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# A simple derivation of the Tracy–Widom distribution of the maximal eigenvalue of a Gaussian unitary random matrix

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**Abstract.** In this paper, we first briefly review some recent results on the distribution of the maximal eigenvalue of an  $(N \times N)$  random matrix drawn from Gaussian ensembles. Next we focus on the Gaussian unitary ensemble (GUE) and by suitably adapting a method of orthogonal polynomials developed by Gross and Matytsin in the context of Yang–Mills theory in two dimensions, we provide a rather simple derivation of the Tracy–Widom law for GUE. Our derivation is based on the elementary asymptotic scaling analysis of a pair of coupled nonlinear recursion relations. As an added bonus, this method also allows us to compute the precise subleading terms describing the right large deviation tail of the maximal eigenvalue distribution. In the Yang–Mills language, these subleading terms correspond to non-perturbative (in  $1/N$  expansion) corrections to the two-dimensional partition function in the so called ‘weak’ coupling regime.

**Keywords:** Painlevé equations, random matrix theory and extensions

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**1. Introduction**

Quite a long time ago, Wigner [1] introduced random matrices in the context of nuclear physics. He suggested that the highly excited energy levels of complex nuclei can locally be well represented by the eigenvalues of a large random matrix. A big nucleus is a rather complex system composed of many strongly interacting quantum particles and it is practically impossible to describe its spectral properties via first principle calculations. The idea of Wigner was to model the spectral properties of the complex Hamiltonian of such a big nucleus by those of a large random matrix preserving the same symmetry. This was a very successful approach in nuclear physics. Since then, the random matrix theory (RMT) has gone beyond nuclear physics and has found a wide number of applications in various fields of physics and mathematics including quantum chaos, disordered systems, string theory and even number theory [2]. A case of special interest is the one of Gaussian random matrices (originally introduced by Wigner himself) where the entries of the matrix are Gaussian random variables.

Depending on the symmetry of the problem, Dyson distinguished three classes for the matrix  $X$  [3]:

- the Gaussian orthogonal ensemble (GOE):  $X$  is real symmetric.
- the Gaussian unitary ensemble (GUE):  $X$  is complex Hermitian.
- the Gaussian symplectic ensemble (GSE):  $X$  is quaternionic Hermitian.

Let us write  $X^\dagger$  the adjoint of  $X$ , i.e. the transpose of  $X$  for the GOE, the complex conjugate transpose for the GUE and the quaternionic conjugate transpose for the GSE.

A Gaussian random matrix is an  $N \times N$  self-adjoint matrix  $X$ , i.e.  $X^\dagger = X$  distributed according to the law

$$\mathcal{P}(X) \propto e^{-(\beta/2) \text{Tr}(X^2)} \quad \text{with } \beta = \begin{cases} 1 & \text{for GOE} \\ 2 & \text{for GUE} \\ 4 & \text{for GSE} \end{cases} \quad (1)$$

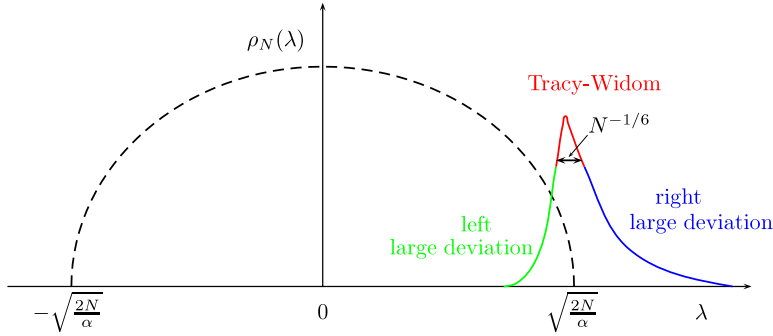
where, for convenience, we have chosen the prefactor  $\beta$  of the  $\text{Tr}(X^2)$  to be  $\beta = 1$  for the GOE,  $\beta = 2$  for the GUE and  $\beta = 4$  for the GSE. For instance, for the GUE we have  $\beta = 2$  and  $\mathcal{P}(X) \propto e^{-\text{Tr}(X^2)} \propto e^{-\sum_{i,j} |X_{i,j}|^2}$  as  $X^2 = X^\dagger X = \sum_{i,j} |X_{i,j}|^2$ . This means that  $X$  is an  $N \times N$  complex Hermitian matrix with entries  $\text{Re } X_{i,j}$  and  $\text{Im } X_{i,j}$  for  $i < j$  that are independent (real) random variables distributed according to the same centred Gaussian law with variance  $1/4$  and the  $X_{i,i}$  are (real) independent Gaussian variables with mean 0 and variance  $1/2$ . In the case of GSE, there are  $2N$  eigenvalues, each of them two-fold degenerate and  $\text{Tr}$  in (1) for  $\beta = 4$  is defined so that only one of the two-fold degenerate eigenvalues in  $X$  is counted.

Self-adjoint matrices can be diagonalized and have real eigenvalues. The joint distribution of eigenvalues of the Gaussian ensemble is well known [4, 2]

$$\mathcal{P}(\lambda_1, \dots, \lambda_N) = B_N e^{-(\beta/2) \sum_{i=1}^N \lambda_i^2} \prod_{j < k} |\lambda_j - \lambda_k|^\beta \quad (2)$$

where  $B_N$  is a normalization constant such that  $\int (\prod_i d\lambda_i) \mathcal{P}(\lambda_1, \dots, \lambda_N) = 1$  (it depends on  $\beta$ ) and the power  $\beta$  of the Vandermonde term is called the Dyson index  $\beta = 1, 2$  or 4 depending on the ensemble (resp. GOE, GUE or GSE). Note that we have chosen the prefactor of the  $\text{Tr}(X^2)$  term in (1) to be the same as the Dyson index  $\beta$  just for convenience. This prefactor is not very important as it can be absorbed by rescaling the matrix entries by a constant factor. In contrast, the value of the Dyson index  $\beta = 1, 2$  or 4, characterizing the power of the Vandermonde term, plays a crucial role. The normalization constant  $B_N$  can be computed using Selberg's integral [2]:  $B_N = \beta^{(N/2)+\beta(N(N-1)/4)} (2\pi)^{-N/2} \Gamma(1 + \beta/2)^N / [\prod_{j=1}^N \Gamma(1 + j\beta/2)]$ .

Because of the presence of the Vandermonde determinant  $\prod_{j < k} (\lambda_j - \lambda_k)$  in equation (2), the eigenvalues are strongly correlated random variables, they repel each other. In this paper, our focus is on the statistical properties of the extreme (maximal) eigenvalue  $\lambda_{\max} = \max(\lambda_1, \lambda_2, \dots, \lambda_N)$ . Had the Vandermonde term been not there in the joint distribution (2), the joint distribution would factorize and the eigenvalues would thus be completely independent random variables, each with a Gaussian distribution. For such independent and identically distributed random variables  $\{\lambda_i\}$ , the extreme value statistics is well understood [5] and the distribution of the maximum, properly shifted and scaled, belongs to one of the three universality classes Gumbel, Fréchet or Weibull (for large  $N$ ) depending on the tail of the distribution of individual  $\lambda_i$ s. However, in the case of random matrix theory, the eigenvalues  $\lambda_i$ s are strongly correlated variables. For strongly correlated random variables there is no general theory for the distribution of the maximum. In the case of Gaussian random matrices, where the joint distribution (2) is explicitly known, much progress has been made in understanding the distribution of  $\lambda_{\max}$  following the seminal work by Tracy and Widom [6, 7]. This then provides a very useful solvable model for the extreme value distribution in a strongly correlated system and hence is of special interest.



**Figure 1.** Average density of the eigenvalues of a Gaussian random matrix  $\rho_N(\lambda)$  as a function of  $\lambda$  (black dashed line). The density has a semi-circular shape (‘Wigner semi-circle’) and a finite support  $[-\sqrt{2N/\alpha}, \sqrt{2N/\alpha}]$ . The maximal eigenvalue has mean value  $\langle \lambda_{\max} \rangle \approx \sqrt{2N/\alpha}$  for large  $N$  and its distribution close to the mean value, over a scale of  $O(N^{-1/6})$  has the Tracy–Widom form (red solid line). However, over a scale  $(\sqrt{N})$  the distribution has large deviation tails shown by solid green (left large deviations) and solid blue (right large deviations) lines.

Let us first summarize some known properties of the random variable  $\lambda_{\max}$ . Its average value can be easily obtained from the right edge of the well-known Wigner semi-circle describing the average density of eigenvalues. For a Gaussian random matrix of large size  $N$ , the average density of eigenvalues (normalized to unity)  $\rho_N(\lambda) = \langle (1/N) \sum_i \delta(\lambda - \lambda_i) \rangle$  has a semi-circular shape on a finite support  $[-\sqrt{2N}, \sqrt{2N}]$  called the Wigner semi-circle [1]:

$$\rho_N(\lambda) \approx \frac{1}{\sqrt{N}} g\left(\frac{\lambda}{\sqrt{N}}\right) \quad \text{with } g(x) = \frac{1}{\pi} \sqrt{2 - x^2} \quad \text{for large } N. \quad (3)$$

The quantity  $\rho_N(\lambda)d\lambda$  represents the average fraction of eigenvalues that lie within the small interval  $[\lambda, \lambda+d\lambda]$ . Therefore, equation (3) means that the eigenvalues of a Gaussian random matrix lie on average within the finite interval  $[-\sqrt{2N}, \sqrt{2N}]$ . Note also that one can rewrite, using the joint distribution in (2)

$$\rho_N(\lambda) = \left\langle \frac{1}{N} \sum_i \delta(\lambda - \lambda_i) \right\rangle = \int \mathcal{P}(\lambda, \lambda_2, \dots, \lambda_N) d\lambda_2 \cdots d\lambda_N. \quad (4)$$

Hence the average density of states  $\rho_N(\lambda)$  can also be interpreted as the marginal distribution of one of the eigenvalues (say the first one). Thus, the marginal distribution also has the shape of a semi-circle. Figure 1 shows the average density  $\rho_N(\lambda)$  ( $\alpha = 1$  here).

It then follows that the average value of the maximal eigenvalue  $\lambda_{\max}$  is given for large  $N$  by the upper bound of the density support:

$$\langle \lambda_{\max} \rangle \approx \sqrt{2N} \quad \text{for large } N. \quad (5)$$

However,  $\lambda_{\max}$  fluctuates around this average value from one realization to another and has a distribution around its mean value  $\sqrt{2N}$  (see figure 1 with  $\alpha = 1$ ). What is the full

probability distribution of  $\lambda_{\max}$ ? From the joint distribution of eigenvalues in equation (2), it is easy to write down formally the cumulative distribution function (cdf) of  $\lambda_{\max}$  as a multiple integral

$$\mathbb{P}_N(\lambda_{\max} \leq t) = B_N \prod_{i=1}^N \int_{-\infty}^t d\lambda_i \prod_{j < k} |\lambda_j - \lambda_k|^\beta e^{-(\beta/2) \sum_{i=1}^N \lambda_i^2} \quad (6)$$

which can be interpreted as a partition function of a Coulomb gas in the presence of a hard wall at the location  $t$  (see the discussion in section 2). The question is how does  $\mathbb{P}_N(\lambda_{\max} \leq t)$  behave for large  $N$ ? It turns out that the fluctuations of  $\lambda_{\max}$  around its mean  $\sqrt{2N}$  have two scales for large  $N$ . While typical fluctuations scale as  $N^{-1/6}$ , large fluctuations scale as  $N^{1/2}$  and their probability distributions are described by different functional forms (see figure 1 with  $\alpha = 1$ ).

*Typical fluctuations.* From an asymptotic analysis of the multiple integral in equation (6), Forrester [8], followed Tracy and Widom [6, 7] deduced that for large  $N$ , small and typical fluctuations of the maximal eigenvalue around its mean value  $\sqrt{2N}$  are of order  $O(N^{-1/6})$  and can be written as

$$\lambda_{\max} \approx \sqrt{2N} + a_\beta N^{-1/6} \chi \quad (7)$$

where  $a_{1,2} = 1/\sqrt{2}$  (for GOE and GUE) and  $a_4 = 2^{-7/6}$  (GSE) and  $\chi$  is a random variable characterizing the typical fluctuations. Tracy and Widom [6, 7] proved that for large  $N$ , the distribution of  $\chi$  is independent of  $N$ :  $P(\chi \leq x) = F_\beta(x)$ . The function  $F_\beta(x)$  depends explicitly on  $\beta$  and is called the Tracy–Widom distribution. For example, for  $\beta = 2$  [6, 7],

$$F_2(x) = \exp \left[ - \int_x^\infty (z - x) q^2(z) dz \right] \quad (8)$$

where  $q(z)$  satisfies the special case of  $\alpha = 0$  of the Painlevé II equation

$$q''(z) = 2q^3(z) + zq(z) + \alpha. \quad (9)$$

For  $\alpha = 0$ , the solution only requires the right tail boundary condition for its unique specification:  $q(z) \sim \text{Ai}(z)$  as  $z \rightarrow \infty$ , where  $\text{Ai}(z)$  is the Airy function that satisfies the differential equation  $\text{Ai}''(z) - z\text{Ai}(z) = 0$  and vanishes as,  $\text{Ai}(z) \approx (1/2\sqrt{\pi}z^{1/4}) e^{-(2/3)z^{3/2}}$  as  $z \rightarrow \infty$ . This solution of the special case  $\alpha = 0$  of the Painlevé-II equation is called the Hastings–McLeod solution [9]. For  $\beta = 2$  and 4, one has [6, 7]

$$F_1(x) = [F_2(x)]^{1/2} \exp \left[ -\frac{1}{2} \int_x^\infty q(z) dz \right] \quad (10)$$

$$F_4(x) = [F_2(x)]^{1/2} \cosh \left[ \frac{1}{2} \int_x^\infty q(z) dz \right]. \quad (11)$$

Note that  $F_\beta(x) = \text{Prob}(\chi \leq x)$  is the cumulative probability of the scaled random variable  $\chi$  and hence it approaches 1 as  $x \rightarrow \infty$  and vanishes to 0 as  $x \rightarrow -\infty$ . The corresponding probability density function (pdf)  $F'_\beta(x) = dF_\beta(x)/dx$  vanishes as  $x \rightarrow \pm\infty$

in an asymmetric fashion. For  $\beta = 1, 2$

$$F'_\beta(x) \sim \exp\left[-\frac{\beta}{24}|x|^3\right] \quad \text{as } x \rightarrow -\infty \quad (12)$$

$$\sim \exp\left[-\frac{2\beta}{3}x^{3/2}\right] \quad \text{as } x \rightarrow \infty \quad (13)$$

$$F'_4(x) \sim \exp\left[-\frac{1}{24}|x|^3\right] \quad \text{as } x \rightarrow -\infty$$

$$\sim \exp\left[-\frac{4}{3}x^{3/2}\right] \quad \text{as } x \rightarrow \infty.$$

Over the last decade or so, the Tracy–Widom distribution has appeared in a wide variety of problems ranging from statistical physics and probability theory all the way to growth models and biological sequence matching problems (for reviews see [10]–[13]). These include the longest increasing subsequence or the Ulam problem [14, 15, 10], a wide variety of  $(1 + 1)$ -dimensional growth models [16]–[20], directed polymers in a random medium [21] and the continuum Kardar–Parisi–Zhang equation [22]–[25], the Bernoulli matching problem between two random sequences [26], and non-intersecting Brownian motions (see e.g. [27, 28] and references therein). This distribution has also been measured in a variety of recent experiments, e.g., in the height distribution of fronts generated in a paper burning experiment [29], in turbulent liquid crystals [30] and more recently in coupled fibre laser systems [31].

*Large deviations:* Tracy–Widom distribution describes the probability of typical fluctuations of  $\lambda_{\max}$  around its mean (on a scale of  $N^{-1/6}$ ), but not the atypical large fluctuations, i.e., fluctuations of order  $O(\sqrt{N})$  around the mean value  $\sqrt{2N}$ . Questions regarding such large/rare fluctuations do arise in various contexts [32]–[34] and have recently been computed [32]–[35] to dominant order for large  $N$ . As a summary, the probability density of  $\lambda_{\max}$ ,  $\mathcal{P}(\lambda_{\max} = t) = (d/dt)[\mathbb{P}_N(\lambda_{\max} \leq t)]$ , is given for large  $N$  by:

$$\begin{aligned} & \mathcal{P}(\lambda_{\max} = t) \\ & \approx \begin{cases} \exp\left\{-\beta N^2 \psi_-\left(\frac{t}{\sqrt{N}}\right) + \dots\right\} & \text{for } t < \sqrt{2N} \quad \text{and} \quad |t - \sqrt{2N}| \approx O(\sqrt{N}) \\ \frac{1}{a_\beta N^{-1/6}} F'_\beta\left(\frac{t - \sqrt{2N}}{a_\beta N^{-1/6}}\right) & \text{for } |t - \sqrt{2N}| \approx O(N^{-1/6}) \\ \exp\left\{-\beta N \psi_+\left(\frac{t}{\sqrt{N}}\right) + \dots\right\} & \text{for } t > \sqrt{2N} \quad \text{and} \quad |t - \sqrt{2N}| \approx O(\sqrt{N}) \end{cases} \quad (14) \end{aligned}$$

where  $F_\beta(x)$  is the Tracy–Widom distribution and where  $\psi_-$  and  $\psi_+$  are respectively the left and right large deviation functions describing the tails of the distribution of  $\lambda_{\max}$ . The rate function  $\psi_-(z)$  was explicitly computed in [32, 33], while  $\psi_+(z)$  was computed in [35], both by simple physical methods exploiting the Coulomb gas analogy. A more complicated, albeit mathematically rigorous, derivation of  $\psi_+(z)$  in the context of spin

glass models can be found in [36]. These rate functions read

$$\begin{aligned} \psi_-(z) &= \frac{z^2}{3} - \frac{z^4}{108} - \sqrt{z^2 + 6} \frac{(z^3 + 15z)}{108} \\ &\quad - \frac{1}{2} \ln \left[ \frac{\sqrt{z^2 + 6} + z}{\sqrt{2}} \right] + \frac{\ln 3}{2}, \quad \text{for } z < \sqrt{2} \\ \psi_+(z) &= \frac{z\sqrt{z^2 - 2}}{2} + \ln \left[ \frac{z - \sqrt{z^2 - 2}}{\sqrt{2}} \right], \quad \text{for } z > \sqrt{2}. \end{aligned} \quad (15)$$

Note that in [35], the function  $\psi_+(z)$  was expressed in terms of a complicated hypergeometric function, which however can be reduced to a simple algebraic function as presented above in equation (15). Note also that while  $F_\beta(x)$  depends explicitly on  $\beta$ , the rate functions  $\psi_-(z)$  and  $\psi_+(z)$  are independent of  $\beta$ . These rate functions only give the dominant order for large  $N$  in the exponential. In other words, the precise meaning of  $\approx$  is that for large  $N$ :  $\lim_{N \rightarrow \infty} (1/\beta N^2) \ln \mathcal{P}(\lambda_{\max} = z\sqrt{N}) = -\psi_-(z)$  for  $z < \sqrt{2}$  and  $\lim_{N \rightarrow \infty} (1/\beta N) \ln \mathcal{P}(\lambda_{\max} = z\sqrt{N}) = -\psi_+(z)$  for  $z > \sqrt{2}$ . When  $z$  approaches  $\sqrt{2}$  (from below or above) it is easy to see that the rate functions vanish respectively as

$$\psi_-(z) \rightarrow \frac{1}{6\sqrt{2}}(\sqrt{2} - z)^3 + \dots \quad (16)$$

$$\psi_+(z) \rightarrow \frac{2^{7/4}}{3}(z - \sqrt{2})^{3/2} + \dots \quad (17)$$

Note that the physics of the left tail [32, 33] is very different from the physics of the right tail [35]. In the former case, the semi-circular charge density of the Coulomb gas is pushed by the hard wall ( $z < \sqrt{2}$ ) leading to a reorganization of all the  $N$  charges that gives rise to an energy difference of  $O(N^2)$  [32, 33]. In contrast, for the right tail  $z > \sqrt{2}$ , the dominant fluctuations are caused by pulling a single charge away (to the right) from the Wigner sea, leading to an energy difference of  $O(N)$  [35].

The different behaviour of the probability distribution for  $z < \sqrt{2}$  and  $z > \sqrt{2}$  leads to a ‘phase transition’ strictly in the  $N \rightarrow \infty$  limit at the critical point  $z = \sqrt{2}$  in the following sense. Indeed, if one scales  $\lambda$  by  $\sqrt{N}$  and takes the  $N \rightarrow \infty$  limit, one obtains

$$\begin{aligned} - \lim_{N \rightarrow \infty} \frac{1}{\beta N^2} \ln \mathbb{P}_N(\lambda_{\max} \leq z\sqrt{N}) &= \psi_-(z) \quad \text{for } z < \sqrt{2} \\ &= 0 \quad \text{for } z > \sqrt{2}. \end{aligned} \quad (18)$$

Note that since  $\mathbb{P}_N(\lambda_{\max} \leq t)$  can be interpreted as a partition function of a Coulomb gas (see equation (6)), its logarithm has the interpretation of a free energy. Since  $\psi_-(z) \sim (\sqrt{2} - z)^3$  as  $z \rightarrow \sqrt{2}$  from below, the third derivative of the free energy is discontinuous at the critical point  $z = \sqrt{2}$ . Hence, this can be interpreted as a third order phase transition.

However, for finite but large  $N$ , it follows from (14) that the behaviour to the left of  $z = \sqrt{2}$  smoothly crosses over to the behaviour to the right as one varies  $z$  through its critical point  $z = \sqrt{2}$  and the Tracy–Widom distribution in (14) around the critical point is precisely this crossover function. Indeed, if one zooms in close to the mean value  $\sqrt{2N}$  by setting  $t = \sqrt{2N} + xN^{-1/6}/\sqrt{2}$  (for  $\beta = 1, 2$ ) in the rate functions  $\psi_-(t/\sqrt{N})$  and



$\psi_+(t/\sqrt{N})$  in (14), one expects to recover, by taking large  $N$  limit, respectively the left and the right tail of the Tracy–Widom distribution. With this scaling, and using (17), one finds  $\psi_+(t/\sqrt{N}) \approx 2x^{3/2}/3N$  and thus  $\mathcal{P}(\lambda_{\max} = t) \sim \exp\{-(2\beta/3)x^{3/2}\}$ , which indeed matches the dominant order in the far right tail of the Tracy–Widom distribution for  $\beta = 1, 2$  in (13). Similarly for the left tail ( $x < 0$ ), using (16), one finds  $\psi_-(t/\sqrt{N}) \approx |x|^3/24N^2$ , thus  $\mathcal{P}(\lambda_{\max} = t) \sim \exp\{-(\beta/24)|x|^3\}$ , which matches the left tail of the Tracy–Widom distribution in (12).

More recently, higher order corrections for large  $N$  have been computed for the left tail of the distribution [37] using methods developed in the context of matrix models. Note that in [37] a different notation for  $\beta$  was used:  $\beta = 1/2$  (GOE),  $\beta = 1$  (GUE) and  $\beta = 2$  (GSE). To avoid confusion, we present below the results in terms of the standard Dyson index  $\beta = 1, 2, 4$ .

$$\mathcal{P}(\lambda_{\max} = t) \approx \exp\left\{-\Phi_N\left(\frac{t}{\sqrt{N}}, \beta\right)\right\} \quad \text{for } t < \sqrt{2N} \quad \text{and} \quad |t - \sqrt{2N}| \approx O(\sqrt{N}) \quad (19)$$

where

$$\Phi_N(z, \beta) = \beta N^2 \psi_-(z) + N(\beta - 2)\Phi_1(z) + \phi_\beta \ln N + \Phi_0(\beta, z) \quad (20)$$

with  $\psi_-(z)$  given in equation (15) (dominant order). The subleading terms are given by [37]

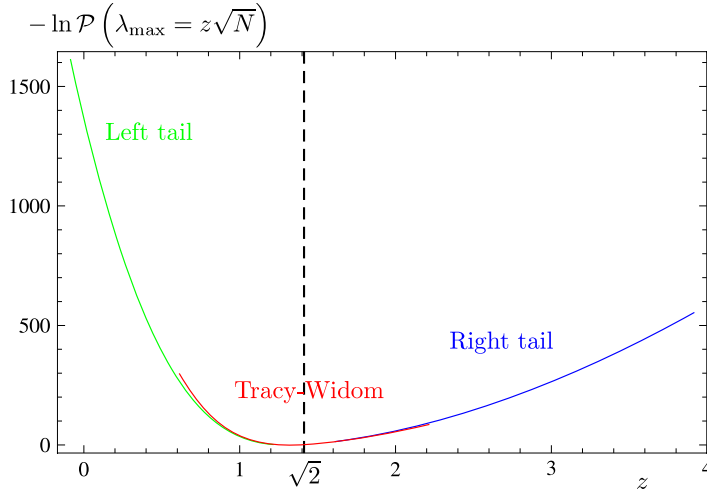
$$\begin{aligned} \Phi_1(z) = & \frac{z^2}{6} - \frac{z\sqrt{z^2+6}}{12} + \frac{z}{4\sqrt{3}}(z^2+6)^{1/4}(\sqrt{z^2+6}-2z)^{1/2} \\ & + \frac{\ln 18}{4} - \frac{1}{2} \ln[2\sqrt{z^2+6} - z + \sqrt{3}(z^2+6)^{1/4}(\sqrt{z^2+6}-2z)^{1/2}] \end{aligned} \quad (21)$$

and

$$\phi_\beta = -\frac{7}{4} - \frac{1}{12} \left( \frac{\beta}{2} + \frac{2}{\beta} \right), \quad (22)$$

and (see equations (4)–(35) in [37])

$$\begin{aligned} \Phi_0(\beta, z) = & \left( \frac{1}{12}\beta + \frac{1}{3\beta} - \frac{1}{3} \right) \ln 2 + \left( \frac{19}{12\beta} + \frac{19\beta}{48} + \frac{9}{8} \right) \ln 3 \\ & + \frac{1}{2} \ln \pi + \left( \frac{-21}{48} + \frac{11}{24} \left( \frac{1}{\beta} + \frac{\beta}{4} \right) \right) \ln[6 + z^2] \\ & + \left( \frac{3}{8} - \frac{1}{4\beta} - \frac{\beta}{16} \right) \ln[-2z + \sqrt{6 + z^2}] \\ & + \left( \frac{1}{2} - \frac{1}{3\beta} - \frac{\beta}{12} \right) \ln[z + \sqrt{6 + z^2}] \\ & + \left( \frac{-4}{3} + \frac{4}{3\beta} + \frac{\beta}{3} \right) \ln \left[ \sqrt{-2z + \sqrt{6 + z^2}} + \sqrt{3}(z^2 + 6)^{1/4} \right] \\ & + \frac{5}{3} \left( 1 - \frac{1}{\beta} - \frac{\beta}{4} \right) \ln \left[ -z + 2\sqrt{6 + z^2} + \sqrt{3}(z^2 + 6)^{1/4} \sqrt{\sqrt{6 + z^2} - 2z} \right] \\ & - \ln \left[ (-18 + z^2)z + (6 + z^2)^{3/2} \right] - \frac{\ln \beta}{2} - \kappa_\beta \end{aligned} \quad (23)$$



**Figure 2.** Rate function  $-\ln \mathcal{P}(\lambda_{\max} = z\sqrt{N})$  associated with the distribution  $\mathcal{P}(\lambda_{\max} = t)$  of the maximal eigenvalue of a random matrix from the GUE for large  $N$ . Close to the mean value  $z = \sqrt{2}$ , the distribution is a Tracy–Widom law (red line), it describes the small typical fluctuations around the mean value. Atypical large fluctuations are described by the large deviations: the left large deviation in green ( $z < \sqrt{2}$ ), the right deviation in blue ( $z > \sqrt{2}$ ).

where  $\kappa_\beta$  is a complicated function of  $\beta$ . For  $\beta/2$  integer, it reduces to [37]

$$\kappa_\beta = \left( \frac{\beta/2 + 1}{4} \right) \ln(2\pi) + \frac{2\zeta'(-1)}{\beta} - \frac{\ln(\beta/2)}{6\beta} - \sum_{m=1}^{(\beta/2)-1} \frac{2m}{\beta} \ln \Gamma \left( 2\frac{m}{\beta} \right). \quad (24)$$

For instance, for the GUE ( $\beta = 2$ ), we find  $\kappa_2 = (\ln(2\pi)/2) + \zeta'(-1)$ . For  $\beta = 1, 2$  and 4, the expression in equation (20) matches the left asymptotics of the Tracy–Widom distribution, i.e. the asymptotic behaviour of  $F'_\beta(x)$  for  $x \rightarrow -\infty$ , see [38]. However, for the right tail of the distribution of  $\lambda_{\max}$ , the corrections to dominant order for large  $N$  are, to our knowledge, not known until now. In fact, one of the results of this paper is to compute these right tail corrections for the GUE ( $\beta = 2$ ). Both left and right large deviations are plotted in figure 2 for the GUE. The left tail is described by  $\Phi_N(z, 2)$  in equation (20), the right tail is described by our result given in equation (26).

Another result of this paper concerns a simpler and pedestrian derivation of the Tracy–Widom distribution for the GUE case. The original derivation of the Tracy–Widom law for the distribution of typical fluctuations of  $\lambda_{\max}$  [6, 7] is somewhat complex as it requires a rather sophisticated and nontrivial asymptotic analysis of the Fredholm determinant involving the Airy Kernel [6, 7]. Since this distribution appears in so many different contexts, it is quite natural to ask if there is any other simpler (more elementary) derivation of the Tracy–Widom distribution. In this paper, we provide such a derivation for the GUE case. Our method is based on a suitable modification of a technique of orthogonal polynomials developed by Gross and Matytsin [39] in the context of two-dimensional Yang–Mills theory. In fact, the partition function of the continuum two-dimensional pure Yang–Mills theory on a sphere (with gauge group  $U(N)$ ) can be written

(up to a prefactor) as a discrete multiple sum over integers [40, 41]

$$Z(A, N) = \sum_{n_1, n_2, \dots, n_N = -\infty}^{\infty} \prod_{1 \leq i < j \leq N} (n_i - n_j)^2 e^{-(A/2N) \sum_{j=1}^N n_j^2} \quad (25)$$

where  $A$  is the area of the sphere. In the  $N \rightarrow \infty$  limit, the free energy  $\ln Z$ , as a function of  $A$ , undergoes a third order phase transition known as the Douglas–Kazakov transition [42] at the critical value  $A_c = \pi^2$ . For  $A > A_c$ , the system is in the ‘strong’ coupling phase while for  $A < A_c$ , it is in the ‘weak’ coupling phase. For finite but large  $N$ , there is a crossover between the two phases as one passes through the vicinity of the critical point. In the so called double scaling limit (where  $A \rightarrow A_c$ ,  $N \rightarrow \infty$  but keeping the product  $(A - A_c)N^{2/3}$  fixed), the singular part of the free energy satisfies a Painlevé II equation [39]. Gross and Matytsin (see also [43]) used a method based on orthogonal polynomials to analyse the partition sum in the double scaling limit, as well as in the weak coupling regime ( $A < A_c$ ), where they were able to compute non-perturbative (in  $1/N$  expansion) corrections to the free energy. Actually, a similar third order phase transition from a weak to strong coupling phase in the  $N \rightarrow \infty$  limit was originally noticed in the lattice formulation (with Wilson action) of the two-dimensional  $U(N)$  gauge theory [44, 45, 14] and in the vicinity of the transition point the singular part of the free energy was shown to satisfy a Painlevé II equation [46]. Note that similar calculations involving the asymptotic analysis of partition functions using orthogonal polynomials were used extensively in the early 1990s to study the double scaling limit of the so called one-matrix model (for a recent review and developments, see e.g. [47]).

In our case, for the distribution of  $\lambda_{\max}$ , we need to analyse the asymptotic large  $N$  behaviour of a multiple indefinite integral in equation (6), as opposed to the discrete sum in equation (25). However, we show that one can suitably modify the orthogonal polynomial method of Gross and Matytsin to analyse the multiple integral in equation (6) in the limit of large  $N$ . In fact, we find a similar third order phase transition (in the  $N \rightarrow \infty$  limit) in the largest eigenvalue distribution  $\mathbb{P}_N(\lambda_{\max} \leq t)$  as a function of  $t$  at the critical point  $t_c = \sqrt{2N}$ . The regime of left large deviation of  $\mathbb{P}_N(\lambda_{\max} \leq t)$  ( $t < t_c$ ) is similar to the ‘strong’ coupling regime ( $A > A_c$ ) of the Yang–Mills theory, while the right large deviation tail of  $\mathbb{P}_N(\lambda_{\max} \leq t)$  ( $t > t_c$ ) is similar to the ‘weak’ coupling regime ( $A < A_c$ ) of the Yang–Mills theory. For finite but large  $N$ , the crossover function across the critical point that connects the left and right large deviation tails is precisely the Tracy–Widom distribution. Thus the Tracy–Widom distribution corresponds precisely to the double scaling limit of the Yang–Mills theory and one finds the same Painlevé II equation. A similar third order phase transition was also found recently in a model of non-intersecting Brownian motions by establishing an exact correspondence between the reunion probability in the Brownian motion model and the partition function in the 2d Yang–Mills theory on a sphere [27, 48].

The advantage of this orthogonal polynomial method to analyse the maximum eigenvalue distribution is two fold: (i) one gets the Tracy–Widom distribution in a simple elementary way (basically one carries out a scaling analysis of a pair of nonlinear recursion relations near the critical point and shows that the scaling function satisfies a Painlevé II differential equation) and (ii) as an added bonus, we also obtain precise subleading corrections to the leading right large deviation tail ( $t > \sqrt{2N}$ ). The subleading corrections,

in the Yang–Mills language, correspond to the non-perturbative corrections in the weak coupling regime as derived by Gross and Matytsin [39]. More precisely we show that

$$\mathcal{P}(\lambda_{\max} = t) \approx \frac{\sqrt{N}}{2\pi\sqrt{2}(t^2 - 2N)} e^{-2N\psi_+(t/\sqrt{N})} \quad \text{for } t > \sqrt{2N}, \quad |t - \sqrt{2N}| \approx O(\sqrt{N}) \quad (26)$$

where  $\psi_+(z)$  is given in equation (15). Note that only the leading behaviour  $\exp[-2N\psi_+(z)]$  was known before [35], but the subleading corrections are, to our knowledge, new results. We also verify that our expression matches the precise right asymptotics of the Tracy–Widom distribution. Figure 2 shows the distribution of  $\lambda_{\max}$  for the GUE: close to the mean value it is described by the Tracy–Widom distribution, whereas the tails are described by the large deviations. The right tail (right large deviation) is given by our result in equation (26). Together with the subleading terms in the left tail in equation (20), our new result in equation (26) then provides a rather complete picture of the tail behaviours of the distribution of  $\lambda_{\max}$  on both sides of the mean  $\sqrt{2N}$ .

The rest of the paper is organized as follows. In section 2, we start with some general notations and scaling remarks for the GUE. In section 3, we explain the method of orthogonal polynomials on a semi-infinite interval and derive some basic recursion relations. In section 4, we compute the right tail of the distribution of  $\lambda_{\max}$  (dominant order and corrections for the GUE): it describes atypical large fluctuations of  $\lambda_{\max}$  to the right of its mean value. In section 5, using results of the previous sections and basic scaling remarks, we derive the Tracy–Widom law (with  $\beta = 2$  for the GUE) that describes small typical fluctuations close to the mean value.

## 2. Notations and scaling

In the rest of the paper we focus only on Gaussian random matrices  $X$  drawn from the GUE ( $\beta = 2$ ). They are Hermitian random matrices  $X^\dagger = X$  such that  $\mathcal{P}(X) \propto e^{-\alpha \text{Tr}(X^2)}$ , where we have introduced an additional parameter  $\alpha > 0$  for the purpose of certain mathematical manipulations that will be clear later. Setting  $\alpha = 1$  at the end of the calculation, we will recover the usual GUE. With the additional parameter  $\alpha$ , the joint distribution of the eigenvalues of  $X$  is given by (see equation (2)):

$$\mathcal{P}(\lambda_1, \dots, \lambda_N) = B_N(\alpha) e^{-\alpha \sum_{i=1}^N \lambda_i^2} \prod_{j < k} (\lambda_j - \lambda_k)^2 \quad (27)$$

where  $B_N(\alpha) = (2\alpha)^{N^2/2} (2\pi)^{-N/2} / [\prod_{n=1}^N n!]$  is the normalization constant. The Vandermonde determinant appears with a power 2 as we consider the GUE ( $\beta = 2$ ). This determinant indeed comes from a Jacobian for the change of variables from the entries of the matrix to its eigenvalues. The power is related to the number of real degrees of freedom of an element of the matrix, which is 2 for complex entries, i.e., for GUE (it is 1 for real entries GOE and 4 for quaternion entries GSE). As we will see later, this power 2 is crucial for the method of orthogonal polynomials to work. The technique of Gross and Matytsin [39] that we adapt here also works only for the GUE  $\beta = 2$  case.

Given the joint distribution of eigenvalues in equation (27), it is easy to write down the cumulative distribution of the maximal eigenvalue  $\lambda_{\max}$

$$\mathbb{P}_N(\lambda_{\max} \leq y) = \text{Prob}[\lambda_1 \leq y, \lambda_2 \leq y, \dots, \lambda_N \leq y] = \frac{Z_N(y, \alpha)}{Z_N(\infty, \alpha)} \quad (28)$$

where the partition function  $Z_N$  is given by the multiple indefinite integral

$$Z_N(y, \alpha) = \frac{1}{N!} \prod_{i=1}^N \int_{-\infty}^y d\lambda_i \prod_{j < k} (\lambda_j - \lambda_k)^2 e^{-\alpha \sum_{i=1}^N \lambda_i^2}. \quad (29)$$

The normalization  $Z_N(\infty, \alpha)$  is actually related to  $B_N(\alpha)$  in equation (27) as  $B_N(\alpha) = 1/(N! Z_N(\infty, \alpha))$ . Note that, by the trivial rescaling  $\sqrt{\alpha} \lambda_i \rightarrow \lambda_i$  in (29), it follows from (28) that

$$\mathbb{P}_N(\lambda_{\max} \leq y) = \frac{Z_N(y, \alpha)}{Z_N(\infty, \alpha)} = \frac{Z_N(y\sqrt{\alpha}, 1)}{Z_N(\infty, 1)}. \quad (30)$$

Thus  $y$  and  $\alpha$  always appear in a single scaling combination  $y\sqrt{\alpha}$ .

We will henceforth focus on the large  $N$  limit. For fixed  $\alpha$ , one can easily figure out from the joint pdf in equation (27) how a typical eigenvalue  $\lambda_{\text{typ}}$  scales with  $N$  for large  $N$ . Let us rewrite the joint distribution of eigenvalues in equation (27) as

$$\mathcal{P}(\lambda_1, \dots, \lambda_N) \propto \exp \left\{ -\alpha \sum_i \lambda_i^2 + 2 \sum_{j < k} \ln |\lambda_j - \lambda_k| \right\} \quad (31)$$

which can then be interpreted as a Boltzmann weight  $\propto \exp[-E_{\text{eff}}]$ , with effective energy  $E_{\text{eff}} = \alpha \sum_i \lambda_i^2 - 2 \sum_{j < k} \ln |\lambda_j - \lambda_k|$ . The eigenvalues can thus be seen as the positions of  $N$  charges of a 2D Coulomb gas (but restricted to be on the real line) which repel each other via a logarithmic Coulomb potential (coming from the Vandermonde determinant in equation (27)) [12]. In addition, the charges are subjected to an external confining parabolic potential. For large  $N$ , the first term of the energy is of order  $N\lambda_{\text{typ}}^2$ , whereas the second is of order  $N^2$  because of the double sum. Balancing the two terms  $N\lambda_{\text{typ}}^2 \sim N^2$  gives the scaling of a typical eigenvalue for large  $N$ :  $\lambda_{\text{typ}} \sim \sqrt{N}$ . For large  $N$ , the eigenvalues are close to each other and they can be described by a continuous charge density (normalized to unity)  $\rho_N(\lambda) = (1/N) \sum_i \delta(\lambda - \lambda_i)$ . The average density of states for large  $N$  can be obtained by evaluating the full partition function  $Z_N(\infty, \alpha)$  (the denominator in equation (28)) via a saddle point method. The saddle point density is the density that minimizes the effective energy (see the book of Mehta [2])  $E_{\text{eff}} = \alpha N \int d\lambda \rho_N(\lambda) \lambda^2 - N^2 \int d\lambda \int d\lambda' \rho_N(\lambda) \rho_N(\lambda') \ln |\lambda - \lambda'|$  (in its continuous version). This gives the well-known semi-circle law (which is exactly the same as in equation (3) for  $\alpha = 1$ ):

$$\rho_N(\lambda) = \frac{1}{\sqrt{N}} \rho \left( \frac{\lambda}{\sqrt{N}} \right) \quad \text{with } \rho(x) = \frac{\alpha}{\pi} \sqrt{\frac{2}{\alpha} - x^2}. \quad (32)$$

The density is plotted in figure 1.

The average value of  $\lambda_{\max}$  is again given for large  $N$  by the upper bound of the density support (see figure 1):

$$\langle \lambda_{\max} \rangle \approx \sqrt{\frac{2N}{\alpha}} \quad \text{for large } N. \quad (33)$$

For  $\alpha = 1$ , this evidently reduces to the usual expression for  $\langle \lambda_{\max} \rangle$ . The typical scaling for large  $N$  is thus  $\lambda_{\max} \sim \sqrt{N}$ . Hence, we will use  $\lambda_{\max} = z\sqrt{N}$ , where  $z$  is of order one.

### 3. Orthogonal polynomials

In this section, we introduce the method of orthogonal polynomials to evaluate the partition function in equation (29). As mentioned in the introduction, to evaluate this multiple indefinite integral we will adapt the method developed by Gross and Matytsin [39] to enumerate the partition sum (25) in the two-dimensional Yang–Mills theory. Orthogonal polynomials are very useful to handle the square Vandermonde determinant in the distribution of the eigenvalues in equation (27). A Vandermonde determinant can indeed be written as  $\prod_{i < j} (\lambda_j - \lambda_i) = \det(\lambda_i^{j-1})_{i,j} = \det(p_{j-1}(\lambda_i))_{i,j}$ , where  $p_j(\lambda) = \lambda^j + \dots$  is any polynomial of degree  $j$  with leading coefficient one. The idea is to choose these polynomials  $p_j$  well in order to simplify the computation of the integral.

We define an operation on pairs of polynomials as follows:

$$\langle f, g \rangle = \int_{-\infty}^y d\lambda e^{-\alpha\lambda^2} f(\lambda)g(\lambda). \quad (34)$$

We consider a family  $\{p_n\}_{n \geq 0}$  of orthogonal polynomials with respect to the operation defined above, i.e. with weight  $e^{-\alpha\lambda^2}$  on the interval  $]-\infty, y]$ . Without any loss of generalization, we define the polynomial  $p_n(\lambda)$  of degree  $n$  such that the coefficient of the  $\lambda^n$  term is always fixed to be 1, i.e.,  $p_n(\lambda) = \lambda^n + \dots$ . These polynomials satisfy the orthogonality property:  $\langle p_n, p_m \rangle = 0$  for all  $n \neq m$ . We also write  $h_n = \langle p_n, p_n \rangle$ . Thus,

$$\langle p_n, p_m \rangle = \delta_{n,m} h_n \quad \text{for all } n \geq 0. \quad (35)$$

Note that  $p_n(\lambda)$  and  $h_n$  are implicitly functions of  $\alpha$  and  $y$ , i.e.  $p_n(\lambda) = p_n(\lambda|y, \alpha)$  and  $h_n = h_n(y, \alpha)$ . In particular, we can easily compute by hand the first few  $p_n(\lambda)$ s for fixed  $\alpha > 0$  and  $y$ , but the expressions become rather complex as  $n$  increases and it is hard to find a closed form expression for  $p_n(\lambda)$  for every  $n$  (except for the limiting case  $y \rightarrow \infty$ ):

$$p_0(\lambda) = 1$$

$$p_1(\lambda) = \lambda + \frac{e^{-\alpha y^2}}{\sqrt{\pi\alpha}[1 + \operatorname{erf}(y\sqrt{\alpha})]}$$

$$p_2(\lambda) = \lambda^2 + \frac{-2y\sqrt{\alpha} - e^{y^2\alpha}\sqrt{\pi}(1 + 2y^2\alpha)(1 + \operatorname{erf}[y\sqrt{\alpha}])}{\sqrt{\alpha}(2 + 2e^{y^2\alpha}\sqrt{\pi}y\sqrt{\alpha}(1 + \operatorname{erf}[y\sqrt{\alpha}]) - e^{2y^2\alpha}\pi(1 + \operatorname{erf}[y\sqrt{\alpha}])^2)}\lambda + \frac{-4 - 4e^{y^2\alpha}\sqrt{\pi}y\sqrt{\alpha}(1 + \operatorname{erf}[y\sqrt{\alpha}]) + e^{2y^2\alpha}\pi(1 + \operatorname{erf}[y\sqrt{\alpha}])^2}{4\alpha + 4e^{y^2\alpha}\sqrt{\pi}y\alpha^{3/2}(1 + \operatorname{erf}[y\sqrt{\alpha}]) - 2e^{2y^2\alpha}\pi\alpha(1 + \operatorname{erf}[y\sqrt{\alpha}])^2}.$$

Thus we get for instance  $h_0 = \langle p_0|p_0 \rangle = (\sqrt{\pi}/[2\sqrt{\alpha}])[1 + \operatorname{erf}(y\sqrt{\alpha})]$  and  $h_1 = \langle p_1|p_1 \rangle = (-2y\sqrt{\alpha}e^{-\alpha y^2} + \sqrt{\pi}[1 + \operatorname{erf}(y\sqrt{\alpha})] - 2e^{-2\alpha y^2}/(\sqrt{\pi}[1 + \operatorname{erf}(y\sqrt{\alpha})]))/4\alpha^{3/2}$ , etc. In the limit  $y \rightarrow \infty$ , we recover the Hermite polynomials:  $p_0 = 1$ ,  $p_1 = \lambda$ ,  $p_2 = \lambda^2 - 1/(2\alpha)$  and  $h_0 = \sqrt{\pi}/\sqrt{\alpha}$ ,  $h_1 = \sqrt{\pi}/(2\alpha^{3/2})$ ,  $h_2 = \sqrt{\pi}/(2\alpha^{5/2})$ .



### 3.1. Partition function

The partition function  $Z_N(y, \alpha)$  in equation (29) can be written as a function of the  $h_n$ s. By combination of rows, the Vandermonde determinant in the joint distribution of the eigenvalues can indeed be written

$$\prod_{j < k} (\lambda_k - \lambda_j) = \det(\lambda_i^{j-1})_{i,j} = \det(p_{j-1}(\lambda_i))_{i,j}.$$

Then, the partition function can be expressed as

$$\begin{aligned} Z_N(y, \alpha) &= \frac{1}{N!} \prod_{i=1}^N \int_{-\infty}^y d\lambda_i \prod_{j < k} (\lambda_j - \lambda_k)^2 e^{-\alpha \sum_{i=1}^N \lambda_i^2} \\ &= \frac{1}{N!} \prod_{i=1}^N \int_{-\infty}^y d\lambda_i \det(p_{j-1}(\lambda_i))_{i,j} \det(p_{l-1}(\lambda_k))_{k,l} e^{-\alpha \sum_{i=1}^N \lambda_i^2} \\ &= \det \left[ \int_{-\infty}^y d\lambda e^{-\alpha \lambda^2} p_{i-1}(\lambda) p_{j-1}(\lambda) \right]_{i,j} \\ &= \det(\langle p_{i-1} | p_{j-1} \rangle)_{i,j} = \det(\delta_{i,j} h_{i-1})_{i,j} = \prod_{i=0}^{N-1} h_i \end{aligned}$$

where in going from the second to the third line we have used the Cauchy–Binet formula [2]. Note that this step works only for  $\beta = 2$ . Therefore the partition function reduces to:

$$Z_N(y, \alpha) = \prod_{n=0}^{N-1} h_n(y, \alpha). \tag{36}$$

Thus the integral  $Z_N(y, \alpha)$  is now expressed as a product of the coefficients  $h_n$ . The goal of next subsection is to find recursion relations for the  $h_n$ s in order to compute them and subsequently analyse their product  $Z_N$  in equation (36) in the large  $N$  limit.

### 3.2. Recursion relations

In general, for orthogonal polynomials (with any reasonable weight function), one can write a recursion relation of the form:

$$\lambda p_n(\lambda) = p_{n+1}(\lambda) + S_n p_n(\lambda) + R_n p_{n-1}(\lambda) \tag{37}$$

where  $S_n$  and  $R_n$  are real coefficients. This relation comes from the fact that  $p_n = \lambda^n + \dots$  and that  $\langle p_n | q \rangle = 0$  for any polynomial  $q(\lambda)$  of degree strictly smaller than  $n$ . The coefficients  $S_n$  and  $R_n$  are functions of  $\alpha$  and  $y$ , i.e.  $S_n = S_n(y, \alpha)$  and  $R_n = R_n(y, \alpha)$ .

Let us first demonstrate that the coefficients  $R_n$  and  $S_n$  can be expressed in terms of  $h_n$ s. From equation (37), we get:  $\langle p_{n-1} | \lambda p_n \rangle = R_n \langle p_{n-1} | p_{n-1} \rangle = R_n h_{n-1}$ . On the other hand, we have  $\langle p_{n-1} | \lambda p_n \rangle = \langle \lambda p_{n-1} | p_n \rangle = \langle p_n + S_{n-1} p_{n-1} + R_{n-1} p_{n-2} | p_n \rangle = \langle p_n | p_n \rangle = h_n$ . Therefore  $R_n h_{n-1} = h_n$ , thus  $R_n = h_n / h_{n-1}$ .

From equation (37) again, we also get:  $\langle p_n | \lambda p_n \rangle = S_n \langle p_n | p_n \rangle = S_n h_n$ . By definition we have  $\langle p_n | \lambda p_n \rangle = \int_{-\infty}^y d\lambda e^{-\alpha \lambda^2} \lambda p_n^2(\lambda)$ . After integrating by parts we find:  $\langle p_n | \lambda p_n \rangle = -(1/(2\alpha)) e^{-\alpha y^2} p_n^2(y) = -(1/(2\alpha)) \partial h_n(y, \alpha) / \partial y$ . The last equality

follows from the definition of  $h_n$ . As  $h_n = \int_{-\infty}^y d\lambda e^{-\alpha\lambda^2} p_n^2(\lambda)$ , we have indeed  $\partial h_n(y, \alpha)/\partial y = e^{-\alpha y^2} p_n^2(y|y, \alpha) + 2\langle p_n|\partial p_n/\partial y\rangle$ . However,  $\langle p_n|\partial p_n/\partial y\rangle = 0$  as  $\partial p_n/\partial y$  is a polynomial of degree strictly smaller than  $n$  (as  $p_n(\lambda|y, \alpha) = \lambda^n + \dots$ ). Therefore  $S_n = -(1/(2\alpha)) \partial \ln h_n/\partial y$ .

Combining these relations,  $R_n$  and  $S_n$  are then given by:

$$R_n(y, \alpha) = \frac{h_n(y, \alpha)}{h_{n-1}(y, \alpha)} \quad \text{and} \quad S_n(y, \alpha) = -\frac{1}{2\alpha} \frac{\partial \ln h_n(y, \alpha)}{\partial y} \quad (38)$$

Iterating the recursion relation  $h_n = R_n h_{n-1}$  starting from  $n = 1$ , we can write  $h_n$  in terms of  $R_k$ s:

$$h_n = \left( \prod_{k=1}^n R_k \right) h_0. \quad (39)$$

Substituting this result in equation (36), we see that the partition function  $Z_N$  can be entirely expressed in terms of a product over the  $R_n$ s. Thus, if we can determine  $R_n$ s, we can evaluate the partition function explicitly.

Thus the next task is to determine the  $R_n$ s. To do this, we will first derive a set of coupled recursion relations for  $R_n$ s and  $S_n$ s. We have  $\partial h_n/\partial \alpha = \partial \langle p_n|p_n\rangle/\partial \alpha = -\int_{-\infty}^y d\lambda e^{-\alpha\lambda^2} \lambda^2 p_n^2(\lambda) = -\langle \lambda p_n|\lambda p_n\rangle$  where we have used the fact that  $\langle p_n|\partial p_n/\partial \alpha\rangle = 0$  (which follows from the observation that  $\partial p_n/\partial \alpha$  is a polynomial of degree strictly less than  $n$  and hence orthogonality dictates that it is identically zero). On the other hand, using equation (37) we find:  $\langle \lambda p_n|\lambda p_n\rangle = \langle p_{n+1} + S_n p_n + R_n p_{n-1}|p_{n+1} + S_n p_n + R_n p_{n-1}\rangle = h_{n+1} + S_n^2 h_n + R_n^2 h_{n-1} = h_n(R_{n+1} + S_n^2 + R_n)$ . Therefore,

$$-\frac{\partial \ln h_n}{\partial \alpha} = R_n + R_{n+1} + S_n^2. \quad (40)$$

We can eliminate  $h_n$  from the relations (38) and (40) and get a pair of coupled nonlinear recursion relations for  $R_n$  and  $S_n$ . Using equation (40) for  $n$  and  $n - 1$ , and as  $R_n = h_n/h_{n-1}$ , we find  $-\partial \ln R_n/\partial \alpha = R_{n+1} - R_{n-1} + S_n^2 - S_{n-1}^2$ . Using equation (38), we also find  $\partial \ln R_n/\partial y = 2\alpha(S_{n-1} - S_n)$ . Finally, we then get our desired recursion relations:

$$R_{n+1} = -\frac{\partial \ln R_n}{\partial \alpha} + R_{n-1} - S_n^2 + S_{n-1}^2 \quad (41)$$

$$S_n = S_{n-1} - \frac{1}{2\alpha} \frac{\partial \ln R_n}{\partial y}. \quad (42)$$

It is easy to show by induction that the two relations (41) and (42) together with the initial conditions  $R_0$ ,  $R_1$  and  $S_0$  uniquely determine  $R_n$  and  $S_n$ . The additional initial condition  $h_0$  is enough to determine  $h_n$  as given in equation (39).

By definition,  $p_0$  is a polynomial of degree 0 with dominant coefficient 1. Thus  $p_0(\lambda|y, \alpha) = 1$ . Therefore  $h_0(y, \alpha) = \langle p_0|p_0\rangle = \int_{-\infty}^y d\lambda e^{-\alpha\lambda^2} = (1/2)\sqrt{\pi/\alpha}(1 + \text{erf}(y\sqrt{\alpha}))$ . We also have  $R_0(y, \alpha) = 0$  as the recursion relation in equation (37) must reduce for  $n = 0$  to  $\lambda p_0(\lambda) = p_1(\lambda) + S_0 p_0(\lambda)$ , i.e.,  $p_1(\lambda) = \lambda - S_0$ . Moreover we get from equation (38)  $S_0 = -(1/(2\alpha))\partial \ln h_0/\partial y = -e^{-\alpha y^2}/(2\alpha h_0) = -e^{-\alpha y^2}/[\sqrt{\pi\alpha}(1 + \text{erf}(y\sqrt{\alpha}))]$ . We now have  $R_0$ ,  $S_0$  and  $h_0$ , we can thus determine  $R_1$  from equation (40) for  $n = 0$ :



$R_1 = -\partial \ln h_0 / \partial \alpha - R_0 - S_0^2 = yS_0 + 1/(2\alpha) - S_0^2$ . Thus, the initial conditions can be summarized as:

$$\begin{aligned} R_0(y, \alpha) &= 0, & p_0(\lambda|y, \alpha) &= 1, \\ h_0(y, \alpha) &= \int_{-\infty}^y d\lambda e^{-\alpha\lambda^2} = \frac{1}{2} \sqrt{\frac{\pi}{\alpha}} (1 + \operatorname{erf}(y\sqrt{\alpha})) \\ S_0(y, \alpha) &= -\frac{1}{2\alpha} \frac{\partial \ln h_0}{\partial y} = -\frac{e^{-\alpha y^2}}{2\alpha h_0} = -\frac{e^{-\alpha y^2}}{\sqrt{\pi\alpha} (1 + \operatorname{erf}(y\sqrt{\alpha}))} \\ R_1(y, \alpha) &= yS_0 + \frac{1}{2\alpha} - S_0^2. \end{aligned} \tag{43}$$

### 3.3. Normalization: limit $y \rightarrow \infty$

As  $y \rightarrow \infty$ , we can explicitly compute the functions  $S_n$ ,  $R_n$  and  $h_n$ . As mentioned above, in the limit  $y \rightarrow \infty$  the polynomials  $p_n$  are indeed simply the Hermite polynomials. Hence everything can then be computed explicitly in this case. We have  $h_0(\infty, \alpha) = \int_{-\infty}^{\infty} d\lambda e^{-\alpha\lambda^2} = \sqrt{\pi/\alpha}$  and  $S_0(\infty, \alpha) = 0$ . Then, by recurrence it is easy to show that:

$$S_n(\infty, \alpha) = 0 \quad R_n(\infty, \alpha) = \frac{n}{2\alpha}. \tag{44}$$

Finally, using equation (39) we get

$$h_n = \sqrt{\frac{\pi}{\alpha}} \frac{n!}{(2\alpha)^n} \tag{45}$$

and thus (see equation (36))

$$Z_N(\infty, \alpha) = (2\pi)^{N/2} (2\alpha)^{-N^2/2} \prod_{n=1}^N \Gamma(n) \tag{46}$$

which could also have been computed directly using Selberg's integral. We recover the normalization  $B_N = 1/(N! Z_N)$  in equation (27).

## 4. Right tail of the distribution of $\lambda_{\max}$ : large deviation function

In section 3 we derived a pair of coupled recursion equations (41) and (42) with initial conditions given in equation (43) that determines uniquely  $R_n$ ,  $S_n$  and thus  $h_n$  and subsequently  $Z_N$  via (36). However, these equations are hard to solve explicitly for general  $n$  and  $y$  (apart from the case  $y = +\infty$  as shown in section 3.3). In this section, we derive an approximate solution for  $Z_N$  and hence for the cdf (28), in the large  $N$  limit where  $N$  is the number of eigenvalues of the matrix  $X$  and we will see that this solution for cdf is valid only for  $\lambda_{\max} > \langle \lambda_{\max} \rangle = \sqrt{2N/\alpha}$ , i.e., it only describes the right tail of the probability distribution.

We have seen in section 2 that for large  $N$  and fixed  $\alpha$ , the maximal eigenvalue typically scales as  $\lambda_{\max} \sim \sqrt{N}$ . We are going to work on this scale, hence in the definition of the maximal eigenvalue cdf in (28) and (29), we will set  $y = z\sqrt{N}$  with  $z \approx O(1)$ . We will then work in the limit of large  $N$  with  $z$  fixed. With this scaling, the operation  $\langle f, g \rangle$  defined in equation (34) for polynomials depends on  $N$  (since the upper limit of

integration in (34) is now  $z\sqrt{N}$ ). The coefficients  $R_n$ ,  $S_n$  and  $h_n$ , for a given  $n$ , are also now implicitly functions of  $N$ . We can make an expansion of these parameters for large  $N$  and fixed  $n$ . The dominant order will be given by the  $y = +\infty$  case (as in section 3.3) as  $y = z\sqrt{N} \rightarrow \infty$  as  $N \rightarrow \infty$  for fixed  $z$ . In this section, we want to determine the first correction to this dominant term.

However the partition function  $Z_N$  and thus the cumulative distribution (cdf) of  $\lambda_{\max}$  is a product of all the  $h_n$ s for  $0 \leq n < N$ . Our expansion will thus give us the behaviour of the cdf of  $\lambda_{\max}$  for large  $N$  only if we can show that it is valid not only for fixed  $n$  but also for  $n$  of order up to  $N$ . This constraint of validity will be discussed later. We will see that this expansion is actually valid only on the right of the mean value, i.e. for  $y > \sqrt{2N/\alpha}$  or equivalently  $z > \sqrt{2/\alpha}$ . This method allows us to describe the right tail of the large deviation of the distribution of  $\lambda_{\max}$ , i.e.  $\mathcal{P}(\lambda_{\max} = t)$  with  $t > \sqrt{2N/\alpha}$  and  $|t - \sqrt{2N/\alpha}| \approx O(\sqrt{N})$ .

#### 4.1. Expansion of $R_n$ and $S_n$

Let us start by expanding the initial conditions for large  $N$ . With the scaling  $y = z\sqrt{N}$  with  $z \approx O(1)$ , equations (43) become (for  $z > 0$ ) for large  $N$ :

$$\begin{aligned} h_0(z\sqrt{N}, \alpha) &\approx \sqrt{\frac{\pi}{\alpha}} - \frac{1}{2z\alpha\sqrt{N}} e^{-N\alpha z^2} + \dots \\ S_0(z\sqrt{N}, \alpha) &\approx -\frac{1}{2\sqrt{\alpha\pi}} e^{-N\alpha z^2} \quad R_1(z\sqrt{N}, \alpha) \approx \frac{1}{2\alpha} - \frac{z\sqrt{N}}{2\sqrt{\alpha\pi}} e^{-N\alpha z^2} \end{aligned} \quad (47)$$

The dominant term for large  $N$  corresponds to the limit  $y \rightarrow \infty$  (see section 3):  $\sqrt{\pi/\alpha} = \int_{-\infty}^{+\infty} d\lambda e^{-\alpha\lambda^2} = h_0(\infty, \alpha)$ . Therefore, ignoring the exponentially small correction for large  $N$  leads to  $R_n(z\sqrt{N}, \alpha) \approx R_n(\infty, \alpha) = n/(2\alpha)$  and  $S_n(z\sqrt{N}, \alpha) \approx 0$ .

We want to determine the first correction for large  $N$ . Let us make the following ansatz:

$$R_n(z\sqrt{N}, \alpha) \approx \frac{n}{2\alpha} + c_n e^{-N\alpha z^2} \quad \text{and} \quad S_n(z\sqrt{N}, \alpha) \approx d_n e^{-N\alpha z^2} \quad (48)$$

where  $c_n = c_n(z\sqrt{N}, \alpha)$  and  $d_n = d_n(z\sqrt{N}, \alpha)$  are expected to be polynomials of  $z\sqrt{N}$ . This will be valid as long as  $c_n(y, \alpha) e^{-N\alpha z^2} \ll n/(2\alpha)$ , where  $y = z\sqrt{N}$ .

The initial conditions in equation (47) give:

$$c_0(y, \alpha) = 0, \quad c_1(y, \alpha) = -\frac{y}{2\sqrt{\alpha\pi}} \quad \text{and} \quad d_0(y, \alpha) = -\frac{1}{2\sqrt{\alpha\pi}}. \quad (49)$$

Let us replace  $R_n$  and  $S_n$  in the recursion equations (41) and (42) by the ansatz in equation (48). We see that  $S_n^2$  and  $S_{n-1}^2$  are actually negligible in the equation (41) giving  $R_{n+1}$ , as they are exponentially smaller than the  $R_k$ s. We thus get recursion relations for the coefficients  $c_n(y, \alpha)$  and  $d_n(y, \alpha)$ :

$$c_{n+1} - c_{n-1} = \frac{2}{n} \left\{ -\frac{\partial(\alpha c_n)}{\partial \alpha} + \alpha y^2 c_n \right\} \quad d_n - d_{n-1} = \left( 2\alpha y c_n - \frac{\partial c_n}{\partial y} \right) \frac{1}{n} \quad (50)$$

$h_n$  is the product of the  $R_k$ , it can thus be expressed in terms of  $c_k$ . As  $Z_N$  is a function of the  $h_n$ s, it is thus a function of  $h_0$  and the  $R_n$ s for  $1 \leq n < N$ , we can use the ansatz (48)

only if  $c_n(z\sqrt{N}, \alpha)e^{-N\alpha z^2} \ll n/(2\alpha)$  for all  $n < N$ . In this case we can write:

$$\begin{aligned} \ln h_n(y, \alpha) &= \ln h_0 + \sum_{k=1}^n \ln R_k \approx \ln h_0 + \sum_{k=1}^n \left[ \ln \left( \frac{k}{2\alpha} \right) + \frac{2\alpha c_k}{k} e^{-N\alpha z^2} \right] \\ &\approx \ln h_n(\infty, \alpha) + \left[ -\frac{1}{2y\sqrt{\pi}} + \sum_{k=1}^n \frac{2\alpha c_k(y, \alpha)}{k} \right] e^{-N\alpha z^2} \end{aligned} \quad (51)$$

where  $y = z\sqrt{N}$ , and  $\ln Z_N(y, \alpha) = \sum_{n=0}^{N-1} \ln h_n(y, \alpha)$  is thus given by

$$\ln Z_N(y, \alpha) \approx \ln Z_N(\infty, \alpha) + \left[ -\frac{N}{2y\sqrt{\pi}} + \sum_{n=0}^{N-1} \sum_{k=1}^n \frac{2\alpha c_k(y, \alpha)}{k} \right] e^{-N\alpha z^2}. \quad (52)$$

The partition function only depends on the  $c_k$ s. We want now to solve the recursion relation for the  $c_k$ s in equation (50). We do not need to determine the  $d_k$ s.

#### 4.2. Solution of the recursion for the $c_n$

Let us define  $\xi$  and  $G_n$  such that:

$$\xi = \alpha y^2 = N\alpha z^2 \quad \text{and} \quad c_n(y, \alpha) = -\frac{y^2}{2\sqrt{\pi\xi}} G_n(\xi) \quad (53)$$

$\xi$  is large for large  $N$  (proportional to  $N$ ).  $G_n(\xi)$  depends only on  $\xi = \alpha y^2$ . This can easily be shown by recurrence with initial condition  $G_1(\xi) = 1$  (as  $c_1$  is given by equation (49)). The recursion (50) for the  $c_n$ s becomes

$$G_{n+1}(\xi) - G_{n-1}(\xi) = \frac{2}{n} \left\{ \left( \xi - \frac{1}{2} \right) G_n(\xi) + \xi G_n'(\xi) \right\} \quad (54)$$

with initial condition  $G_0(\xi) = 0$  and  $G_1(\xi) = 1$ . By recurrence again, it is easy to show that  $G_n(\xi)$  is a polynomial of  $\xi$  of degree  $(n-1)$ , with leading coefficient  $2^{n-1}/(n-1)!$ .

Let us consider the generating function of the  $\{G_n(\xi)\}$ :

$$F(\xi, x) = \sum_{n=1}^{\infty} x^n G_n(\xi). \quad (55)$$

The  $G_n(\xi)$  are obtained from  $F$  by a contour integration:

$$G_n(\xi) = \oint_C \frac{dx}{2i\pi} \frac{1}{x^{n+1}} F(\xi, x) \quad (56)$$

where  $C$  is a contour in the complex plane that encircles the origin  $x = 0$  in such a way that all singularities of  $F(\xi, x)$  (as a function of  $x$  for fixed  $\xi$ ) are outside the contour.

From equation (54) and the definition of  $F$ , we deduce that  $F(\xi, x)$  satisfies the following partial differential equation:

$$(1 - x^2) \frac{\partial F}{\partial x} + 2\xi \frac{\partial F}{\partial \xi} = \left[ x + \frac{1}{x} + 2\xi - 1 \right] F. \quad (57)$$

This equation together with the requirement that  $F(\xi, x) \approx x + O(x^2)$  as  $x \rightarrow 0$  (as  $G_1 = 1$ ) determines uniquely  $F(\xi, x)$ . We find:

$$F(\xi, x) = \frac{x}{(1+x)\sqrt{1-x^2}} e^{2\xi x/(x+1)} \quad (58)$$

$G_n(\xi)$  is given by the contour integral in equation (56) where the contour  $C$  encircles  $x = 0$  in such a way that  $x = 1$  and  $-1$  are outside of the contour.

Let us compute  $G_n(\xi)$  with  $\xi = N\alpha z^2$  for large  $N$ , fixed  $z$  and  $n = cN$  with fixed  $0 < c \leq 1$ . We have

$$G_n(\xi) = \oint_C \frac{dx}{2i\pi} \frac{1}{x^{n+1}} F(\xi, x) = \oint_C \frac{dx}{2i\pi} \frac{e^{N\Phi_c(x)}}{(1+x)\sqrt{1-x^2}} \quad (59)$$

where

$$\Phi_c(x) = \frac{2\alpha z^2 x}{x+1} - c \ln x \quad \text{with } c = \frac{n}{N} \quad (60)$$

is of order one for large  $N$  when  $n = cN$  with  $c$  of order one. For fixed  $c \leq 1$  and for large  $N$ , the contour integral can thus be computed using a saddle point method. The integral will be dominated by the neighbourhood of  $x^*$  such that:

$$\left. \frac{d\Phi_c}{dx} \right|_{x^*} = 0 \quad \text{i.e.} \quad \frac{2\alpha z^2}{(1+x^*)^2} = \frac{n}{Nx^*} = \frac{c}{x^*}. \quad (61)$$

There exists a real solution for  $x^*$  iff  $z^2 > 2c/\alpha$ , i.e.  $z^2 > 2n/(\alpha N)$ . We want this condition to be satisfied for all  $n < N$ , therefore we must have  $z^2 > 2/\alpha$ , i.e.  $z > \sqrt{2/\alpha}$ . Our method can only describe the regime  $z > \sqrt{2/\alpha}$ , i.e.  $y > \sqrt{2N/\alpha}$ , which corresponds to the right tail of the distribution  $P(\lambda_{\max} \leq y)$  (region where  $\lambda_{\max}$  is above its mean value). Let us call the critical point  $y_{\text{cr}} = \sqrt{2N/\alpha}$ , i.e.  $z_{\text{cr}} = \sqrt{2/\alpha}$ .

For  $y > y_{\text{cr}}$ , there are two real solutions  $x^* = -1 + (N\alpha z^2/n)[1 \pm \sqrt{1 - 2n/(N\alpha z^2)}]$ . The contour  $C$  must encircle 0 but not 1 and  $-1$ , therefore we impose  $-1 < x^* < 1$ . This implies

$$\begin{aligned} x^* &= -1 + \frac{N\alpha z^2}{n} \left[ 1 - \sqrt{1 - \frac{2n}{N\alpha z^2}} \right] = \frac{N\alpha z^2}{2n} \left[ 1 - \sqrt{1 - \frac{2n}{N\alpha z^2}} \right]^2 \\ &= -1 + \frac{\xi}{n} \left[ 1 - \sqrt{1 - \frac{2n}{\xi}} \right] = \frac{\xi}{2n} \left[ 1 - \sqrt{1 - \frac{2n}{\xi}} \right]^2. \end{aligned} \quad (62)$$

Thus

$$\Phi_c(x^*) = \alpha z^2 \left[ 1 - \sqrt{1 - \frac{2c}{\alpha z^2}} \right] - 2c \ln \left\{ \sqrt{\frac{\alpha z^2}{2c}} - \sqrt{\frac{\alpha z^2}{2c} - 1} \right\}. \quad (63)$$

The saddle point gives for large  $N$ :

$$G_n(\xi) \approx \frac{1}{2\pi} \frac{e^{N\Phi_c(x^*)}}{(1+x^*)\sqrt{1-x^{*2}}} \sqrt{\frac{2\pi}{N|d^2\Phi_c/dx^2|_{x^*}}} \quad (64)$$

where

$$\left| \frac{d^2 \Phi_c}{dx^2} \Big|_{x^*} \right| = \frac{4\sqrt{(\alpha z^2/c) - 2}}{(\sqrt{(\alpha z^2/c) - 2} - \sqrt{\alpha z^2/c})^4 \sqrt{\alpha z^2/c}}.$$

In this subsection we have found, as given in equation (64), the expression of  $G_n(\xi)$  and thus the solution  $c_n(y, \alpha) = -y^2/(2\sqrt{\pi\xi}) G_n(\xi)$  (with  $\xi = N\alpha z^2 = \alpha y^2$ ) of the recursion relation (50) for large  $N$  and  $n = cN$  with fixed  $0 < c \leq 1$ . We have also shown that the validity of our approximation is the regime  $y > y_{\text{cr}}$  with  $y_{\text{cr}} = \sqrt{2N/\alpha} = \langle \lambda_{\text{max}} \rangle$ .

### 4.3. Computation of the distribution of $\lambda_{\text{max}}$ for large $N$

We want to compute for large  $N$  and for  $y > y_{\text{cr}}$  (with the scaling  $y = z\sqrt{N}$  for large  $N$ ) the cdf  $\mathbb{P}_N(\lambda_{\text{max}} \leq y) = Z_N(y, \alpha)/Z_N(\infty, \alpha)$ . Using equation (52) and the definition of  $G_n$  in equation (53), we get:

$$\begin{aligned} \ln \mathbb{P}_N(\lambda_{\text{max}} \leq y) &= \ln Z_N(y, \alpha) - \ln Z_N(\infty, \alpha) \\ &\approx -\frac{e^{-\xi}}{2\sqrt{\pi\xi}} \left\{ 1 + 2\xi \sum_{n=1}^{N-1} \left( \frac{N-n}{n} \right) G_n(\xi) \right\}. \end{aligned} \quad (65)$$

Therefore we need to compute  $I_N(\xi) \equiv \sum_{n=1}^{N-1} ((N-n)/n) G_n(\xi)$  for large  $N$  and  $\xi = N\alpha z^2$  (with fixed  $\alpha$  and  $z$ ). For that purpose, we do not actually need to use the approximate expression of  $G_n$  for  $n$  of order  $N$  that we derived in section 4.2. We can use the formal expression of  $G_n$  as a contour integral  $G_n(\xi) = \oint_C dx / (2i\pi) (1/x^{n+1}) F(\xi, x)$  and compute the sum over  $n$  before computing the integral by the saddle point method. In particular we have:

$$\sum_{n=1}^{N-1} G_n(\xi) = \oint_C \frac{dx}{2i\pi} \frac{F(\xi, x)}{x(x-1)} + \oint_C \frac{dx}{2i\pi} \frac{F(\xi, x)}{x^N(1-x)} = \oint_C \frac{dx}{2i\pi} \frac{F(\xi, x)}{x^N(1-x)}.$$

The function  $F(\xi, x)/[x(x-1)] = (1/[(x^2-1)\sqrt{1-x^2}]) e^{2\xi x/(x+1)}$  has indeed no singularity at the origin, its integral is thus zero. On the other hand, we have:

$$\sum_{n=1}^{N-1} \frac{G_n(\xi)}{n} = \oint_C \frac{dx}{2i\pi} \left( \sum_{n=1}^{N-1} \frac{1}{nx^{n+1}} \right) F(\xi, x) = \oint_C \frac{dx}{2i\pi} \frac{F(\xi, x)}{x^N} \frac{{}_2F_1(1, 1-N, 2-N, x)}{N-1}$$

where  ${}_2F_1$  is a hypergeometric function:  ${}_2F_1(a, b, c, z) = \sum_{k \geq 0} ((a)_k (b)_k / (c)_k) z^k / k!$  with  $(a)_k = a(a+1) \cdots (a+k-1)$ . For large  $N$ , we find:

$$\begin{aligned} \frac{{}_2F_1(1, 1-N, 2-N, x)}{N-1} &= \sum_{k \geq 0} \frac{x^k}{-1-k+N} \approx \sum_{k \geq 0} x^k \left( \frac{1}{N} + \frac{1-k}{N^2} + \cdots \right) \\ &\approx \frac{1}{N(1-x)} + \frac{1}{(1-x)^2 N^2} + \cdots. \end{aligned}$$

Therefore we get for large  $N$

$$I_N(\xi) = \sum_{n=1}^{N-1} \left( \frac{N-n}{n} \right) G_n(\xi) \approx \frac{1}{N} \oint_C \frac{dx}{2i\pi} \frac{F(\xi, x)}{x^N(1-x)^2} + \cdots. \quad (66)$$

Equivalently we can write:

$$I_N(\xi) \approx \frac{1}{N} \oint_{\mathcal{C}} \frac{dx}{2i\pi} \frac{x}{(1+x)^{3/2}(1-x)^{5/2}} e^{N\Phi_1(x)} \quad (67)$$

where  $\Phi_1(x) = (2\alpha z^2 x/[x+1]) - \ln x$  (see (60)). For large  $N$ , the saddle point method thus gives

$$I_N(\xi) \approx \frac{1}{N} \frac{1}{2\pi} \frac{x^*}{(1+x^*)^{3/2}(1-x^*)^{5/2}} e^{N\Phi_1(x^*)} \sqrt{\frac{2\pi}{N|d^2\Phi_c/dx^2|_{x^*}}} \quad (68)$$

where  $x^*$  is given in equation (62) with  $n = N$ :

$$x^* = -1 + \alpha z^2 \left[ 1 - \sqrt{1 - \frac{2}{\alpha z^2}} \right] = \frac{\alpha z^2}{2} \left[ 1 - \sqrt{1 - \frac{2}{\alpha z^2}} \right]^2. \quad (69)$$

Thus we find

$$I_N(\xi) \approx \frac{1}{N^{3/2}} \frac{1}{\sqrt{2\pi}} \frac{1}{4\sqrt{\alpha z^2}(\alpha z^2 - 2)^{3/2}} e^{N\Phi_1(x^*)} \quad (70)$$

with

$$\Phi_1(x^*) = \alpha z^2 \left[ 1 - \sqrt{1 - \frac{2}{\alpha z^2}} \right] - 2 \ln \left\{ \sqrt{\frac{\alpha z^2}{2}} - \sqrt{\frac{\alpha z^2}{2} - 1} \right\}. \quad (71)$$

Therefore

$$\begin{aligned} \ln \mathbb{P}_N(\lambda_{\max} \leq y) &= \ln Z_N(y, \alpha) - \ln Z_N(\infty, \alpha) \\ &\approx -\frac{e^{-\xi}}{2\sqrt{\pi\xi}} \{1 + 2\xi I_N(\xi)\} \\ &\approx -\frac{e^{-N\alpha z^2}}{2\sqrt{\pi N\alpha z^2}} \left\{ 1 + \frac{\sqrt{\alpha z^2}}{\sqrt{N}2\sqrt{2\pi}(\alpha z^2 - 2)^{3/2}} e^{N\Phi_1(x^*)} \right\}. \end{aligned}$$

As  $\Phi_1(x^*) > 0$  for  $z > z_{\text{cr}}$ , i.e.  $\alpha z^2 > 2$ , the first term in the parentheses can be neglected for large  $N$ :

$$\ln \mathbb{P}_N(\lambda_{\max} \leq y) \approx -\frac{e^{-N\alpha z^2}}{4\pi N} \frac{1}{\sqrt{2}(\alpha z^2 - 2)^{3/2}} e^{N\Phi_1(x^*)} \quad (72)$$

with  $y = z\sqrt{N}$ . Therefore we get the expression of the right tail of the cdf of  $\lambda_{\max}$ :

$$\ln \mathbb{P}_N(\lambda_{\max} \leq z\sqrt{N}) \approx -\frac{1}{4\pi N\sqrt{2}(\alpha z^2 - 2)^{3/2}} e^{-2N\psi_+(z)} \quad \text{for } z > \sqrt{\frac{2}{\alpha}} \quad (73)$$

where the rate function  $\psi_+(z) = (\alpha z^2 - \Phi_1(x^*))/2$  is given by:

$$\psi_+(z) = \frac{\alpha z^2}{2} \left[ \sqrt{1 - \frac{2}{\alpha z^2}} \right] + \ln \left\{ \sqrt{\frac{\alpha z^2}{2}} - \sqrt{\frac{\alpha z^2}{2} - 1} \right\}. \quad (74)$$

We have thus found

$$\mathbb{P}_N(\lambda_{\max} \leq z\sqrt{N}) \approx 1 - \frac{1}{4\pi N\sqrt{2}(\alpha z^2 - 2)^{3/2}} e^{-2N\psi_+(z)} \quad \text{for } z > \sqrt{\frac{2}{\alpha}}. \quad (75)$$

Deriving with respect to  $t = z\sqrt{N}$  we get an equivalent of the probability density function of  $\lambda_{\max}$  for large  $N$ :

$$\mathcal{P}(\lambda_{\max} = t) \approx \frac{\sqrt{\alpha N}}{2\pi\sqrt{2}(\alpha t^2 - 2N)} e^{-2N\psi_+(t/\sqrt{N})} \quad \text{for } t > \sqrt{\frac{2N}{\alpha}},$$

$$\left| t - \sqrt{\frac{2N}{\alpha}} \right| \approx O(\sqrt{N}). \quad (76)$$

We thus recover the dominant order for large  $N$  (given by  $\psi_+(z)$ ) that was derived in [35] by a Coulomb gas method, but in addition this method also provides us with the first correction term to the dominant order.

Let us now see how this precise right tail large deviation behaviour in (75) behaves when  $z \rightarrow \sqrt{2/\alpha}$  from the right. Using the leading expansion of  $\psi_+(z)$  around  $z = \sqrt{2}$  in (17) and setting  $y = \sqrt{2N/\alpha} + N^{-1/6}(x/\sqrt{2\alpha})$ , i.e.,  $z = y/\sqrt{N} = \sqrt{2/\alpha} + N^{-2/3}(x/\sqrt{2\alpha})$  one finds from (75)

$$\mathbb{P}_N(\lambda_{\max} \leq y) \approx 1 - \frac{1}{16\pi x^{3/2}} e^{-(4/3)x^{3/2}}. \quad (77)$$

On the other hand, using the boundary condition  $q(z) \approx \text{Ai}(z) \approx (1/[2\sqrt{\pi}z^{1/4}])e^{-(2/3)z^{3/2}}$  as  $z \rightarrow \infty$  in the definition of the Tracy–Widom function (8), one can easily derive the precise leading asymptotics of its right tail,  $F_2(x) \approx 1 - (1/[16\pi x^{3/2}])e^{-(4/3)x^{3/2}}$  as  $x \rightarrow \infty$ . Thus, our right large deviation function for small argument (when the fluctuation of  $\lambda_{\max}$  to the right of its mean value  $\sqrt{2N/\alpha}$  is of  $O(N^{-1/6})$ ) in (77), matches smoothly with the precise right tail of the Tracy–Widom distribution  $F_2(x)$ .

## 5. Double scaling limit and Tracy–Widom distribution

In this section, we provide an elementary derivation of the Tracy–Widom law for the GUE based on simple scaling analysis of the recursion relations derived in section 2 in the vicinity of the critical point  $y = y_{\text{cr}} = \sqrt{2N/\alpha}$ . This derivation, in our opinion, is mathematically simpler than the original derivation by Tracy and Widom [6, 7] as it avoids the sophisticated asymptotic analysis of Fredholm determinants. The derivation of the Painlevé II equation from the scaling analysis of recursion relations that we follow here is similar in spirit (though rather different in details) to the analysis of the partition function in the two-dimensional Yang–Mills theory on a sphere by Gross and Matytsin [39].

Let us recall that the Tracy–Widom distribution  $F_2(x)$  is defined as

$$F_2(x) = \exp \left\{ - \int_x^\infty ds (s - x)q^2(s) \right\} \quad (78)$$

where  $q(x)$  satisfies the Painlevé II equation with the boundary condition

$$q''(x) = 2q^3(x) + xq(x) \quad \text{Painlevé II}$$

$$q(x) \approx \frac{1}{2\sqrt{\pi}x^{1/4}} e^{-(2/3)x^{3/2}} \quad \text{as } x \rightarrow \infty. \quad (79)$$

From equation (78), it follows that  $d^2 \ln F_2(x)/dx^2 = -q^2(x)$ .



We want to show that for large  $N$  the probability of small typical fluctuations of  $\lambda_{\max}$  around its mean value  $\sqrt{2N/\alpha}$  are described by the Tracy–Widom distribution. For this, we need to first estimate how do these typical fluctuations scale with  $N$  for large  $N$ . In the vicinity of the mean  $\sqrt{2N/\alpha}$ , let us then write

$$\lambda_{\max} \approx \sqrt{\frac{2N}{\alpha}} + \frac{1}{\sqrt{2\alpha}} N^\gamma x \tag{80}$$

where  $N^\gamma$  is the scale of the typical fluctuation and the random variable  $x$  has an  $N$  independent distribution for large  $N$ . Evidently the exponent  $\gamma < 1/2$  (so that the fluctuation is less than the mean) whose precise value is yet to be determined. Note also that since  $\lambda_{\max}$  always appears in the distribution  $\mathbb{P}_N(\lambda_{\max} \leq y)$  in the scaling combination  $y\sqrt{\alpha}$  (see equation (30)), we have chosen the prefactor of the fluctuation term as  $1/\sqrt{2\alpha}$ , which then ensures that the random variable  $x$  describing the typical fluctuation is also independent of  $\alpha$ .

One way to estimate the exponent  $\gamma$  is from the right large deviation tail computed in (75) in section 4. The right tail in (75) describes the probability of large fluctuations of  $O(\sqrt{N})$  to the right of the mean. Assuming that the right tail behaviour continues to hold even for fluctuations smaller than  $\sqrt{N}$ , we substitute  $z\sqrt{N} = \sqrt{(2N/\alpha)} + (1/\sqrt{2\alpha})N^\gamma x$  in (75). This gives

$$\mathbb{P}_N \left( \lambda_{\max} \leq \sqrt{\frac{2N}{\alpha}} + \frac{1}{\sqrt{2\alpha}} N^\gamma x \right) \approx 1 - \frac{1}{N^{(1/4)+(3\gamma/2)}} \frac{1}{16\pi x^{3/2}} e^{-(4/3)x^{3/2}N^{(3\gamma/2)+1/4}} \tag{81}$$

valid for  $x > 0$ ,  $x$  large. Assuming that this continues to hold even for not so large  $x$  (so that it even captures the tail of the distribution of typical small fluctuations), we expect that in terms of this rescaled variable  $x$ , the tail of the distribution in (81) is independent of  $N$  for large  $N$ . Clearly, for this to happen the power of  $N$  must be zero both inside the exponential as well as in the prefactor in (81), indicating that  $(1/4) + (3\gamma/2) = 0$ , thus  $\gamma = -1/6$ . Hence, the correct scaling describing typical fluctuations, for large  $N$ , is given by

$$\lambda_{\max} = \sqrt{\frac{2N}{\alpha}} + \frac{1}{\sqrt{2\alpha}} N^{-1/6} x \tag{82}$$

where  $x$  has an  $N$  independent distribution that we now have to compute and show that it is given by the Tracy–Widom function  $F_2(x)$ .

The meaning of the double scaling limit is now clear. It simply says the following. Consider the cdf  $\mathbb{P}_N(\lambda_{\max} \leq y)$  or rather its logarithm (for convenience)  $\ln \mathbb{P}_N(\lambda_{\max} \leq y)$ . In general, it is a function of two variables  $y$  and  $N$ . However, in the vicinity of the mean  $y \rightarrow \sqrt{2N/\alpha}$ , if one takes the limit  $y - \sqrt{2N/\alpha} \rightarrow 0$  and  $N \rightarrow \infty$ , but keeping the scaling combination  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$  fixed, this function of two variables collapses into a function of the single scaled variable  $x$

$$\ln \mathbb{P}_N(\lambda_{\max} \leq y) \rightarrow f \left( \sqrt{2\alpha} N^{1/6} \left( y - \sqrt{\frac{2N}{\alpha}} \right) \right) \tag{83}$$

and our job is to show that this scaling function  $f(x) = \ln F_2(x)$ , where  $F_2(x)$  is the Tracy–Widom function defined in (78). In other words, we want to show that  $f''(x) = -q^2(x)$  where  $q(x)$  satisfies the Painlevé II equation (79).



Our starting point is the definition of the cdf in (28). From equation (36) it is easy to see that the partition function  $Z_N$  satisfies the recursion

$$\frac{Z_{N-1}(y, \alpha) Z_{N+1}(y, \alpha)}{Z_N^2(y, \alpha)} = \frac{h_N(y, \alpha)}{h_{N-1}(y, \alpha)} = R_N(y, \alpha). \quad (84)$$

Taking the logarithm and using the definition in (28) we get

$$\ln \mathbb{P}_{N+1}(\lambda_{\max} \leq y) + \ln \mathbb{P}_{N-1}(\lambda_{\max} \leq y) - 2 \ln \mathbb{P}_N(\lambda_{\max} \leq y) = \ln \left( \frac{R_N(y, \alpha)}{R_N(\infty, \alpha)} \right). \quad (85)$$

In the double scaling limit, we will now substitute the anticipated scaling form in (83) for the logarithm of the cdf on the left-hand side of (85). But we need to first evaluate  $\ln \mathbb{P}_{N\pm 1}(\lambda_{\max} \leq y)$ . Replacing  $N$  by  $N \pm 1$  in (83) and expanding for large  $N$ , with  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$  fixed, we get

$$\begin{aligned} \ln \mathbb{P}_{N\pm 1}(\lambda_{\max} \leq y) &= f \left( \sqrt{2\alpha} (N \pm 1)^{1/6} \left( y - \sqrt{\frac{2(N \pm 1)}{\alpha}} \right) \right) \\ &= f \left( x \mp N^{-1/3} \pm \frac{x}{6N} \pm \frac{N^{-4/3}}{12} + \dots \right) \\ &= f(x) \mp N^{-1/3} f'(x) + \frac{N^{-2/3}}{2} f''(x) + O(N^{-1}). \end{aligned} \quad (86)$$

Substituting this result in (85) we get for the left-hand side

$$\begin{aligned} \ln \mathbb{P}_{N+1}(\lambda_{\max} \leq y) + \ln \mathbb{P}_{N-1}(\lambda_{\max} \leq y) - 2 \ln \mathbb{P}_N(\lambda_{\max} \leq y) \\ \approx N^{-2/3} f''(x) + O(N^{-1}). \end{aligned} \quad (87)$$

From equations (85) and (87), we get for large  $N$

$$N^{-2/3} f''(x) \approx \ln \left( \frac{R_N(y, \alpha)}{R_N(\infty, \alpha)} \right) \approx \ln \left( \frac{R_N(y, \alpha)}{N/(2\alpha)} \right) \quad (88)$$

as  $R_N(\infty, \alpha) = N/(2\alpha)$  (see equation (44)). This suggests that in this scaling limit,  $R_N$  must scale as  $R_N(y, \alpha) \approx (N/2\alpha)(1 + N^{-2/3} f''(x) + \dots)$ . More precisely, this leads us to the following large  $N$  expansion of  $R_N(y, \alpha)$  in the double scaling limit

$$R_N(y, \alpha) \approx \frac{N}{2\alpha} (1 + N^{-2/3} r_1(x) + N^{-1} r_2(x) + N^{-4/3} r_3(x) + \dots), \quad (89)$$

where

$$f''(x) = r_1(x) \quad (90)$$

and  $r_2(x)$ ,  $r_3(x)$  etc describing the higher order scaling corrections. Thus, if we can now determine the first subleading scaling function  $r_1(x)$  in the expansion of  $R_N(y, \alpha)$ , then we can determine  $f(x)$  by integrating  $r_1(x)$  twice. So, our next task is to determine  $r_1(x)$  by analysing the recursion relations (41) and (42) (setting  $n = N$ ) in the double scaling limit.

We now know, from (89), how  $R_N(y, \alpha)$  behaves in the scaling limit with the scaling combination  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$  fixed. In order to analyse the recursion relations (41) and (42), we also need to know how  $S_N(y, \alpha)$  behaves in this scaling limit.

In order to match the leading  $N$  behaviour of  $R_N(y, \alpha)$  with  $x$  fixed in (42), it is not difficult to see that to leading order for large  $N$ ,  $S_N(y, \alpha)$  must have the following scaling behaviour

$$S_N(y, \alpha) \approx \frac{N^{-1/6}}{\sqrt{2\alpha}} s_1 \left( \sqrt{2\alpha} N^{1/6} \left( y - \sqrt{\frac{2N}{\alpha}} \right) \right) + O(N^{-1/2}) \quad (91)$$

where  $s_1(x)$  is the leading order scaling function. Let us first evaluate the difference  $S_{N-1}(y, \alpha) - S_N(y, \alpha)$  that appears in (42). Replacing  $N$  by  $N - 1$  in (91), expanding for large  $N$ , we get

$$S_{N-1}(y, \alpha) - S_N(y, \alpha) \approx \frac{N^{-1/2}}{\sqrt{2\alpha}} s_1'(x) + O(N^{-5/6}). \quad (92)$$

It rests to evaluate the partial derivative  $\partial \ln R_N(y, \alpha) / \partial y$  in (42). From the definition of the scaling variable  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$ , it follows, using the chain rule,

$$\begin{aligned} \frac{\partial \ln R_N(y, \alpha)}{\partial y} &= \frac{\partial \ln R_N(y, \alpha)}{\partial x} \frac{\partial x}{\partial y} \\ &= \sqrt{2\alpha} N^{1/6} \frac{\partial \ln R_N(y, \alpha)}{\partial x} \\ &= \sqrt{2\alpha} N^{-1/2} r_1'(x) + O(N^{-5/6}). \end{aligned} \quad (93)$$

Finally substituting (92) and (93) in (42) (with  $n = N$ ) we get

$$\begin{aligned} \sqrt{2\alpha} (r_1'(x) N^{-1/2} + O(N^{-5/6})) &\approx \frac{\partial \ln R_N}{\partial y} = 2\alpha (S_{N-1} - S_N) \\ &\approx \sqrt{2\alpha} (N^{-1/2} s_1'(x) + O(N^{-5/6})). \end{aligned}$$

Matching the leading order  $N^{-1/2}$  term gives a relation between  $s_1(x)$  and  $r_1(x)$ :  $s_1'(x) = r_1'(x)$ , i.e.,  $s_1(x) = r_1(x) + c_0$  with  $c_0$  a constant. From (44) and the fact that when  $y \rightarrow \infty$ ,  $x \rightarrow \infty$ , it follows that both the scaling functions  $r_1(x)$  and  $s_1(x)$  must vanish as  $x \rightarrow \infty$ . Thus the constant  $c_0 = 0$  and we have, for all  $x$ ,

$$r_1(x) = s_1(x). \quad (94)$$

Having determined the relation  $r_1(x) = s_1(x)$ , we need one more relation between these two scaling functions in order to determine them individually. This will now be done by substituting the scaling solutions for  $R_N(y, \alpha)$  (given in (89)) and  $S_N(y, \alpha)$  (given in (91)) into the remaining recursion relation (41).

To analyse (41) (setting  $n = N$ ), we need to evaluate the derivative  $\partial \ln R_N(y, \alpha) / \partial \alpha$ . From the definition of the scaling variable,  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$ , it follows that  $\partial x / \partial \alpha = x / (2\alpha) + (N^{2/3} / \alpha)$ . We then use the chain rule and (89) to express

$$\frac{\partial \ln R_N(y, \alpha)}{\partial \alpha} = \frac{r_1'(x) - 1}{\alpha} + \frac{N^{-1/3}}{\alpha} r_2'(x) + \frac{N^{-2/3}}{2\alpha} [x r_1'(x) - 2r_1(x) r_1'(x) + 2r_3'(x)] + \dots$$

Again replacing  $N$  by  $N \pm 1$  in (89) and expanding for large  $N$ , keeping  $x = \sqrt{2\alpha} N^{1/6} (y - \sqrt{2N/\alpha})$  fixed, we get

$$R_{N-1} - R_{N+1} \approx \frac{r_1'(x) - 1}{\alpha} + \frac{N^{-1/3}}{\alpha} r_2'(x) + \frac{N^{-2/3}}{6\alpha} [-2r_1(x) - xr_1'(x) + 6r_3'(x) + r_1'''(x)] + \dots$$

and similarly from (91)

$$S_{N-1}^2 - S_N^2 \approx \frac{N^{-2/3}}{\alpha} s_1(x) s_1'(x) + \dots$$

Substituting these results in (41) and matching the leading order term ( $O(N^{-2/3})$ ), we get the desired second relation between  $r_1(x)$  and  $s_1(x)$

$$xr_1'(x) - 2r_1(x)r_1'(x) = -\frac{2}{3}r_1(x) - \frac{x}{3}r_1'(x) + \frac{1}{3}r_1'''(x) + 2s_1(x)s_1'(x).$$

Eliminating  $s_1(x)$  by using  $s_1(x) = r_1(x)$  we finally get a single closed equation for  $r_1(x)$

$$2xr_1'(x) + r_1(x) = \frac{1}{2}r_1'''(x) + 6r_1(x)r_1'(x). \tag{95}$$

Let us write

$$r_1(x) = -u^2(x). \tag{96}$$

Equation (95) then becomes an equation for  $u(x)$ :

$$u(u''' - 6u^2u' - xu' - u) = -3u'(u'' - xu - 2u^3). \tag{97}$$

Let  $W(x) = u''(x) - xu(x) - 2u^3(x)$ . Then (97) becomes

$$u(x) \frac{dW(x)}{dx} = -3u'(x)W(x) \tag{98}$$

which can simply be integrated to give

$$W(x) = \frac{A}{u(x)^3} \tag{99}$$

where  $A$  is an arbitrary constant. Hence we have

$$u''(x) - xu(x) - 2u^3(x) = \frac{A}{u(x)^3}. \tag{100}$$

From the boundary condition  $r_1(x) \rightarrow 0$  as  $x \rightarrow \infty$  (which follows from (44)), it follows using  $r_1(x) = u^2(x)$  that  $u(x) \rightarrow 0$  as  $x \rightarrow \infty$ . Taking  $x \rightarrow \infty$  in (100) then fixes the value of the constant  $A = 0$ . Finally, from (90), we have  $f''(x) = r_1(x) = -u^2(x)$ , where  $u(x)$  satisfies the Painlevé II equation

$$u''(x) = xu(x) + 2u^3(x). \tag{101}$$

To fix the boundary condition for  $u(x)$ , we again invoke the matching with the right large deviation tail in (81). Taking logarithm of (81) with  $\gamma = -1/6$  and using

$\ln \mathbb{P}_N(\lambda_{\max} \leq y, \alpha) = f(x)$  we find that

$$f(x) \approx -\frac{1}{16\pi x^{3/2}} e^{-(4/3)x^{3/2}} \quad \text{as } x \rightarrow \infty. \quad (102)$$

Hence  $u^2(x) = -f''(x) \approx e^{-(4/3)x^{3/2}} / (4\pi\sqrt{x})$  and consequently as  $x \rightarrow \infty$

$$u(x) = \sqrt{-f''(x)} \approx e^{-(2/3)x^{3/2}} \frac{1}{2\sqrt{\pi}x^{1/4}}. \quad (103)$$

Finally integrating  $f''(x)$  twice and using the appropriate boundary condition as  $x \rightarrow \infty$ , we get

$$f(x) = -\int_x^\infty ds (s-x)u^2(s) \quad (104)$$

where  $u(x)$  satisfies the Painlevé II equation (101) with the boundary condition (103). Comparing to (78), we have thus shown that the scaling function  $f(x) = \ln F_2(x)$ , where  $F_2(x)$  is the Tracy–Widom function ( $\beta = 2$ ). This then constitutes our derivation for the Tracy–Widom distribution for the GUE ( $\beta = 2$ ).

Using recursion relations that we have derived for orthogonal polynomials on a semi-infinite interval, we have shown that the large  $N$  asymptotics of the distribution of the maximal eigenvalue of a Gaussian random matrix (from the GUE) is described in the double scaling regime by the Painlevé II equation. Similar recursion relations for other orthogonal polynomials leading to different kinds of Painlevé equations have also been established in a number of papers, see [49] and references therein, in particular [50, 46, 47].

## 6. Conclusion

In this paper, we have provided a rather simple and pedestrian derivation of the Tracy–Widom law for the distribution of the largest eigenvalue of a Gaussian unitary random matrix. This was done by suitably adapting a method of orthogonal polynomials developed by Gross and Matytsin [39] in the context of two-dimensional Yang–Mills theory. Our derivation requires just elementary asymptotic scaling analysis of a pair of coupled nonlinear recursion relations. Strictly in the  $N \rightarrow \infty$  limit, there is a third order phase transition in the form of the probability distribution of  $\lambda_{\max}$  as  $\lambda_{\max}$  crosses its mean value from left to right. For finite but large  $N$ , the two regions are connected by a smooth crossover function and the shape of this crossover function is precisely the Tracy–Widom distribution that describes the ‘typical’ small fluctuations of  $\lambda_{\max}$  around its mean. The ‘atypical’ large fluctuations to the left and right of the mean are described by large deviation tails that correspond to the two ‘phases’ across this phase transition. In qualitative analogy to the two-dimensional Yang–Mills theory, the left (left large deviation) and the right (right large deviation) phases correspond respectively to the ‘strong’ and ‘weak’ coupling phases of the two-dimensional QCD. Apart from the simple derivation of the Tracy–Widom GUE law, we were also able to compute the precise right large deviation tail of the maximal eigenvalue distribution that is not described by the Tracy–Widom distribution. In the language of QCD, these right tail corrections are similar to the non-perturbative (in  $1/N$  expansion) corrections to the QCD partition function in 2d [39].

One drawback of our method is that it works only for the GUE case (with Dyson index  $\beta = 2$ ). It would be challenging to see if this method can be extended/generalized to derive the Tracy–Widom law for the other two Gaussian ensembles, namely the GOE ( $\beta = 1$ ) and the GSE ( $\beta = 4$ ). In addition, this method for  $\beta = 2$  should be useful to compute the largest eigenvalue distribution for other non-Gaussian matrix ensembles, such as the Laguerre (Wishart matrices) or the Jacobi ensembles.

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