions for the (left) large deviation of the distribution of \( \lambda_{\text{max}} \).

4.2.2. Modified algorithm: conditional probabilities. It is difficult to reach small (or large) values of \( \lambda_{\text{max}} \) with a standard Metropolis algorithm. The idea is thus to force the algorithm to explore a region \( \lambda_{\text{max}} \leq a_c \) for different values of \( a_c \). We thus add in the algorithm the constraint \( \lambda_{\text{max}} \leq a_c \). More precisely, we start with an initial configuration that satisfies the constraint \( \lambda_{\text{max}} \leq a_c \). At each step, the proposed move \( \{\lambda_i\} \rightarrow \{\lambda'_i\} \) is rejected if \( \lambda'_{\text{max}} > a_c \). If \( \lambda'_{\text{max}} \leq a_c \), then the move is accepted or rejected exactly with the same Metropolis rules as before. Because of the new constraint \( \lambda_{\text{max}} \leq a_c \), the moves are rejected much more often than before. Therefore the variance of the Gaussian distribution \( P(\varepsilon) \) has to be taken smaller to achieve a rejection rate 1/2.

doi:10.1088/1742-5468/2011/11/P11024
We run the program for several values of \( a_c \) (about 60 different values on the left of the mean value) and we construct a histogram of \( \lambda_{\text{max}} \) for each value \( a_c \). This gives the conditional probability distribution \( P(\lambda_{\text{max}} = a | \lambda_{\text{max}} \leq a_c) \). Again, as the distribution of the maximal eigenvalue is highly peaked, the algorithm can only explore a very small range of values of \( a \), even for a large running time (about \( 10^9 \) steps). The difference with the previous algorithm is that we can now explore small regions of the form \( a_c - \eta \leq a \leq a_c \) for every \( a_c \), whereas before we could only explore the neighborhood of the mean value.

The distribution of the maximal eigenvalue is given by

\[
P(\lambda_{\text{max}} = a) = P(\lambda_{\text{max}} = a | \lambda_{\text{max}} \leq a_c)P(\lambda_{\text{max}} \leq a_c) \quad \text{(for} \ a < a_c) \tag{4.30}.
\]

Therefore the rate function is

\[
- \ln P(\lambda_{\text{max}} = a) = - \left[ \ln P(\lambda_{\text{max}} = a | \lambda_{\text{max}} \leq a_c) + \ln P(\lambda_{\text{max}} \leq a_c) \right]. \tag{4.31}
\]

The histogram constructed by the algorithm with the constraint \( \lambda_{\text{max}} \leq a_c \) is the rate function \( \Phi_{a_c}(a) \equiv - \ln P(\lambda_{\text{max}} = a | \lambda_{\text{max}} \leq a_c) \). It differs from the exact rate function \( \Phi(a) = - \ln P(\lambda_{\text{max}} = a) \) by an additive constant that depends on \( a_c \). In order to get rid of this constant, we construct from the histogram giving \( \Phi_{a_c}(a) \) the derivative \( \frac{d\Phi_{a_c}(a)}{da} \). This derivative is equal to \( \frac{d\Phi(a)}{da} \) and the constants disappear.

### 4.3. Numerical simulations: results

We now come back to \( \Phi(a) \) from its derivative using an interpolation of the data for the derivative and a numerical integration of the interpolation. This allows us to compare directly the numerical results \( \Phi_{\text{num}}(a) \) with the theoretical rate function \( \Phi(a) = - \ln P(\lambda_{\text{max}} = a) = - \ln [\partial_a P(a)] \). The interpolation and numerical integration also help us to get rid of the fluctuations in numerical data (by smoothening the plot) and thus increase the precision of the results.

We can follow the same steps to explore the region on the right of the mean value by adding in the simulations the condition \( \lambda_{\text{max}} \geq a_c \) (instead of \( \lambda_{\text{max}} \leq a_c \)) for several values of \( a_c \). At the end, we can reconstruct the full plot of the probability density function (pdf) of \( \lambda_{\text{max}} \). There remains only an additive constant in the rate function that is not yet determined (as we perform a numerical integration of the derivative). This constant can be numerically determined by imposing that the integral (numerical integration of the data) of the pdf of \( \lambda_{\text{max}} \) is one (normalization of the probability). We can now compare the simulations with analytical predictions.

Our analytical predictions give (see equation (C.13))

\[
P(a) = \frac{\mathcal{I}_{N,\beta}(a)}{\mathcal{I}_{N,\beta}(\infty)} = \frac{C_{N,\beta}}{\mathcal{I}_{N,\beta}(\infty)} e^{F(a)} = \exp \left\{ F(a) + \ln C_{N,\beta} - \ln \mathcal{I}_{N,\beta} \right\} \tag{4.32}
\]

with

\[
\ln C_{N,\beta} - \ln \mathcal{I}_{N,\beta} = \beta N^2 \left( \frac{3}{4} - \frac{\ln t}{2} \right) + N(\beta - 1) \left( -\frac{1}{2} + \ln(2\pi) + \frac{\ln t}{2} \right) \\
+ \ln N \left( \frac{\beta + \beta^{-1} - 3}{12} \right) + \left( \frac{5 - 2(\beta + \beta^{-1})}{6} \right) \ln 2 \\
- \frac{\ln(2\pi)}{2} - \frac{\ln \beta}{2} + \kappa_{\beta} \tag{4.33}
\]

doi:10.1088/1742-5468/2011/11/P11024

and $F(a)$ is given up to order one (for large $N$) by
\begin{equation}
F(a) = \frac{N^2}{t^2} \beta F^{0,0}(a) + N \frac{1}{t} (\beta - 1) F^{0,1}(a) + \frac{1}{2} F^{1,0}(a) + (\beta + \beta^{-1} - 2) F^{0,2}(a).
\end{equation}

Thus
\begin{align}
\Phi(a) &\equiv -\ln \mathbb{P}(\lambda_{\text{max}} = a) = -\ln [\partial_a \mathcal{P}(a)]
\notag\\notag \\
&= -\{F(a) + \ln C_{N,\beta} - \ln \mathcal{I}_{N,\beta} + \ln [\partial_a F(a)]\}
\notag\notag \\
&= N^2 \Phi_2(a) + N \Phi_1(a) + \ln N \phi + \Phi_0(a) + \cdots
\end{align}

where
\begin{align}
\Phi_2(a) &= -\frac{1}{t^2} \beta F^{0,0}(a) - \beta \left(\frac{3}{4} - \frac{\ln t}{2}\right)
\notag\notag \\
\Phi_1(a) &= -\frac{1}{t} (\beta - 1) F^{0,1}(a) - (\beta - 1) \left(-\frac{1}{2} + \ln(2\pi) + \frac{\ln t}{2}\right)
\notag\notag \\
\phi &= -\left(\frac{\beta + \beta^{-1} - 3}{12}\right) - 2
\notag\notag \\
\Phi_0(a) &= -F^{1,0}(a) - (\beta + \beta^{-1} - 2) F^{0,2}(a) - \ln \left(\partial_a F^{0,0}(a)\right)
\notag\notag \\
&\quad - \left(\frac{5 - 2(\beta + \beta^{-1})}{6}\right) \ln 2 + \frac{\ln(2\pi)}{2} + \frac{\ln \beta}{2} - \kappa_\beta - \ln \left(\frac{\beta}{t^2}\right).
\end{align}

We typically run the simulations for $N = 30$ with $10^9$ iterations.

Figure 1 shows a plot of the rate function $\Phi_{\text{num}}(a)/N^2 \equiv -\ln P(\lambda_{\text{max}} = a)/N^2$ for different values of $\beta$: in red, $\beta = \frac{5}{8} = 0.625$; in blue, $\beta = 1$ (Gaussian unitary ensemble) and in orange, $\beta = \frac{8}{5} = 1.6$. Numerical data $\Phi_{\text{num}}(a)$ (discs, squares and triangles) are compared with analytical predictions for the left tail $\Phi(a) = N^2 \Phi_2(a) + N \Phi_1(a) + (\ln N) \phi + \Phi_0(a)$ (solid lines). In this figure, we see that the agreement between numerics and theory is very good, at least to dominant order for large $N$ (order $N^2$).

Figure 2 shows the next order in the large $N$ expansion (order $N$) and figure 3 shows the order one for large $N$. Up to order $O(N)$, the agreement between numerics and theory is very good (see figures 1 and 2). To order $O(1)$, the theoretical lines do not fit the numerical data as well as to order $O(N)$ or $O(N^2)$ but the agreement is still quite good. At this order, the number of significant digits that is required is very high. A better precision would require a longer running time (to have more samples) and a refined histogram. Also, the determination of the constant depends on the reconstruction of the full histogram (both left and right tails). However, we cannot go numerically toward infinity, we have to choose a finite interval. The precision could thus be increased by exploring a wider range of values (by going deeper in the left and right tails).

5. Scaling limit at the edge of the spectrum

Now, we wish to take the scaling limit $N \rightarrow \infty, a \rightarrow \hat{a}$ the edge of the spectrum, in a one-cut model with arbitrary polynomial potential. Let us mention that using an asymptotic expansion of eigenvalue statistics to study limiting distribution is not a new idea. For instance, within GUE, Gustavsson [21] has found limiting distributions for the $k$th eigenvalue in various regimes, using an asymptotic expansion for the first few terms of eigenvalue correlation densities.
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Figure 1. Case of a quadratic potential (Gaussian ensemble): rate function
\[ \frac{\Phi(a)}{N^2} = - \frac{\ln P(\lambda_{\text{max}} = a)}{N^2} \]
plotted against \( a \) for different values of \( \beta \): \( \beta = 5/8 = 0.625 \) (red discs and red line), \( \beta = 1 \) (blue squares and blue line) and \( \beta = 8/5 = 1.6 \) (orange triangles and orange line). The points (discs, squares, triangles) are numerical data obtained with the modified Metropolis algorithm. The solid lines are the analytical predictions for the left tail of the large deviation of the pdf of \( \lambda_{\text{max}} \): \( P(a) \) is given in equation (C.13).

5.1. Blow up of the spectral curve and unstable correlators

Formula (3.68) computes the stable \( F_{g,k}^{g}(a) \) only in terms of the spectral curve \( y, \omega_{2,0} \) and \( \omega_{1,0}^{\pm} \):

\[
(S_{a}) : \quad \begin{cases}
  x(z) &= a + \frac{(z-1)^2}{\gamma} \\
  y(z) &= \frac{z+1}{2} \prod_{j=1}^{d-1} \frac{z-s_{j}}{(z-1/(s_{j})}
\end{cases}
\]

\( \omega_{1,0}^{\pm}(z) = \frac{1}{2(z-1)} - \frac{1}{2(z+1)} - \sum_{j=1}^{d-1} \frac{1}{(z-1/s_{j}) - 1/z} \) (5.2)

\( \omega_{2,0}^{0}(z_{1},z_{2}) = \frac{1}{(z_{1}z_{2} - 1)^{2}} \) (5.3)

We shall use the notation \( F_{g,k}^{g}(a) = F_{g,k}(\Sigma_{a}) \) for the geometric quantities associated with the curve \( \Sigma_{a} \).

Let us define \( \epsilon \) such that \( s_{1} - 1 \sim \sqrt{\epsilon} \) and \( \epsilon \to 0 \) when \( a \to a^{*} \). Let us also assume that \( \gamma \to \bar{\gamma} \neq 0 \) and \( s_{j} \to \bar{s}_{j} \neq 1 \) for \( 2 \leq j \leq d-1 \). See figure 4. With the rescaling \( z = 1 + \sqrt{\epsilon} \sigma \),

\[ \text{doi:10.1088/1742-5468/2011/11/P11024} \]
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Figure 2. Case of a quadratic potential (Gaussian ensemble): order $N$ of the rate function, $(\Phi_{\text{num}}(a) - N^2 \Phi_2(a))/N$, plotted against $a$ for different values of $\beta$: $\beta = 5/8 = 0.625$ (red), $\beta = 1$ (blue) and $\beta = 8/5 = 1.6$ (orange). The points (discs, squares, triangles) are numerical data obtained with the modified Metropolis algorithm. The solid lines are the analytical predictions: $(N \Phi_1(a) + (\ln N) \phi + \Phi_0(a))/N$.

Figure 3. Case of a quadratic potential (Gaussian ensemble): order one of the rate function, $\Phi_{\text{num}}(a) - N^2 \Phi_2(a) - N \Phi_1(a) - \ln N \phi$, plotted against $a$ for different values of $\beta$: $\beta = 5/8 = 0.625$ (red), $\beta = 1$ (blue) and $\beta = 8/5 = 1.6$ (orange). The points (discs, squares, triangles) are numerical data obtained with the modified Metropolis algorithm. The solid lines are the analytical predictions: $\Phi_0(a)$. 

doi:10.1088/1742-5468/2011/11/P11024
Figure 4. When a moving singularity \((x = x(s_i)\) here) merges to a fixed singularity (the hard edge at \(x = a\)), the blow up around \(a\) of the spectral curve (here, of the form equation \((5.1)\)) can be generically described by the situation on the left. Indeed, \((\hat{\Sigma}) = (C, \hat{x}, \hat{y})\) is a universal object. We may call it the Tracy–Widom curve. In our parameterization, the physical sheet corresponds to \(\text{Re} \sigma \geq 0\). There is only one simple branch point, at \(\sigma = 0\), and we have \(\hat{\sigma} = -\sigma\) (which is globally defined). \(y\) has a pair of zeros at \(\sigma = -1, 1\).

we have

\[
(\Sigma_a) \sim \begin{cases} 
  x(z) = (a - 2\epsilon \hat{\gamma}) + \hat{\gamma}\epsilon \hat{x}(\sigma) \\
  y(z) = M_+ \epsilon^{1/2} \hat{y}(\sigma)
\end{cases}
\]  

(5.4)

\[
\hat{\omega}_1^{0,1}(z) \sim \epsilon^{-1/2} \hat{\omega}_1^{0,1}(\sigma)
\]  

(5.5)

\[
\hat{\omega}_2^{0,0}(z_1, z_2) \sim \epsilon^{-1} \hat{\omega}_2^{0,0}(\sigma_1, \sigma_2)
\]  

(5.6)

with the constant

\[
M_\pm = t_d \hat{\gamma}^{d-1} \prod_{j=2}^{d-1} (\pm 1 - \hat{s}_j)(\pm 1 - 1/\hat{s}_j).
\]

We have

\[
(\hat{\Sigma}) : \begin{cases} 
  \hat{x}(\sigma) = \sigma^2 \\
  \hat{y}(\sigma) = \sigma - \frac{1}{\sigma}
\end{cases}
\]  

(5.7)

\[
\hat{\omega}_1^{0,1}(\sigma) = \frac{1}{2\sigma} - \frac{1}{\sigma + 1}
\]  

(5.8)

\[
\hat{\omega}_2^{0,0}(\sigma_1, \sigma_2) = \frac{1}{(\sigma_1 + \sigma_2)^2}.
\]  

(5.9)

In other words, \(\hat{\Sigma}\) is the plane curve of equation \(y^2 = x + 1/x - 2\).
5.2. Scaling limit of unstable $F^{g,k}$

It is easy to obtain the scaling limit of unstable free energies from their expressions. We find

$$F^{0,0}(a) = F^{0,0}(a) + ((\hat{\gamma} \hat{M}_+)^{2/3} \epsilon)^2 F^{0,0} + o(\epsilon^2)$$

$$F^{0,1}(a) = F^{0,1}(a) + ((\hat{\gamma} \hat{M}_+)^{2/3} \epsilon)^{3/2} F^{0,1} + o(\epsilon^{3/2})$$

$$F^{1,0}(a) = \ln((\hat{\gamma} \hat{M}_+)^{2/3} \epsilon) \hat{F}^{1,0} + 0(1)$$

$$F^{0,2}(a) = \ln((\hat{\gamma} \hat{M}_+)^{2/3} \epsilon) \hat{F}^{0,2} + 0(1)$$

with

$$\hat{F}^{0,0} = -\frac{2}{3},$$

$$\hat{F}^{0,1} = \frac{4}{3},$$

$$\hat{F}^{1,0} = -\frac{1}{8}, \quad \hat{F}^{0,2} = \frac{1}{8}$$

and

$$\delta^{1,0} = -\frac{1}{24} \ln\left(2^{10} \hat{M}_+ \hat{M}_- \hat{\gamma}^2 t^{-4}\right)$$

$$\delta^{0,2} = \frac{1}{12} \ln\left[2^{-7} t_a \hat{\gamma}^d t^{-2} \prod_{j=2}^{d-1} \frac{s_j^{23} (s_j - 1)^{11}}{(s_j + 1)^{11}}\right] - \sum_{2 \leq j < l \leq d-1} \ln\left(1 - \frac{1}{s_j s_l}\right)$$

$\hat{F}^{0,0}, \hat{F}^{0,1}, \hat{F}^{1,0}$ and $\hat{F}^{0,2}$ are universal and can be considered as the unstable free energies associated with the curve $\hat{\Sigma}$. In contrast, $\delta^{1,0}$ and $\delta^{0,2}$ are non-universal. They are relevant to compute the constant prefactor in the scaling limit of $F(a)$.

5.3. Scaling limit of stable $F^{g,k}$

The scaling of the correlators is very simple to obtain from the residue formula (equation (3.49)). The scaling of unstable correlators is computed at hand, and the result for stable correlators is proved by recursion. $\omega^{0,0}$ is special:

$$\omega^{0,0}(z) = \frac{\hat{\gamma}}{2} V'(\hat{a}) \epsilon^2 - \hat{\gamma} M_+ \epsilon^{3/2} \hat{y}(\sigma) + O(\epsilon^2).$$

Also, if $(n, g, k) \neq (1, 0, 0)$:

$$\omega^{g,k}(z_1, \ldots, z_n) = (\hat{\gamma} \hat{M}_+ \epsilon^{3/2})^{2-2g-k-n} \hat{\omega}^{g,k}(\sigma_1, \ldots, \sigma_n) + O(\epsilon^{2-2g-k+n}).$$

The $\hat{\omega}^{g,k}$ are the result of the residue formula applied to $\hat{\Sigma}$ and the data of $\omega^{0,0}$ and $\omega^{0,1}$. The standard form of the topological recursion (formula (3.59)) computes $\hat{\omega}^{g,k}$ with the limit of the recursion kernel:

$$K(\sigma, \sigma_0) = \frac{1}{4 \sigma (\sigma^2 - 1)(\sigma^2 - \sigma_0)}.$$

In the stable $\omega^{g,k}$, the leading behavior corresponds to the diverging power of $\epsilon$. Thus, only the poles at infinitesimal distance of the hard edge are relevant to find $\hat{\omega}^{g,k}$. In other
words, the poles at \( z_i = 1 \) are replaced by poles at \( \sigma_i = 0 \), the poles at \( z_i = 1/s_1 \) are replaced by poles at \( \sigma_i = -1 \) and the poles at \( z_i = 1/z_l \) are replaced by poles at \( \sigma_i = -\sigma_l \), whereas the poles at \( z = 0 \) and \( 1/s_j \) do not give contributions to \( \hat{\omega}_{g,k} \).

The integration formula for stable \( F^{g,k} \) (equation (3.68)) yields the scaling

\[
F^{g,k}(a) = (\hat{\gamma} \hat{M}_+ \epsilon^{3/2})^{2-2g-k} \hat{F}^{g,k}
\]

where \( \hat{F}^{g,k} \) are the geometric quantities associated with the curve \( \hat{\Sigma} \):

\[
\hat{F}^{g,k} = \frac{1}{2 - 2g - k} \text{Res}_{\sigma \rightarrow 0} d\sigma \hat{\phi}(\sigma) \hat{\omega}^{g,k}_1(\sigma)
\]

\( \hat{\phi} \) is the following primitive of \( \hat{y} d\hat{x} \):

\[
\hat{\phi}(\sigma) = \int_0^\sigma \hat{y} d\hat{x}
\]

\[
= 2 \left( \frac{\sigma^3}{3} - \sigma \right).
\]

### 5.4. Summary: double scaling limit

To find the scaling limit of \( F(a) \) when \( a \) goes to the edge of the spectrum, a scaling variable appears naturally:

\[
\eta = \left( \frac{\hat{\gamma} \hat{M}_+}{t} \right)^{2/3} N^{2/3} \epsilon
\]

Notice that the \( N^{2/3} \) scaling is the same for all \( \beta \), as pointed out long ago and proved recently in [23]. When \( \eta \) is of order 1, each term in the topological expansion of \( F(a(\eta)) \) is also of order 1. It is the so-called ‘double scaling limit’:

\[
F(a) = \sum_{g,k \geq 0} \left( \frac{N}{t} \right)^{2-2g-k} \beta^{1-g}(1 - \beta^{-1})^k \hat{F}^{g,k}(\Sigma_a)
\]

\[
\sim \frac{N^2 \beta}{t^2} F^{0,0}(\hat{a}) + \frac{N(\beta - 1)}{t} F^{0,1}(\hat{a}) + \frac{3 - \beta - \beta^{-1}}{12} \ln \left( \frac{N \hat{\gamma} \hat{M}_+}{t} \right) + \delta^{1,0} + (\beta + \beta^{-1} - 2) \delta^{0,2} + \sum_{g,k \geq 0} \beta^{1-g}(1 - \beta^{-1})^k \eta^{(3/2)(2-2g-k)} \hat{F}^{g,k}
\]

where it is implied that \( \eta^0 \) should be replaced by \( \ln(\eta) \). In particular, for Hermitian matrices (\( \beta = 1 \)):

\[
F(a) \sim \frac{N^2}{t^2} F^{0,0} + \frac{1}{12} \ln \left( \frac{N \hat{\gamma} \hat{M}_+}{t} \right) + \delta_{1,0} + \sum_{g \geq 0} \eta^{3(1-g)} \hat{F}^{g,0}.
\]
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The curve \( \hat{\Sigma} \) was described in section 5.1. We call the curve the Tracy–Widom curve, because its geometric quantities \( \hat{F}^{g,k} \) are related to the coefficients of the left tails of the Tracy–Widom laws (see section 5.6.2).

Eventually, the complete answer for \( \mathcal{P}(a) = \mathbb{P}(\lambda_{\text{max}} \leq a) \) is

\[
\ln \mathcal{P}(a) = \ln(C_{N,\beta}) - \ln(I_{N,\beta}) + \frac{N^2 \beta}{t^2} \hat{F}^{0,0}(\hat{a}) + \frac{N(\beta - 1)}{t} \hat{F}^{0,1}(\hat{a}) \\
+ \frac{3 - \beta - \beta^{-1}}{12} \ln \left( \frac{N\gamma M_+}{t} \right) + \delta^{1,0} + (\beta + \beta^{-1} - 2)\delta^{0,2} \\
\times \sum_{g,k \geq 0} \beta^{1-g}(1 - \beta^{-1})^k \eta^{(3/2)(2-2g-k)} \hat{F}^{g,k}. \tag{5.28}
\]

Notice a posteriori that \( \hat{a} \) must be such that the divergent terms when \( N \to \infty \) cancel in this expression.

5.5. First few \( \hat{F}^{g,k} \)

We give a few of these geometric quantities associated with \( \hat{\Sigma} \). We have already seen

\[
\hat{F}^{0,0} = -\frac{2}{3}, \quad \hat{F}^{0,1} = \frac{4}{3}, \quad \hat{F}^{1,0} = -\frac{1}{12}, \quad \hat{F}^{0,2} = \frac{1}{12}.
\]

We can also compute

\[
\hat{F}^{1,1} = \frac{31}{2^7 \times 3}, \quad \hat{F}^{0,3} = -\frac{9}{2^6}, \quad \hat{F}^{2,0} = \frac{3}{2^9}, \\
\hat{F}^{1,2} = \frac{487}{2^{12} \times 3}, \quad \hat{F}^{0,4} = -\frac{595}{2^{11} \times 3}, \quad \hat{F}^{2,1} = \frac{8831}{2^{17} \times 3}, \\
\hat{F}^{1,3} = \frac{9281}{2^{17} \times 3^2}, \quad \hat{F}^{0,5} = -\frac{19977}{2^{16} \times 3}, \quad \hat{F}^{3,0} = \frac{63}{2^{14}}, \\
\hat{F}^{2,2} = \frac{62761}{2^{20}}, \quad \hat{F}^{1,4} = -\frac{539193}{2^{23}}, \quad \hat{F}^{0,6} = -\frac{577879}{2^{22}}.
\]

5.6. Example: Gaussian \( \beta \) ensemble

To illustrate the general framework explained above, we carry on direct computation in the Gaussian case from the results of section 4. The edge of the spectrum is located at \( \hat{a} = 2\sqrt{t} \). This value is actually the mean value of the largest eigenvalue of the Gaussian model:

\[
\lim_{N \to \infty} \mathbb{E}(\lambda_{\text{max}}) = 2\sqrt{t}. \tag{5.29}
\]

\( F(a) \) gives the large deviation of the distribution of \( \lambda_{\text{max}} \) on the left of the mean value, which describes atypical large fluctuations of order 1. On the other hand, it is known that the typical small fluctuations of order \( N^{-2/3} \) around the mean value are described by the Tracy–Widom distributions. We expect that the two regimes match smoothly and thus that a scaling limit of \( F(a) \) when \( a \to 2\sqrt{t} \) coincides with the limit \( \xi \to -\infty \) of the Tracy–Widom distribution \( F_{2\beta}(\xi) \). More precisely, we write

\[
a = 2\sqrt{t}(1 - \epsilon) \tag{5.30}
\]

doi:10.1088/1742-5468/2011/11/P11024

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and the correspondence
\[ \epsilon = -c_\beta N^{-2/3} \xi \]  
(5.31)
is expected for some constant \( c_\beta > 0 \).

5.6.1. First terms in \( F(a) \). For bookkeeping, let us write several useful expansion when \( \epsilon \to 0 \):

\[ \alpha = -\sqrt{t} \epsilon \left( 1 + \frac{\epsilon}{8} + \frac{\epsilon^2}{32} + \frac{\epsilon^3}{512} \right) + O(\epsilon^5) \]  
(5.32)
\[ \gamma = \sqrt{t} \left( 1 - \frac{\epsilon}{2} + \frac{\epsilon^2}{16} + \frac{\epsilon^3}{64} + \frac{\epsilon^4}{1024} \right) + O(\epsilon^5). \]  
(5.33)
The other zero of \( \rho(z) \) outside the unit disc is
\[ s_1 = 1 + \sqrt{\epsilon} + \frac{\epsilon}{2} + \frac{7\epsilon^{3/2}}{16} + \frac{5\epsilon^2}{16} + \frac{103\epsilon^{5/2}}{512} + \frac{9\epsilon^3}{64} + \frac{615\epsilon^{7/2}}{8192} + \frac{45\epsilon^4}{1024} + \frac{10187\epsilon^{9/2}}{524288} + O(\epsilon^5). \]  
(5.34)
We see that the definition of \( \epsilon \) through equation (5.30) is compatible with the notation in section 5.1. Moreover, the natural scaling variable is \( \eta = t^{-1/2}\epsilon \). We have for the nondecaying terms of \( F(a) \)

\[ F^{0,0}(2\sqrt{t}) = t^2 \left( -\frac{3}{4} + \frac{\ln t}{2} \right) \]  
(5.35)
\[ F^{0,1}(2\sqrt{t}) = t \left( \frac{1}{2} - \ln(2\pi) - \frac{\ln t}{2} \right) \]  
(5.36)
\[ F^{1,0}(2\sqrt{t}), \quad F^{0,2}(2\sqrt{t}) \quad \text{diverges logarithmically} \]
\[ F^{0,0}(2\sqrt{t}) \quad \text{diverges logarithmically}. \]

Also

\[ F^{0,0}(a) - F^{0,0}(2\sqrt{t}) = t^2 \left( \frac{2\epsilon^3}{3} + \frac{\epsilon^4}{8} \right) + O(\epsilon^5) \]  
(5.37)
\[ F^{0,1}(a) - F^{0,1}(2\sqrt{t}) = t \left( \frac{4\epsilon^{3/2}}{3} - \frac{\epsilon^2}{2} - \frac{\epsilon^{5/2}}{20} - \frac{\epsilon^3}{24} \right. \]
\[ + \frac{23\epsilon^{7/2}}{896} - \frac{\epsilon^4}{128} + \frac{143\epsilon^{9/2}}{18432} \) + O(\epsilon^5) \]  
(5.38)
\[ F^{1,0}(a) = -\frac{\ln(\epsilon)}{8} - \frac{\ln 2}{4} + \frac{5\epsilon}{4} + \frac{5\epsilon^2}{64} + \frac{5\epsilon^3}{1024} - \frac{5\epsilon^4}{2048} - \frac{147\epsilon^5}{131072} + O(\epsilon^5) \]  
(5.39)
\[ F^{0,2}(a) \]
\[ = \frac{\ln \epsilon}{8} - \frac{7\ln 2}{12} + \frac{\sqrt{\epsilon}}{8} + \frac{13\epsilon}{64} + \frac{13\epsilon^{3/2}}{96} - \frac{69\epsilon^2}{1024} + \frac{147\epsilon^{5/2}}{5120} - \frac{97\epsilon^3}{6144} \]
\[ - \frac{103\epsilon^{7/2}}{114688} + \frac{83\epsilon^4}{131072} - \frac{35005\epsilon^{9/2}}{9437184} + O(\epsilon^5). \]  
(5.40)

\( \text{doi:10.1088/1742-5468/2011/11/P11024} \)
5.6.2. Scaling regime of $F(a)$. We have argued in section 5 that, when $\epsilon$ is of order $N^{-2/3}$, all the terms in $F(a)$ become of order 1. This can be checked explicitly in equations (5.37)–(5.40). $N^{-2/3}$ is indeed the typical scale expected of small fluctuations around $E(\lambda_{\text{max}})$. Let us write $\mathcal{P}(a)$ in the scaling variable $\eta$ such that

$$a = \sqrt{t}(2 - 2N^{-2/3}\eta)$$

and taking into account the first terms $F^{0,0}$, $F^{1,0}$, $F^{0,1}$ and $F^{0,2}$. When we use the value in the large $N$ limit of $C_{N,\beta}$ (appendix C.3) and $I_{N,\beta}$ (appendix C.2), we find as expected that the following limit exists:

$$\lim_{N \to \infty} \mathcal{P}[a = 2\sqrt{t}(1 - N^{-2/3}\eta)] = \mathcal{P}_{\beta\text{SE}}^*(\eta).$$

For a better comparison to the literature, we set $\xi = -2\eta$ and we find

$$\mathcal{P}_{\beta\text{SE}}^*(\eta) = \tau_\beta \exp \left[ -\frac{\beta|\xi|^3}{12} + \frac{\sqrt{2}(\beta - 1)}{3}|\xi|^{3/2} + \frac{\beta + \beta^{-1} - 3}{8} \ln|\xi| + O(|\xi|^{3/2}) \right]$$

$$\tau_\beta$$ is a constant given by

$$\ln \tau_\beta = \left( \frac{17}{8} - \frac{25}{24}(\beta + \beta^{-1}) \right) \ln 2 - \frac{\ln(2\pi)}{2} - \frac{\ln \beta}{2} + \kappa_\beta$$

$\kappa_\beta$ is the finite part in the large $N$ asymptotic expansion of $\sum_{j=1}^{N} \ln \Gamma(1 + j\beta)$. It is studied in appendix C.1 and computed in a somewhat explicit way for rational values of $\beta$ in appendix C.4.2.

| Value of $\beta$ | $\ln(\tau_\beta)$ | Numerical value of $\tau_\beta$ |
|-----------------|-------------------|---------------------------------|
| 1               | $-\frac{\ln 2}{12} + \zeta'(1)$ | 0.8723714 |
| $\frac{1}{2}$   | $-\frac{37 \ln 2}{48} + \frac{\zeta'(-1)}{2}$ | 0.5395545 |
| $\frac{1}{3}$   | $-\frac{97 \ln 2}{72} - \frac{7 \ln 3}{36} + \frac{\ln(2\pi)}{6} + \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{3}$ | 0.3071491 |
| $\frac{1}{4}$   | $-\frac{87 \ln 2}{32} - \frac{\ln(2\pi)}{4} + \frac{\ln \Gamma(1/4)}{2} + \frac{\zeta'(-1)}{4}$ | 0.1752911 |
| $\frac{1}{2}$   | $-\frac{11 \ln 2}{48} + \frac{\zeta'(-1)}{2}$ | 0.7854042 |
| $\frac{1}{3}$   | $-\frac{97 \ln 2}{72} + \frac{5 \ln 3}{18} - \frac{\ln(2\pi)}{6} + \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{3}$ | 0.5160081 |
| $\frac{1}{4}$   | $-\frac{185 \ln 2}{96} - \frac{\ln(2\pi)}{4} + \frac{\ln \Gamma(1/4)}{2} + \frac{\zeta'(-1)}{4}$ | 0.3034417 |
| $\frac{2}{3}$   | $-\frac{65 \ln 2}{144} + \frac{2 \ln 3}{3} + \frac{\ln(2\pi)}{6} - \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{6}$ | 0.8882751 |
| $\frac{3}{2}$   | $\frac{17 \ln 2}{144} - \frac{25 \ln 3}{72} + \frac{\ln(2\pi)}{6} - \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{6}$ | 0.7051367 |

doi:10.1088/1742-5468/2011/11/P11024
5.6.3. Comparison to Tracy–Widom law. This expansion should match the left tail of the Tracy–Widom law [24, 25] for the usual values of $\beta = 1/2, 1, 2$. In fact, the leading term in $t^3$ was obtained in [13, 14] from equation (4.13). The correspondence between the scaling variable $\xi$ in the large deviation and the Tracy–Widom variable $\chi$ should be the following:

- Gaussian unitary ensemble ($\beta = 1$):
  \[ F_2(\chi) = \mathcal{P}_{\text{GUE}}^*(-\chi). \]  

If we specialize equation (5.43), we obtain
\[ F_2(\chi) = \chi \to -\infty \frac{2^{1/24} \exp(-|\chi|^3/12)}{|\chi|^{1/8}} \left(1 + \mathcal{O}(|\chi|^{-3})\right). \]  

Notice that only $F_{g,0}$ are involved in equation (5.28) when $\beta = 1$, so $\mathcal{O}(|\chi|^{-3/2})$ is, in fact, a $\mathcal{O}(|\chi|^{-3})$ according to equation (5.21).

- Gaussian orthogonal ensemble ($\beta = 1/2$):
  \[ F_1(\chi) = \mathcal{P}_{\text{GOE}}^*(-\chi). \]  

We obtain
\[ F_1(\chi) = \chi \to -\infty \frac{2^{-11/48} \exp(-|\chi|^3/12)}{|\chi|^{1/16}} (1 + \mathcal{O}(|\chi|^{-3/2})). \]

- Gaussian symplectic ensemble ($\beta = 2$):
  \[ F_4(\chi) = \mathcal{P}_{\text{GSE}}^*(-2^{-2/3} \chi) \]  

We obtain
\[ F_4(\chi) = \chi \to -\infty \frac{2^{-35/48} \exp(-|\chi|^3/12)}{|\chi|^{1/16}} (1 + \mathcal{O}(|\chi|^{-3/2})). \]

These expansions match earlier results. Usually, the constant term is considered as the difficult part of the asymptotics. For $\beta = 1$, this constant was first obtained by Deift, Its and Krasovsky by Riemann–Hilbert asymptotic analysis [15]. For $\beta = 1/2$ and 2, it was determined by Baik, Buckingham and DiFranco by a technique allowing representation of total integrals of the Hastings–MacLeod solution of Painlevé II [5]. In these works, as in ours, an important step was the evaluation of the asymptotic of a Selberg-type integral (see section 3.8 and appendix A). We remark that, for $\beta \neq 1/2, 1, 2$, the choice of the normalization of the scaling variable is somewhat arbitrary, since at present there is no other definition of a ‘$\beta$ Tracy–Widom’ distribution than
\[ "F_{2\beta} = \mathcal{P}_{\text{G}\beta\text{E}}^". \]

Our $\tau_\beta$ gives the constant of $F_{2\beta}(\chi) = \mathcal{P}_{\text{G}\beta\text{E}}^*(\eta = -\chi/2)$.
6. Conclusion

- We showed how to determine to all orders the left large deviation function of the maximal eigenvalue of a random matrix in the $\beta$ ensemble, for all values of $\beta$. We have presented in detail the case where the large $N$ spectrum is connected (one-cut case), but the method combined with existing literature would give as well the multi-cut case. We performed numerics to check our results on the case of the Gaussian model at orders $O(N^2)$ and $O(N)$ with a very good agreement, and at order $O(1)$ with a less convincing agreement.

- We described the repartition function $F_{2\beta}(\chi)$ of the maximal eigenvalue when it reaches the edge of the spectrum from the left (existence of this distribution was proved in [23] with probabilistic methods). More precisely, we gave its full asymptotic expansion when $\chi \to -\infty$ in terms of geometric quantities associated with a spectral curve of the equation

$$\hat{\Sigma} : y^2 = x + \frac{1}{x} - 2.$$ (6.1)

Since it is well known that $F_{1/2}$, $F_1$, $F_2$ are the Tracy–Widom laws, we have related these Tracy–Widom laws, defined from the world of Painlevé equations and integrable systems, to the topological recursion defined from algebraic geometry. For $\beta = 1$, in a related work [7], two of us proved directly that the resummation of the symplectic invariants of $S^*$ are solution to an ODE related to Painlevé II, as it should be according to the work of Tracy and Widom. For general $\beta$, the topological recursion allows us to define a $F_{2\beta}$ for any value of $\beta$. Here, we have obtained the three first terms of its left tail, the constant term, and we may compute recursively the next terms. We notice that the constant term was computed in three parts, similarly to [5] where the problem was addressed for $\beta = 1/2, 1, 2$. Eventually, it would be interesting to relate the $F_{g,k}[S^*]$, to the parabolic differential equation obtained recently in [6] from which the Tracy–Widom $\beta$ distribution can be extracted. Nevertheless, at present, nothing is known about the relation between integrable systems and $\beta$ matrix models.

- From the mathematical point of view, our work is based on two assumptions: existence of a $1/N$ expansion and possibility to take a double scaling limit. The first one will be justified in a forthcoming work [8], whereas the second one seems hard to justify rigorously at present in the absence of integrability. If one accepts the $1/N$ expansion, the loop equations imposes the special $\beta$ dependence of the expansion. From this, one observes the following duality $\beta \leftrightarrow 1/\beta$ at the level of left-tail asymptotics:

$$P_{G_{1/\beta}E}(\chi) = \frac{e^{K_{1/\beta}}}{\beta K_{1/\beta}} \tilde{P}_{G_{1/\beta}E}(\beta^{2/3} \chi)$$ (6.2)

where $\tilde{\cdot}$ is obtained by taking the other branch of the square root (all half-integer powers of $s$ of the expansion appear with a minus sign). Such a duality is readily observed on the known results for $F_1$ and $F_4$ (compare equations (5.48) and (5.50)). Whether this duality has a counterpart for the functions themselves is an open question.

- One could ask for a combinatorial interpretation of the symplectic invariants of the Tracy–Widom curve $F^{g,0}(S)$ (let alone $F^{g,k}(S)$) for fixed $g$. A related challenge would be to have a closed formula, at $g$ fixed, for $F^{g,0}$ and $\omega_n^{g,0}$ of the Tracy–Widom curve.
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Acknowledgments

GB and BE would like to thank M Bergère, F David and G Schehr for useful and fruitful discussions, as well as the organizers of StatPhys2010 in Brisbane and Cairns. GB would like to thank the scientific hospitality of the SISSA and of the Department of Maths and Statistics of Melbourne University, where part of this work was completed. He also would like to thank in general the organizers and participants of the Random Matrix Theory semester at the MSRI, and in particular J Baik and P Deift for their corrections. He acknowledges the funding of the CEA and CFM for travel to StatPhys conferences. The work of BE is partly supported by the ANR project GranMa ‘Grandes Matrices Aléatoires’ ANR-08-BLAN-0311-01, by the European Science Foundation through the Misgam program and by the Quebec government with the FQRNT. He also would like to thank the CRM (Centre de recherche mathématiques de Montréal, QC, Canada) and CERN for its hospitality.

Appendix A. Determination of the normalization $C_{N,\beta}$

When $a \to -\infty$, we have the asymptotics:

\[ F^{0,0}(a) = -\frac{t|a|^2}{2} - t^2 \ln |a| - \frac{3t^2}{2} + t^2 \ln t + O(|a|^{-2}) \]
\[ F^{0,1}(a) = t \ln |a| + t - t \ln t - t \ln(2\pi) + O(|a|^{-2}) \]
\[ F^{1,0}(a) = \ln \frac{2}{6} + O(|a|^{-4}) \]
\[ F^{0,2}(a) = \frac{\ln 2}{3} + O(|a|^{-2}) \]

and one can prove that all other $F^{g,k}$ are $O(|a|^{-2})$. As $F(a) = \sum_{g,k \geq 0} \nu^{2 - 2g} \eta^k F^{g,k}(a)$, we have the following result:

\[ \lim_{a \to -\infty} \left[ F(a) - (\beta N^2 \{ -|a|^2/(2t) - \ln |a| \} + (\beta - 1)N \ln |a|) \right] = \beta N^2 \left( -\frac{3}{2} + \ln t \right) + (\beta - 1)N (1 - \ln(2\pi) - \ln t) \]
\[ + \left( \frac{\beta + \beta^{-1} - 5}{3} \right) \ln 2. \] \hspace{1cm} (A.1)

On the other hand, consider the bounded eigenvalue integral:

\[ I_{N,\beta}(a) = \int_{|\lambda - a|} d\lambda_1 \cdots d\lambda_N |\Delta(\lambda)|^{2\beta} e^{-N\beta/2t} \sum_{i=1}^N \lambda_i^2. \] \hspace{1cm} (A.2)

One knows that

\[ I_{N,\beta}(a) = C_{N,\beta} e^{F(a)}. \] \hspace{1cm} (A.3)

We want to find $C_{N,\beta}$ by comparing the result of equation (A.1) to the large $N$ asymptotic of $I_{N,\beta}(a)$. Heuristically, when $a \to -\infty$, we do not see a Gaussian weight, but the tail of it starting from the point $a$, which looks like a decreasing exponential. Precisely, for any finite $N$, with the change of variable:

\[ \mu_i = \frac{\beta N|a|}{t}(a - \lambda_i) \] \hspace{1cm} (A.4)

doi:10.1088/1742-5468/2011/11/P11024
we find
\[ I_{N,\beta}(a) = \left( \frac{t}{N\beta|a|} \right)^{\beta N(N-1)+N} e^{-\beta N^2|a|^2/2t} \times \int_{\mathbb{R}_+^N} d\mu_1 \cdots d\mu_N |\Delta(\mu)|^{2\beta} e^{-\sum_{i=1}^{N} \mu_i} e^{-\left(\langle N/2 \beta|a|^2 \right)} \sum_{i=1}^{N} \mu_i^2. \] (A.5)

When \( a \to -\infty \), the integral is of order 1 and converges to a Selberg integral with Laguerre-type weight:
\[ \int_{\mathbb{R}_+^N} d\mu_1 \cdots d\mu_N |\Delta(\mu)|^{2\beta} e^{-\sum_{i=1}^{N} \mu_i} = \frac{\left( \prod_{j=1}^{N} \Gamma(1+j\beta) \right)^2}{\Gamma(1+\beta)^N \Gamma(1+N\beta)}. \] (A.6)

Thus, we have
\[ \lim_{a \to -\infty} \left[ \ln(I_{N,\beta}(a)) - \left( \frac{\beta N^2}{2t} - \ln |a| \right) \right] = \beta N^2 \left( -\frac{3}{2} + \ln t \right) + (\beta - 1)N \left( 1 - \ln(2\pi) - \ln t \right) - \beta N^2 \ln N + \beta N^2 \left( \frac{3}{2} - \ln \beta \right) + (\beta - 1)N \ln N + (\beta - 1)N(-1 + \ln(2\pi) + \ln \beta) - \ln \Gamma(1+N\beta) + \sum_{j=1}^{N} 2 \ln \Gamma(1+j\beta). \] (A.7)

Hence
\[ C_{N,\beta} = \exp\left[-\beta N^2 \ln N + \beta N^2 \left( \frac{3}{2} - \ln \beta \right) + (\beta - 1)N \ln N + (\beta - 1)N(-1 + \ln(2\pi) + \ln \beta) \right]^{2^{5/6-(\beta+\beta^{-1})/3}} \frac{\left( \prod_{j=1}^{N} \Gamma(1+j\beta) \right)^2}{\Gamma(1+\beta)^N \Gamma(1+N\beta)}. \] (A.8)

Note that this normalization constant is the correct one when there is one hard edge. The result is different when there is no hard edge.

**Appendix B. Reminder on special functions**

**B.1. Gamma function**

We recall the definition of the Gamma function. For \( \text{Re} s > 1 \):
\[ \Gamma(z) = \int_{0}^{\infty} d\sigma \sigma^{s-1} e^{-\sigma}. \] (B.1)

It satisfies the functional equation:
\[ \Gamma(z+1) = z\Gamma(z). \] (B.2)

It can be analytically continued on the complex plane, and \( 1/\Gamma(z) \) is an entire function, which can be defined by the product
\[ \frac{1}{\Gamma(z)} = z e^{-\gamma z} \prod_{m=1}^{\infty} \left( 1 + \frac{z}{m} \right) e^{-z/m}. \] (B.3)
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Its asymptotic expansion is given by the Stirling formula:
\[
\ln \Gamma(z + 1) \xrightarrow{z \to \infty} z \ln z - z + \frac{\ln z}{2} + \frac{\ln(2\pi)}{2} + o(1).
\] (B.4)

It satisfies the reflection property:
\[
\Gamma(1 - z)\Gamma(z) = \frac{\pi}{\sin(\pi z)}
\] (B.5)

and the ‘addition of angles’ formula:
\[
\prod_{m=0}^{n-1} \Gamma \left( \frac{m}{p} + z \right) = (2\pi)^{(p-1)/2} p^{-pz+1/2} \Gamma(pz).
\] (B.6)

### B.2. Riemann zeta function

We recall the definition of the zeta function. For \( \text{Re } s > 1 \):
\[
\zeta(s) \equiv \sum_{j=1}^{\infty} \frac{1}{j^s}.
\] (B.7)

It can be extended by analytical continuation on the complex plane, by the reflection relation:
\[
\Gamma \left( \frac{s}{2} \right) \pi^{-s/2} \zeta(s) = \Gamma \left( \frac{1-s}{2} \right) \pi^{(1-s)/2} \zeta(1-s).
\] (B.8)

### B.3. Barnes \( G \) function

We recall that the Barnes \( G \) function [4] is an entire function satisfying
\[
G(z + 1) = \Gamma(z)G(z).
\] (B.9)

At integers, it coincides with the product of factorials:
\[
\forall k \in \mathbb{N} \quad G(k) = \prod_{j=1}^{k-2} j!
\] (B.10)

We take the convention that a void product is equal to 1. Its asymptotic expansion for \( z \to \infty \) is a classical result [26]:
\[
\ln G(z + 1) = z^2 \left( \frac{\ln z}{2} - \frac{3}{4} \right) + \frac{\ln(2\pi)z}{2} - \frac{\ln z}{12} + \zeta'(1) + o(1).
\] (B.11)

We quote some special values and relations [1]:
\[
\ln G(1/2) = \frac{7 \ln 2}{24} - \frac{\ln(2\pi)}{4} + \frac{3\zeta'(-1)}{2}
\] (B.12)
\[
\ln G(1/4) + \ln G(3/4) = \frac{\ln 2}{8} - \frac{\ln(2\pi)}{4} - \frac{\ln \Gamma(1/4)}{2} + \frac{9\zeta'(-1)}{4}
\] (B.13)
\[
\ln G(1/3) + \ln G(2/3) = \frac{7 \ln 3}{36} - \frac{\ln(2\pi)}{3} - \frac{\ln \Gamma(1/3)}{3} + \frac{8\zeta'(-1)}{3}
\] (B.14)
\[
\ln G(1/6) + \ln G(5/6) = -\frac{5 \ln 2}{36} - \frac{25 \ln 3}{72} + \frac{\ln(2\pi)}{6} - \frac{4\ln \Gamma(1/3)}{3} + \frac{5\zeta'(-1)}{3}.
\] (B.15)
In principle, \( G(p/q) \) can be expressed in terms of an increasing number of fundamental constants when \( p \) and \( q \) increase. We do not try in this paper to express \( G(p/q) \) in other ways. We mention an ‘addition of angles’ formula \([16]\):

\[
\sum_{m=0}^{p-1} \ln G \left( \frac{m}{p} + z + 1 \right) = \left( p - \frac{1}{p} \right) \zeta'(1) + \frac{\ln G(pz+1)}{p} - z \ln \Gamma(pz) \\
+ \left( pz^2 - z + \frac{1}{6p} \right) \ln \frac{p}{2} + \sum_{m=0}^{p-1} \left( \frac{m}{p} + z \right) \ln \Gamma \left( \frac{m}{p} + z \right). \tag{B.16}
\]

**Appendix C. Large \( N \) expansion of some Selberg integrals**

We have defined

\[
I_{N,\beta} \equiv I_{N,\beta}(\infty) = \int_{\mathbb{R}^N} d\lambda_1 \cdots d\lambda_N |\Delta(\lambda)|^{2\beta} e^{-(N\beta/2) \sum_{i=1}^N \lambda_i^2}. \tag{C.1}
\]

This is a Selberg integral and it is given by

\[
I_{N,\beta} = (2\pi)^{N/2} \left( \frac{t}{N\beta} \right)^{(\beta N^2/2) - ((\beta-1)N/2)} \prod_{j=1}^N \frac{\Gamma(1+j\beta)}{\Gamma(1+\beta)^N}. \tag{C.2}
\]

In the main text, we are interested in

\[
P(a) = \frac{I_{N,\beta}(a)}{I_{N,\beta}(\infty)} \approx C_{N,\beta}^F(a), \tag{C.3}
\]

up to a factor equivalent to 1. So, we need the large \( N \) asymptotics of \( \ln I_{N,\beta} \) and \( \ln C_{N,\beta} \) up to \( o(1) \).

**C.1. Asymptotics of \( \sum_{j=1}^N \ln \Gamma(1+j\beta) \)**

We first determine the large \( N \) asymptotics of \( \sum_{j=1}^N \ln \Gamma(1+j\beta) \) for a general \( \beta \). We start with an integral representation of \( \ln \Gamma \):

\[
\ln \Gamma(z) = \left( z - \frac{1}{2} \right) \ln z - z + \frac{\ln(2\pi)}{2} + \int_0^{\infty} d\sigma \frac{e^{-\sigma z}}{\sigma^2} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} \right). \tag{C.4}
\]

We have

\[
\sum_{j=1}^N \ln \Gamma(1+j\beta) = \sum_{j=1}^N \ln(j+\beta) + \ln \Gamma(\beta) \\
= \sum_{j=1}^N \ln j + N \ln(\beta) + \sum_{j=1}^N \left( j\beta - \frac{1}{2} \right) \ln(j\beta) - \beta \frac{N(N+1)}{2} + N \frac{\ln(2\pi)}{2} \\
+ \sum_{j=1}^K \int_0^{\infty} d\sigma \frac{e^{-j\beta \sigma}}{\sigma^2} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} \right).
\]

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doi:10.1088/1742-5468/2011/11/P11024

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50
where
\[ \gamma \]
or equivalently
\[ \kappa \]
\[ \kappa \equiv N \ln(\beta) + \ln(2\pi) + K. \]

We recall the classical asymptotic expansions:
\[ \sum_{j=1}^{N} \ln j = \ln \Gamma(N+1) = N \ln N - N + \frac{\ln N}{2} + \frac{\ln(2\pi)}{2} + o(1) \]
\[ \sum_{j=1}^{N} j \ln j = \frac{N^2 \ln N}{2} - \frac{N^2}{4} + \frac{N \ln N}{2} + \frac{N}{12} + \frac{1}{12} - \zeta'(-1) + o(1). \]

Let us determine the large \( N \) asymptotic of the integral:
\[
K \equiv \sum_{j=1}^{N} \int_{0}^{\infty} d\sigma \frac{e^{-j\beta\sigma}}{\sigma^2} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} \right) = \frac{1}{12} \sum_{j=1}^{N} \int_{0}^{\infty} d\sigma e^{-j\beta\sigma} + \sum_{j=1}^{N} \int_{0}^{\infty} d\sigma \frac{e^{-j\beta\sigma}}{\sigma^2} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} - \frac{\sigma^2}{12} \right)
\]
\[
= \frac{1}{12} \sum_{j=1}^{N} \frac{1}{j} \left( \frac{1}{e^{\sigma\beta} - 1} \right) + \int_{0}^{\infty} d\sigma \frac{1}{\sigma^2} \frac{1}{e^{\beta\sigma} - 1} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} - \frac{\sigma^2}{12} \right)
\]
\[
= \frac{\ln N}{12\beta} + \frac{\gamma_E}{12\beta} + \int_{0}^{\infty} d\sigma \frac{1}{\sigma^2} \frac{1}{e^{\beta\sigma} - 1} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} - \frac{\sigma^2}{12} \right) \tag{C.7}
\]
where \( \gamma_E \) is the Euler–Mascheroni constant. Collecting the results, we find
\[
\sum_{j=1}^{N} \ln \Gamma(1 + j\beta) = \frac{\beta N^2 \ln N}{2} + \beta \left( -\frac{3}{4} + \frac{\ln(\beta)}{2} \right) N^2 + \frac{\beta + 1}{2} N \ln N
\]
\[
+ \left( \frac{1}{2} + \frac{\beta \ln(\beta)}{2} - \frac{\beta}{2} + \frac{\ln(\beta)}{2} + \frac{\ln(2\pi)}{2} \right) N
\]
\[
+ \frac{\beta + \beta^{-1} + 3}{12} \ln N + \kappa_{\beta} + o(1) \tag{C.8}
\]
where the finite part \( \kappa_{\beta} \) is given by
\[
\kappa_{\beta} = \frac{\ln(2\pi)}{4} + \beta \left( \frac{1}{12} - \zeta'(-1) \right) + \frac{\gamma_E}{12\beta} + \int_{0}^{\infty} d\sigma \frac{1}{\sigma^2} \frac{1}{e^{\beta\sigma} - 1} \left( \frac{\sigma}{e^\sigma - 1} - 1 + \frac{\sigma}{2} - \frac{\sigma^2}{12} \right) \tag{C.9}
\]
or equivalently
\[
\kappa_{\beta} = \frac{\ln(2\pi)}{4} + \beta \left( \frac{1}{12} - \zeta'(-1) \right) + \frac{\gamma_E}{12\beta} + \int_{0}^{\infty} d\sigma \left[ \frac{6\sigma \coth(\sigma/2) - 12 - \sigma^2}{12\sigma^2 (e^{\beta\sigma} - 1)} \right]. \tag{C.10}
\]
C.2. Asymptotics of $I_{N,\beta}$

If we apply equation (C.8) to the definition of $I_{N,\beta}$ (equation (C.2)), we find

$$\ln I_{N,\beta} = \beta \left( -\frac{3}{4} + \frac{\ln t}{2} \right) N^2 + \beta N \ln N$$

$$+ \left( \beta \ln(\beta) - \ln \Gamma(1 + \beta) - \frac{\beta + 1}{2} + \ln(2\pi) - \frac{\beta - 1}{2} \ln t \right) N$$

$$+ \frac{\beta + \beta^{-1} + 3}{12} \ln N + \kappa_{\beta} + o(1).$$

(C.11)

C.3. Asymptotic of $C_{N,\beta}$

If we apply equation (C.8) to the expression of $C_{N,\beta}$ (equation (A.8)), we find

$$\ln C_{N,\beta} = \beta N \ln N + N(-\beta + \beta \ln \beta + \beta \ln(2\pi) - \ln \Gamma(1 + \beta)) + \left( \frac{\beta + \beta^{-1}}{6} \right) \ln(N)$$

$$+ \left( \frac{5 - 2(\beta + \beta^{-1})}{6} \right) \ln 2 - \frac{\ln(2\pi)}{2} - \frac{\ln \beta}{2} + 2\kappa_{\beta} + o(1).$$

(C.12)

Finally we get the normalization constant for the cumulative distribution of $\lambda_{\text{max}}$. We have indeed

$$P(a) = \frac{I_{N,\beta}(a)}{I_{N,\beta}(\infty)} = \frac{C_{N,\beta}}{I_{N,\beta}} e^{F(a)} = \exp \{ F(a) + \ln C_{N,\beta} - \ln I_{N,\beta} \}$$

(C.13)

with

$$\ln C_{N,\beta} - \ln I_{N,\beta} = \beta N^2 \left( \frac{3}{4} - \frac{\ln t}{2} \right) + N(\beta - 1) \left( -\frac{1}{2} + \ln(2\pi) + \frac{\ln t}{2} \right)$$

$$+ \ln N \left( \frac{\beta + \beta^{-1} - 3}{12} \right) + \left( \frac{5 - 2(\beta + \beta^{-1})}{6} \right) \ln 2$$

$$- \frac{\ln(2\pi)}{2} - \frac{\ln \beta}{2} + \kappa_{\beta}.$$ 

(C.14)

C.4. Another computation of $\kappa_{\beta}$ for $\beta$ rational

C.4.1. $\beta$ integer. By definition we have $\kappa_{\beta} = [\sum_{j=1}^{N} \ln \Gamma(1 + j\beta)]_{\text{finite}}$. Let us come back to the asymptotic of $\sum_{j=1}^{N} \ln \Gamma(1 + j\beta)$ when $\beta$ is a positive integer. Here, we only focus on the finite term of this expansion, which we called previously $\kappa_{\beta}$. Of course, one could check with this method that one does obtain the divergent term of equation (C.8).

For $\beta = 1$, this series is the Barnes function itself:

$$\sum_{j=1}^{N} \ln \Gamma(1 + j) = \ln G(N + 2).$$

(C.15)

Using the asymptotic expansion of $G(z + 1)$ given in equation (B.11), we obtain for any $w$ in the large $z$ limit:

$$[\ln G(z + w + 1)]_{\text{finite}} = \frac{w \ln(2\pi)}{2} + \zeta'(-1) + o(1).$$

(C.16)
Thus, we have
\[ \kappa_1 = \frac{\ln(2\pi)}{2} + \zeta'(-1). \] (C.17)

As a matter of fact, one may use only the asymptotic expansion of the Barnes function (equation (B.11)) to work out the case $\beta$ integer. We start from the ‘addition of angles’ formula (equation (B.6)) and write
\[ \prod_{j=1}^{N} \Gamma(j\beta) = (2\pi)^{(1-\beta)N/2} \beta^{\beta N(N+1)/2-(N/2)} \prod_{m=0}^{\beta-1} \frac{G(N+m/\beta+1)}{G(m/\beta+1)}. \] (C.18)
Then
\[ \sum_{j=1}^{N} \ln \Gamma(1+j\beta) = \ln(N!) + N \ln \beta + N \left( \frac{1-\beta}{2} \right) \ln(2\pi) + \left( \frac{\beta N(N+1)}{2} - \frac{N}{2} \right) \ln \beta 
+ \sum_{m=0}^{\beta-1} \left[ \ln G \left( N + \frac{m}{\beta} + 1 \right) - \ln G \left( \frac{m}{\beta} + 1 \right) \right]. \] (C.19)
Subsequently
\[ \kappa_\beta = \frac{\ln(2\pi)}{2} + \sum_{m=0}^{\beta-1} \left[ \frac{m \ln(2\pi)}{2\beta} + \zeta'(-1) - \ln G \left( \frac{m}{\beta} + 1 \right) \right] 
= \left( \frac{\beta + 1}{4} \right) \ln(2\pi) + \beta \zeta'(-1) - \sum_{m=0}^{\beta-1} \ln G \left( 1 + \frac{m}{\beta} \right). \] (C.20)
For example
\[ \kappa_2 = \frac{3 \ln(2\pi)}{4} + 2 \zeta'(-1) - \ln G(3/2) 
= \frac{5 \ln 2}{24} + \frac{\ln(2\pi)}{2} + \frac{\zeta'(-1)}{2} \] (C.21)
where we used the value of $G(1/2)$ given in equation (B.12) and $\Gamma(1/2) = \sqrt{\pi}$.

Actually, $\kappa_\beta$ can be expressed only in terms of values of $\Gamma$ at rational points, by means of the formula equation (B.16):
\[ \sum_{m=0}^{p-1} \ln G \left( 1 + \frac{m}{p} \right) = \left( p - \frac{1}{p} \right) \zeta'(-1) + \frac{\ln p}{12p} + \sum_{m=1}^{p-1} \frac{m}{p} \ln \Gamma \left( \frac{m}{p} \right). \] (C.22)
Thus, we have
\[ \kappa_\beta = \left( \frac{\beta + 1}{4} \right) \ln(2\pi) + \frac{\zeta'(-1)}{\beta} - \frac{\ln \beta}{12\beta} - \sum_{m=1}^{\beta-1} \frac{m}{\beta} \ln \Gamma \left( \frac{m}{\beta} \right). \] (C.23)
For example
\[ \kappa_3 = \frac{11 \ln 3}{36} + \frac{\ln(2\pi)}{3} + \frac{\zeta'(-1)}{3} + \frac{\ln(1/3)}{3} \] (C.24)
\[ \kappa_4 = \frac{7 \ln 2}{12} + \frac{\ln(2\pi)}{4} + \frac{\zeta'(-1)}{4} + \frac{\ln(1/4)}{2}. \] (C.25)
C.4.2. Value of $\kappa_{\beta}$ for rational $\beta$. The case $\beta = p/q \in \mathbb{Q}$ can be treated similarly (we assume $\gcd(p,q) = 1$). It is convenient to do the computation for a given congruence of $N \mod q$. Let us write

$$N = qL + R, \quad 0 \leq R \leq q - 1.$$

We have according to the duplication formula:

$$\prod_{j=1}^{N} \Gamma\left(\frac{j^2}{q}\right) = (2\pi)^{(1-p)N/2} p^p (p^{N(N+1)/2q})^{-\frac{(N/2)}{2}} \prod_{m=0}^{p-1} \prod_{j=1}^{N} \Gamma\left(\frac{j + m}{p}\right).$$

Then, we obtain

$$\prod_{r=1}^{q} G(L + r/q + m/p + 1) \prod_{r=R+1}^{q} G(L + r/q + m/p) = \frac{\prod_{r=1}^{q} G(r/q + m/p)}{\prod_{r=1}^{q} G(r/q + m/p)}.$$

Then

$$\prod_{j=1}^{N} \Gamma\left(1 + \frac{j^2}{q}\right) = N! \left(\frac{p}{q}\right)^N \prod_{j=1}^{N} \Gamma\left(\frac{j^2}{q}\right).$$

We need the large $z$ asymptotic expansion of $\ln G(1 + w + z/q)$ for fixed $w$. Starting from equation (B.11), we obtain

$$\ln G\left(1 + w + \frac{z}{q}\right) = \left(\frac{1}{12} - \frac{w^2}{2}\right) \ln q + \frac{w \ln(2\pi)}{2} + \zeta(-1).$$

Then, we have

$$\kappa_{p/q} = \frac{\ln(2\pi)}{2} - \sum_{m=0}^{p-1} \sum_{r=1}^{q} \ln G\left(\frac{r}{q} + \frac{m}{p}\right) + pq\zeta'(-1)$$

$$+ \left\{ \sum_{r=1}^{R} \left[ \frac{1}{12} - \frac{1}{2} \left( -\frac{r}{q} + \frac{m}{p}\right)^2 \right] \right\} \ln q$$

$$+ \sum_{r=R+1}^{q} \left[ \frac{1}{12} - \frac{1}{2} \left( -\frac{r}{q} + \frac{m}{p}\right)^2 \right] \ln q$$

$$+ \sum_{m=0}^{p-1} \left( \sum_{r=1}^{R} \left( -\frac{r}{q} + \frac{m}{p}\right) + \sum_{r=R+1}^{q} \left( -\frac{r}{q} + \frac{m}{p}\right) \right) \frac{\ln(2\pi)}{2}$$

$$= \frac{3 - p/q - q/p}{12} \ln q + \frac{p - q + 2}{4} \ln(2\pi) + pq\zeta'(-1)$$

$$- \sum_{m=0}^{p-1} \sum_{r=1}^{q} \ln G\left(\frac{r}{q} + \frac{m}{p}\right).$$

Notice that the result does not depend on the congruence of $N$ modulo $q$. For $q = 1$, we recover the result for $\beta = p$ integer (see (C.20)).

We give below some values of $\kappa_{\beta}$. To complete the table, we have used repeatedly the reflection formula and the ‘duplication of angles’ formula of the $\Gamma$ function and equations (B.12)–(B.15).
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Value of $\beta \kappa$ $
\begin{array}{ll}
1 & \frac{\ln(2\pi)}{2} + \zeta'(-1) \\
2 & \frac{5 \ln 2}{24} + \frac{\ln(2\pi)}{2} + \zeta'(-1) \\
3 & \frac{11 \ln 3}{36} + \frac{\ln(2\pi)}{3} + \frac{2}{3} \zeta'(-1) + \frac{\ln \Gamma(1/3)}{3} \\
4 & \frac{7 \ln 2}{12} + \frac{\ln(2\pi)}{4} + \frac{\zeta'(-1)}{4} + \frac{\ln \Gamma(1/4)}{2} \\
1/2 & -\frac{\ln 2}{4} + \frac{\ln(2\pi)}{2} + \frac{\zeta'(-1)}{4} \\
1/3 & -\frac{2 \ln 3}{9} + \frac{\ln(2\pi)}{2} + \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{3} \\
1/4 & -\frac{5 \ln 2}{8} + \frac{\ln(2\pi)}{4} + \frac{\zeta'(-1)}{4} + \frac{\ln \Gamma(1/4)}{2} \\
2/3 & -\frac{13 \ln 2}{72} + \frac{5 \ln 3}{18} + \frac{2 \ln(2\pi)}{4} - \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{6} \\
3/2 & -\frac{\ln 2}{4} + \frac{11 \ln 3}{72} + \frac{2 \ln(2\pi)}{3} - \frac{\ln \Gamma(1/3)}{3} + \frac{\zeta'(-1)}{6} \\
\end{array}$

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