

Singular Values for Products of Complex Ginibre Matrices with a Source: Hard Edge Limit and Phase Transition

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Abstract: The singular values squared of the random matrix product $Y = G_r G_{r-1} \cdots G_1(G_0 + A)$, where each G_j is a rectangular standard complex Gaussian matrix while A is non-random, are shown to be a determinantal point process with the correlation kernel given by a double contour integral. When all but finitely many eigenvalues of A^*A are equal to bN, the kernel is shown to admit a well-defined hard edge scaling, in which case a critical value is established and a phase transition phenomenon is observed. More specifically, the limiting kernel in the subcritical regime of 0 < b < 1 is independent of b, and is in fact the same as that known for the case b = 0 due to Kuijlaars and Zhang. The critical regime of b = 1 allows for a double scaling limit by choosing $b = (1 - \tau/\sqrt{N})^{-1}$, and for this the critical kernel and outlier phenomenon are established. In the simplest case r = 0, which is closely related to non-intersecting squared Bessel paths, a distribution corresponding to the finite shifted mean LUE is proven to be the scaling limit in the supercritical regime of b > 1 with two distinct scaling rates. Similar results also hold true for the random matrix product $T_r T_{r-1} \cdots T_1(G_0 + A)$, with each T_j being a truncated unitary matrix.

1. Introduction and Main Results

1.1. Introduction. The squared singular values of a matrix X are equal to the eigenvalues of the positive semi-definite Hermitian matrix X^*X , where X^* denotes the Hermitian conjugate of X. An ensemble of random matrices of the form X^*X may then contain x = 0 as the left boundary of support of the eigenvalues. Since the eigenvalue density is strictly zero for x < 0, x = 0 is then called a hard edge (see e.g. [27, Ch. 7]). As an explicit example, consider the ensemble of $n \times N$ ($n \ge N$) rectangular standard complex Gaussian random matrices, namely the joint density of elements being proportional to $\exp\{-tr(X^*X)\}$, and let X be a matrix from this ensemble. Let $\{\lambda_j\}$ denote the eigenvalues of the scaled positive semi-definite matrix $N^{-1}X^*X$. In the limit $N \to \infty$ with n - N fixed, the density of $\{\lambda_i\}$ has support [0, 4]. That the support is a finite interval gives rise to this particular scaling being referred to as global scaling, and the corresponding density as the global density. The explicit functional form of the global density is given by the so-called Marchenko–Pastur law (see e.g. [54])

$$\rho_{(1)}^{\text{MP}}(\lambda) = \frac{1}{2\pi} \sqrt{\frac{4-\lambda}{\lambda}}, \qquad 0 < \lambda \le 4.$$
(1.1)

Note in particular the reciprocal square root singularity as the hard edge $\lambda = 0$ is approached from above, in contrast to the square root singularity as $\lambda \to 4^-$. The point $\lambda = 4$ is an example of what is termed a soft edge, since for finite N the eigenvalue density is not strictly zero for $\lambda > 4$.

Continuing with this example, for large N the eigenvalues in the neighbourhood of the hard edge have spacing $\mathcal{O}(1)$ upon the introduction of the scaled variables $X_j = 4N^2\lambda_j$ (j = 1, ..., N) (see e.g. [27, Sect. 7.2.1]). This will be referred to as hard edge scaling. Moreover, in the limit $N \to \infty$, and with $v_0 = n - N$, the limiting state—referred to as the hard edge state—is an example of a determinantal point process, meaning that the *k*-point correlation function can be written in the form

$$\rho_{(k)}(X_1, \dots, X_k) = \det[K^{\mathsf{h}}(X_j, X_l)]_{j,l=1,\dots,k}$$
(1.2)

with correlation kernel (see e.g. [27, Exercises 7.2 q.1])

$$K^{\rm h}(x, y) = \frac{1}{4} \int_0^1 J_{\nu_0}(\sqrt{xt}) J_{\nu_0}(\sqrt{yt}) dt, \qquad (1.3)$$

where $J_{\nu_0}(x)$ is the Bessel function of the first kind of order ν_0 .

Our interest in this paper is in the functional form and analytic properties of the correlation kernel for the hard edge scaling of the squared singular values of the product of independent random matrices

$$Y = G_r G_{r-1} \cdots G_1 (G_0 + A), \tag{1.4}$$

where each G_j is an $(N + v_j) \times (N + v_{j-1})$ standard complex Gaussian matrix (also referred to as the complex Ginibre matrices since such non-Hermitian random matrices in the square case were first studied by Ginibre [34]) with $v_{-1} = 0$ and integers $v_0, \ldots, v_r \ge 0$, while *A* is of size $(N+v_0) \times N$ and fixed. If we focus on the singular values of *Y*, the definition (1.4) can equivalently be written as a product of independent square matrices G_j now with each being distributed according to the joint density of elements proportional to $\det^{v_j}(G_j^*G_j) \exp\{-\operatorname{tr}(G_j^*G_j)\}$; see e.g. [3,45]. In this case the restriction on the parameters can be relaxed to $v_0, v_1, \ldots, v_r > -1$, and the main results in the present paper (for instance, Proposition 1.1, Theorems 1.2, 1.3, 3.1 and 3.2) also hold true.

In the case that all entries of A are zero, the determinantal representation of the joint eigenvalue density and the limiting hard edge state have been the subject of a number of recent works [3,4,29,45,46,57]. For r nonzero, the product (1.4) is the simplest nontrivial example of the more general product $(G_r + A_r) \cdots (G_1 + A_1)(G_0 + A_0)$, where each G_j is random and each A_j is fixed. However for the latter product with $r \ge 1$, it is not known how to find a closed form of the joint eigenvalue density, which is the starting point of our study (see Proposition 1.1 below), let alone to study asymptotic statistical properties.

The study of products of random matrices goes back to the pioneering work of Furstenberg and Kesten [33] in the context of dynamical systems and their Lyapunov

exponents. Later, applications were found in Schrödinger operator theory [17], in statistical physics relating to disordered and chaotic dynamical systems [20], in wireless communication networks [59] and in combinatorics [56]. Further motivations from theoretical physics include the DMPK equation in mesoscopic quantum transport [11,25,51], fluid turbulence [26, eqn. (16)] and time evolution models [50, eqn. (1)] (the last two models can be immediately recognized after discrete integration in time).

Products of complex Ginibre matrices, and of truncations of Haar distributed random unitary matrices, have attracted much attention as examples of determinatal point processes with kernels possessing special integrability properties. The first advance in this direction was by Akemann and coworkers [3,4], who derived the joint eigenvalue density and corresponding correlation kernel in terms of biorthogonal functions for the product of complex Ginibre matrices; a double integral formula for the correlation kernel was subsequently obtained by Kuijlaars and Zhang, cf. [46, Prop. 5.1]. These advances have opened up the possibility to study local statistical properties, see [29,40,45,46] for the hard edge limit and [48] for bulk and soft edge limits. All these studies form part of a fast paced and very recent literature relating to the integrability and exactly solvable properties of random matrix products. Works relating to this theme which have appeared on the electronic preprint archive over the past few months (as of July 2015) include [19,32,35,39–41,48,61]; we refer the reader to [2] for a recent survey article. Here we contribute to this line of research by undertaking a comprehensive study of the hard edge state formed by the singular values of (1.4).

The first point to note is that the hard edge state in the case A = 0 depends on r, and thus is no longer described by the correlation kernel (1.3). This fact can be anticipated by an analysis of the global density of the squared singular values [5,31,53,56]. The global density, which refers to the limiting density of eigenvalues of $N^{-r-1}Y^*Y$ as $N \to \infty$, is found to exhibit the hard edge singularity (see [56] or [31, eqn (2.16)])

$$\frac{1}{\pi}\sin\frac{\pi}{r+2}\lambda^{-1+\frac{1}{r+2}} \quad \text{as} \ \lambda \to 0^+,$$
(1.5)

which has an *r*-dependent exponent. In fact with the eigenvalues of Y^*Y scaled according to $X_j = Nx_j$ (j = 1, ..., N), as $N \to \infty$ the hard edge state in the case A = 0 forms a determinantal point process with limiting correlation kernel

$$K^{\mathbf{h},r}(x,y) = \frac{1}{(2\pi i)^2} \int_{-1/2-i\infty}^{-1/2+i\infty} du \oint_{\Sigma} dt \prod_{j=-1}^{r} \frac{\Gamma(\nu_j + u + 1)}{\Gamma(\nu_j + t + 1)} \frac{\sin \pi u}{\sin \pi t} \frac{x^t y^{-u-1}}{u - t}$$
$$= \int_{0}^{1} G_{0,r+2}^{1,0} \Big(\underbrace{-\nu_0, -\nu_1, \dots, -\nu_r}_{0,r+1} \Big| ux \Big) G_{0,r+2}^{r+1,0} \Big(\underbrace{-\nu_0, \nu_1, \dots, \nu_r, 0}_{(1.6)} \Big| uy \Big) du,$$
(1.6)

where $G_{p,q}^{m,n}$ denotes the Meijer G-function defined by the contour integral

$$G_{p,q}^{m,n}\binom{a_1,\ldots,a_p}{b_1,\ldots,b_q}|z) = \frac{1}{2\pi i} \int_{\gamma} \frac{\prod_{j=1}^m \Gamma(b_j+s) \prod_{j=1}^n \Gamma(1-a_j-s)}{\prod_{j=m+1}^q \Gamma(1-b_j-s) \prod_{j=n+1}^p \Gamma(a_j+s)} z^{-s} ds,$$
(1.7)

see [49, Sect. 5.2] for the choice of the contour γ and elementary properties of G-functions, or [8] for a gentle introduction; it is worth mentioning a particular relation

between the generalized hypergeometric function ${}_{p}F_{q}$ and the Meijer G-function (cf. eqn (14), [49, Sect. 5.2])

$${}_{p}F_{q}(a_{1},\ldots,a_{p};b_{1},\ldots,b_{q};z) = \frac{\prod_{l=1}^{q}\Gamma(b_{l}+1)}{\prod_{l=1}^{p}\Gamma(a_{l}+1)}G_{p,q+1}^{1,p} {\binom{1-a_{1},\ldots,1-a_{p}}{0,1-b_{1},\ldots,1-b_{q}}} - z),$$
(1.8)

which thus gives an alternative way of writing the first Meijer G-function in the final line of (1.6). These kernels were described in [46] and are named after Meijer G-kernels in [45]. They also appear in the hard edge scaling for products with inverses of Ginibre matrices [29], products of truncated unitary matrices [45], Cauchy two matrix models [12,13,30], and Muttalib–Borodin biorthogonal ensembles [15,52] (cf. [45] for the relationship between Borodin's expression and Meijer G-kernels).

As noted in [46, Sect. 5.3], in the case r = 0 the facts that

$$G_{0,2}^{1,0}\left(-\frac{1}{\nu,\nu}\right) = (ux)^{-\nu/2} J_{\nu}(2\sqrt{ux}), \quad G_{0,2}^{1,0}\left(-\frac{1}{\nu,0}\right) = (uy)^{\nu/2} J_{\nu}(2\sqrt{uy}),$$
(1.9)

show

$$K^{h,0}(x, y) = 4(y/x)^{\nu/2} K^{h}(4x, 4y).$$
(1.10)

The factor of $(y/x)^{\nu/2}$ cancels out of the determinant (1.2), while the factors of 4 are accounted for by this same factor being present in the scaling leading to (1.3); recall the text leading to this equation.

Consider now (1.4) with

$$A = \sqrt{bN} I_{(N+\nu_0) \times N},\tag{1.11}$$

where $I_{(N+\nu_0)\times N}$ denotes the $(N + \nu_0) \times N$ rectangular matrix with 1's on the main diagonal, and 0's elsewhere. It was shown recently in [31, Remark 3.4] that there is a critical value of b = 1 for which as $N \to \infty$ the left hand edge of the support of the global scaled squared singular values equals 0 for the last time as *b* increases from 0. Moreover, it was shown that the singularity of the global density has the leading form

$$\frac{1}{\pi}\sin\frac{2\pi}{2r+3}\,\lambda^{-1+\frac{2}{2r+3}} \quad \text{as} \quad \lambda \to 0^+, \tag{1.12}$$

which gives rise to a different family of exponents to those in (1.5). We remark that the fractional part of the exponents, 1/(r + 2) and 1/(r + 3/2) respectively in (1.5) and (1.12), are the reciprocals of positive integers and half-integers, which given knowledge of the correlation kernel (1.6) and its analogue in relation to (1.12) to be established herein (see Eq. (1.24) below), is coincident with them being the simplest in terms of tractability of the general rational fractional exponents accessible in the Raney family (see e.g. [31, eqn. (2.16)]), so named due to the sequence formed by the moments of the global density.

Let *A* be again given by (1.11), and consider the case r = 0 in (1.4) so that $Y = G_0 + \sqrt{bN}I_{(N+\nu_0)\times N}$. It is well known that the squared singular values allow for an interpretation as the positions of non-intersecting particles on the half line evolving according to the squared Bessel process with parameter $d = 2(\nu_0 + 1)$ (see e.g. [37,38]). In this interpretation the particles all begin at the same point *bN*, evolve for t = 1 time

units, and furthermore are conditioned to remain non-intersecting if the process was to continue to $t \to \infty$. The support of the density of such a process with a delta function initial condition $N\delta(x - a)$ is, for Nt < a equal at leading order to [-Nt + a, Nt + a] (this fact is implied by results in [43], for example), so we see that with a = bN and t = 1, the particles first come in contact with the wall x = 0 as b is decreased to b = 1. A functional form of the hard edge scaled kernel in the critical case b = 1, generalised to a double scaling by setting $b = (1 - \tau/\sqrt{N})^{-1}$, has recently been obtained in [44]. In the present paper an alternative functional form to that in [44] is derived; see Eq. (1.25) below. The kernel (1.25), further specialised to $v_0 = -1/2$ reads

$$\frac{1}{2\pi^2 i} \left(\frac{1}{\xi\eta}\right)^{1/4} \int_0^\infty du \int_{i\mathbb{R}} dv \, \frac{e^{-\tau u + \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v} \frac{\cos(2\sqrt{u\xi})}{u^{1/2}} \cos(2\sqrt{v\eta}). \quad (1.13)$$

With ξ and η replaced by squared variables, (1.13) is identified in [44] as the symmetric Pearcey kernel found in the study [16]. Moreover, our method of derivation of this new functional form in the case r = 0 works equally as well for the double scaling of the critical kernel in the general r case, which is our main theme. The resulting explicit double contour integral expression is given in Theorem 1.2 below.

1.2. Main results. In preparation for the statement of our first key result, let us introduce two auxiliary functions. The first is defined to be

$$\Psi(u;x) = \frac{1}{(2\pi i)^r} \frac{1}{\Gamma(\nu_0+1)} \int_{\gamma_1} dw_1 \cdots \int_{\gamma_r} dw_r \prod_{l=1}^r w_l^{-\nu_l-1} e^{w_l} \\ \times e^{x/(w_1 \cdots w_r)} {}_0F_1 \Big(\nu_0+1; -ux/(w_1 \cdots w_r)\Big),$$
(1.14)

where $\gamma_1, \ldots, \gamma_r$ are paths starting and ending at negative infinity and encircling the origin once in the positive direction, or equivalently (cf. (4.4), Sect. 4 below),

$$\Psi(u;x) = \frac{1}{\Gamma(\nu_0+1)} \frac{1}{2\pi i} \int_{\gamma} dw \, (-x)^{-w} {}_1F_1(w;\nu_0+1;u) \\ \times \Gamma(w) \prod_{l=1}^r \frac{1}{\Gamma(\nu_l+1-w)},$$
(1.15)

with γ encircling all non-positive integers, while the other reads

$$\Phi(v; y) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \, y^{-s} \phi(v; s) \prod_{l=1}^{r} \Gamma(v_l + s), \tag{1.16}$$

where $c > -\min\{v_0, v_1, ..., v_r\}$, and

$$\phi(v;s) = \frac{1}{\Gamma(v_0+1)} \int_0^\infty dt \, t^{v_0+s-1} e^{-t} {}_0F_1(v_0+1;-vt) \tag{1.17}$$

$$= \frac{\Gamma(\nu_0 + s)}{\Gamma(\nu_0 + 1)} {}_1F_1(\nu_0 + s; \nu_0 + 1; -\nu).$$
(1.18)

In the case r = 0 (1.14) is to be interpreted as

$$\Psi(u;x) = \frac{1}{\Gamma(\nu_0+1)} e^x {}_0F_1(\nu_0+1;-ux), \qquad (1.19)$$

and a calculation shows that (1.16) simplifies to read

$$\Phi(v; y) = \frac{1}{\Gamma(v_0 + 1)} y^{v_0} e^{-y} {}_0 F_1(v_0 + 1; -vy).$$
(1.20)

The two auxiliary functions appear in a double contour integral expression for the correlation kernel, which we present next. Its significance is that it provides the starting point for further asymptotic analysis. The special case r = 0 was previously obtained by Desrosiers and one of the present authors; see [24, Prop. 5].

Proposition 1.1. Let Y be defined by (1.4), and suppose that all eigenvalues a_1, \ldots, a_N of A^*A are positive. The joint density of eigenvalues for Y^*Y can be written in the form

$$\mathcal{P}_N(x_1, \dots, x_N) = \frac{1}{N!} \det[K_N(x_i, x_j)]_{i,j=1}^N$$
(1.21)

with correlation kernel

$$K_N(x, y) = \frac{1}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \, u^{v_0} e^{-u+v} \Psi(u; x) \Phi(v; y) \frac{1}{u-v} \prod_{l=1}^N \frac{u+a_l}{v+a_l}, \quad (1.22)$$

where C is a counterclockwise contour encircling $-a_1, \ldots, -a_N$ but not u.

Remark 1.1. When some of the parameters a_l 's are null, the double integral representation (1.22) remains valid provided that $\int_0^\infty du$ is interpreted as $\lim_{\varepsilon \to 0^+} \int_{\varepsilon}^\infty du$, or for given u > 0 C is chosen such that $\operatorname{Re}\{v\} < u$ with any $v \in C$.

One of the main results in the present paper concerns a double scaling limit near the critical point, which permits a new family of limiting kernels.

Theorem 1.2 (Critical kernel). With the kernel (1.22), for $\tau \in \mathbb{R}$ let

$$a_1 = \dots = a_N = N(1 - \tau/\sqrt{N})^{-1}.$$
 (1.23)

Then we have

$$\lim_{N \to \infty} \frac{1}{\sqrt{N}} K_N \left(\frac{\xi}{\sqrt{N}}, \frac{\eta}{\sqrt{N}} \right)
= \frac{1}{2\pi i} \int_0^\infty du \int_{i\mathbb{R}} dv \left(\frac{u}{v} \right)^{\nu_0} \frac{e^{-\tau u - \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v}
\times G_{0,r+2}^{1,0} \Big(_{0, -\nu_0, -\nu_1, \dots, -\nu_r} \left| u\xi \right) G_{0,r+2}^{r+1,0} \Big(_{\nu_0, \nu_1, \dots, \nu_r, 0} \left| v\eta \right)
=: \mathcal{K}^{\mathbf{h}, r}(\xi, \eta; \tau),$$
(1.24)

valid uniformly for ξ , η in any compact set of $(0, \infty)$ and for τ in any compact set of \mathbb{R} .

In the special case r = 0, upon making use of (1.9) we see from (1.24) that

$$\left(\frac{\xi}{\eta}\right)^{\nu_0/2} \mathcal{K}^{\mathrm{h},0}\left(\xi,\eta;\tau\right) = \frac{1}{2\pi i} \int_0^\infty du \int_{i\mathbb{R}} dv \left(\frac{u}{v}\right)^{\nu_0/2} \frac{e^{-\tau u - \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v} \\ \times J_{\nu_0}(2\sqrt{u\xi}) J_{\nu_0}(2\sqrt{v\eta}),$$
(1.25)

where the integral form on the RHS of the above equation is similar to (1.22) with r = 0 (cf. [24, Prop. 5]). In the study [44, displayed equation below (1.34)], this kernel was conjectured to be an equivalent functional form to that derived therein in the case r = 0. Our work thus provides a direct way of deriving (1.25) for the r = 0 critical kernel. Recently, an understanding of the resulting functional identity has been given in [22, Remark 2.26].

Theorem 1.2 quantifies the limiting correlation kernel for the situation that $a_k = N(1 - \tau/\sqrt{N})^{-1}$ (k = 1, ..., N), which is shown to depend on τ , thus justifying the term critical kernel. A variation on this setting is to have at most finitely many source eigenvalues, say $a_1, ..., a_m$, go to infinity at a smaller but appropriate scale and others remain at the same critical value. This gives rise to a multi-parameter deformation of the critical kernel (1.24).

Theorem 1.3 (Deformed critical kernel). *With the kernel* (1.22), *for a fixed nonnegative integer m, let*

$$a_j = \sqrt{N}\sigma_j, \ j = 1, \dots, m \quad and \quad a_k = N(1 - \tau/\sqrt{N})^{-1}, \quad k = m + 1, \dots, N,$$
(1.26)

where $\tau \in \mathbb{R}$ and $\sigma_1, \ldots, \sigma_m > 0$. Then we have

$$\lim_{N \to \infty} \frac{1}{\sqrt{N}} K_N \left(\frac{\xi}{\sqrt{N}}, \frac{\eta}{\sqrt{N}} \right)
= \frac{1}{2\pi i} \int_0^\infty du \int_{-c-i\infty}^{-c+i\infty} dv \left(\frac{u}{v} \right)^{\nu_0} \frac{e^{-\tau u - \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v}
\times \prod_{j=1}^m \frac{u + \sigma_j}{v + \sigma_j} G_{0,r+2}^{1,0} \Big(_{0, -\nu_0, -\nu_1, \dots, -\nu_r} \left| u\xi \right) G_{0,r+2}^{r+1,0} \Big(_{\nu_0, \nu_1, \dots, \nu_r, 0} \left| v\eta \right)
=: \mathcal{K}_m^{h,r}(\xi, \eta; \tau, \sigma),$$
(1.27)

where $0 < c < \min\{\sigma_1, ..., \sigma_m\}$.

In the simplest case r = 0, upon making use of (1.9) we see from (1.27) that

$$\begin{pmatrix} \frac{\xi}{\eta} \end{pmatrix}^{\nu_0/2} \mathcal{K}_m^{h,0} \Big(\xi, \eta; \tau \Big) = \frac{1}{2\pi i} \int_0^\infty du \int_{-c-i\infty}^{-c+i\infty} dv \\ \times \Big(\frac{u}{v} \Big)^{\nu_0/2} \frac{e^{-\tau u - \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v} \\ \times \prod_{j=1}^m \frac{u + \sigma_j}{v + \sigma_j} J_{\nu_0}(2\sqrt{u\xi}) J_{\nu_0}(2\sqrt{v\eta}).$$
(1.28)

Even in this special case, the kernel (1.27) appears to be new.

We remark that the inter-relationship between the interpolating kernel (1.27) and critical kernel (1.24) is similar in form to that between the interpolating Airy kernel and Airy kernel (see e.g. [1,9]). Furthermore, as the parameter *b* displayed in Eq. (1.11) increases from zero, we will establish a phase transition at the hard edge from the Meijer G-kernel (cf. Theorem 3.1) to the critical and deformed critical kernels (cf. Theorems 1.2 and 1.3), then to the shifted mean LUE kernel (cf. Theorem 3.2); see Sect. 3 for more details. A

similar phase transition occurs in another random matrix product $T_r T_{r-1} \cdots T_1(G_0 + A)$, with each T_i being a truncated unitary matrix; see Sect. 4.

The paper is organized as follows. Section 2 is devoted to the joint eigenvalue probability density function (PDF) and a double contour integral representation for the correlation kernel of the squared singular values of the product (1.4). The proof of Proposition 1.1 will be given, and the formulas for the average of the ratio of characteristic polynomials and a single (inverse) characteristic polynomial are also derived. In Sect. 3 the hard edge limits of the kernel in different regimes are evaluated, which include the proofs of Theorems 1.2 and 1.3. Our methods are used to similarly analyse the product of r truncated unitary matrices and one shifted mean Ginibre matrix in Sect. 4. In Sect. 5 further discussions on asymptotics for large variables, and some open problems, are presented.

2. Eigenvalue PDF and Double Integral for Correlation Kernel

2.1. Correlation kernels. Consider (1.4) in the case r = 0. Let x_1, \ldots, x_N and a_1, \ldots, a_N denote the eigenvalues of Y^*Y and A^*A respectively. It is well known (see e.g. [24, Prop. 5], [27, Sect. 11.6]) that the eigenvalue PDF of the random matrix Y^*Y is an example of a biorthogonal ensemble [15]

$$Q_N(x_1, \dots, x_N) = \frac{1}{Z_N} \det[\eta_i(x_j)]_{i,j=1}^N \det[\xi_i(x_j)]_{i,j=1}^N,$$
(2.1)

where $\eta_i(x) = x^{i-1}, \xi_i(x) = x^{\nu_0} e^{-x_0} F_1(\nu_0+1; a_i x)$, and Z_N denotes the normalisation. Our first task is to specify a functional form for the joint eigenvalue PDF of Y^*Y in the case of general *r*. For this purpose use will be made of a recent result due to Kuijlaars and Stivigny [45].

Proposition 2.1 (Special case of [45, Thm. 2.1]). Let W be an $n \times n$ random matrix, and suppose that the eigenvalue PDF of W^*W can be written in the form

$$\prod_{1 \le j < k \le n} (x_k - x_j) \det[f_{k-1}(x_j)]_{j,k=1}^n$$
(2.2)

for some $\{f_{k-1}(x)\}_{k=1,...,n}$. For $v \ge 0$, let G be an $(n + v) \times n$ standard complex Gaussian matrix. The squared singular values of GW, or equivalently the eigenvalues of $(GW)^*GW$, then have their PDF proportional to

$$\prod_{1 \le j < k \le n} (y_k - y_j) \det[g_{k-1}(y_j)]_{j,k=1}^n,$$
(2.3)

where

$$g_k(y) = \int_0^\infty x^{\nu} e^{-x} f_k\left(\frac{y}{x}\right) \frac{dx}{x}, \quad k = 0, \dots, n-1.$$
 (2.4)

Let Y be defined in (1.4) and let a_1, \ldots, a_N denote the eigenvalues of A^*A . Starting with (2.1), application of Proposition 2.1 r times in succession shows that the joint eigenvalue PDF of Y^*Y is equal to

$$\frac{1}{Z_N} \det[\eta_i(x_j)]_{i,j=1}^N \det[\xi_i(x_j)]_{i,j=1}^N,$$
(2.5)

where $\eta_i(x) = x^{i-1}$ and $\xi_j(x) = \Phi(-a_j; x)$, while with $T = t_1 \cdots t_r$

$$\Phi(v; y) = \frac{1}{\Gamma(v_0+1)} \int_0^\infty dt_1 \cdots \int_0^\infty dt_r \prod_{i=1}^r t_i^{\nu_i - 1} e^{-t_i} \left(\frac{y}{T}\right)^{\nu_0} e^{-\frac{y}{T}} {}_0F_1(\nu_0+1; -\nu\frac{y}{T}),$$
(2.6)

valid for $r \ge 1$ (for $r = 0 \xi_j(x)$ is defined as below (2.1)). Here $\Phi(v; y)$ is actually the same as defined in (1.16), for which application of the Mellin transform gives

$$\int_0^\infty y^{s-1} \Phi(v; y) \, dy = \phi(v; s) \prod_{l=1}^r \Gamma(v_l + s), \tag{2.7}$$

while use of the inverse Mellin transform gives the sought expression. We stress that when some of the a_j 's in (2.5) coincide L'Hospital's rule provides the appropriate eigenvalue density.

The significance of the structure (2.5) is that it provides a systematic way to compute the corresponding *k*-point correlation function $\rho_{(k)}(x_1, \ldots, x_k)$, where the normalization has been chosen such that integrating gives $N(N - 1) \cdots (N - k + 1)$, see e.g. [27, eqn (5.1)] for the definition.

Proposition 2.2 ([15, Prop. 2.2]). With $g_{i,j} := \int_0^\infty \eta_i(x)\xi_j(x) dx$, let $[g_{i,j}]_{i,j=1}^n$ be invertible for each n = 1, 2, ... Defining $c_{i,j}$ by

$$\left([c_{i,j}]_{i,j=1}^N \right)^t = \left([g_{i,j}]_{i,j=1}^N \right)^{-1}, \tag{2.8}$$

and setting

$$K_N(x, y) = \sum_{i,j=1}^{N} c_{i,j} \eta_i(x) \xi_j(y),$$
(2.9)

we have that the k-point correlation function is given by

$$\rho_{(k)}(x_1, \dots, x_k) = \det[K_N(x_j, x_l)]_{j,l=1}^k.$$
(2.10)

We are now ready to complete the proof of Proposition 1.1.

Proof of Proposition 1.1. Our first task is to compute $g_{i,j} := \int_0^\infty \eta_i(x)\xi_j(x) dx$. For this purpose, we require the fact (see e.g. [7, eqns. (6.2.15), (6.2.33) and (4.5.2)]) that with $L_n^{\nu_0}(y)$ denoting the Laguerre polynomial of degree *n*, one has the Hankel pair

$$L_n^{\nu_0}(y) = \frac{e^y}{n!\Gamma(\nu_0+1)} \int_0^\infty t^{\nu_0+n} e^{-t} {}_0F_1(\nu_0+1;-yt) dt$$
(2.11)

and

$$t^{n} = \frac{n!e^{t}}{\Gamma(\nu_{0}+1)} \int_{0}^{\infty} y^{\nu_{0}} L_{n}^{\nu_{0}}(y) e^{-y} {}_{0}F_{1}(\nu_{0}+1;-ty) \, dy.$$
(2.12)

Combination of (2.7), (1.17) and (2.11) shows that

$$g_{i,j} = (i-1)! e^{a_j} L_{i-1}^{\nu_0}(-a_j) \prod_{l=1}^r \Gamma(\nu_l + i).$$
(2.13)

According to Proposition 2.2, we must now invert the matrix (2.13). With $G = [g_{i,j}]_{i,j=1}^N$, let $C = (G^{-1})^t$, the entries $c_{i,j}$ of C then satisfy

$$e^{a_k} \sum_{i=1}^{N} (i-1)! L_{i-1}^{\nu_0}(-a_k) \prod_{l=1}^{r} \Gamma(\nu_l+i) c_{i,j} = \delta_{j,k}.$$
 (2.14)

Without loss of generality we assume that a_1, \ldots, a_N are pairwise distinct. In this case the above equations imply

$$\sum_{i=1}^{N} (i-1)! L_{i-1}^{\nu_0}(u) \prod_{l=1}^{r} \Gamma(\nu_l+i) c_{i,j} = e^{-a_j} \prod_{l=1, l \neq j}^{N} \frac{-u-a_l}{a_j-a_l},$$
(2.15)

as can be verified by noting that both sides are polynomials of degree N - 1 in u which are equal at N different points since (2.14) is satisfied. Using this implicit formula for $\{c_{i,j}\}$ we now want to show that (2.9) implies the double contour integral formula (1.22).

Using the integral representation of the reciprocal Gamma function

$$\frac{1}{\Gamma(z)} = \frac{1}{2\pi i} \int_{\gamma} w^{-z} e^w dw, \qquad (2.16)$$

we have from (2.9) that

$$\begin{split} K_{N}(x,y) &= \frac{1}{(2\pi i)^{r}} \sum_{i,j=1}^{N} \xi_{j}(y) \int_{\gamma_{1}} dw_{1} \cdots \int_{\gamma_{r}} dw_{r} \prod_{l=1}^{r} w_{l}^{-\nu_{l}-1} e^{w_{l}} \\ &\times \left(\frac{x}{w_{1} \cdots w_{r}}\right)^{i-1} \prod_{l=1}^{r} \Gamma(\nu_{l}+i) c_{i,j} \\ &= \frac{1}{(2\pi i)^{r}} \sum_{j=1}^{N} \xi_{j}(y) \int_{\gamma_{1}} dw_{1} \cdots \int_{\gamma_{r}} dw_{r} \prod_{l=1}^{r} w_{l}^{-\nu_{l}-1} e^{w_{l}} \sum_{i=1}^{N} (i-1)! \prod_{l=1}^{r} \Gamma(\nu_{l}+i) c_{i,j} \\ &\times \frac{1}{\Gamma(\nu_{0}+1)} e^{\frac{x}{w_{1} \cdots w_{r}}} \int_{0}^{\infty} du \, u^{\nu_{0}} L_{i-1}^{\nu_{0}}(u) \, e^{-u_{0}} F_{1}(\nu_{0}+1; -\frac{ux}{w_{1} \cdots w_{r}}) \\ &= \frac{1}{(2\pi i)^{r}} \sum_{j=1}^{N} \xi_{j}(y) \int_{\gamma_{1}} dw_{1} \cdots \int_{\gamma_{r}} dw_{r} \prod_{l=1}^{r} w_{l}^{-\nu_{l}-1} e^{w_{l}} e^{\frac{x}{w_{1} \cdots w_{r}}} \\ &\times \frac{1}{\Gamma(\nu_{0}+1)} \int_{0}^{\infty} du \, u^{\nu_{0}} e^{-u_{0}} F_{1}(\nu_{0}+1; -\frac{ux}{w_{1} \cdots w_{r}}) e^{-a_{j}} \prod_{l\neq j} \frac{-u-a_{l}}{a_{j}-a_{l}}. \end{split}$$
(2.17)

Here the formulae (2.12) and (2.15) have been made use of respectively in the second and third equalities.

Finally, with (2.6) and (1.14) in mind, these facts substituted into (2.17) imply that

$$K_N(x, y) = \int_0^\infty du \, u^{\nu_0} e^{-u} \Psi(u; x) \sum_{j=1}^N \Phi(-a_j; y) \, e^{-a_j} \prod_{l \neq j} \frac{-u - a_l}{a_j - a_l}.$$
 (2.18)

We recognise the sum over j as the sum of the residues at $\{a_l\}$ of

$$e^{-v}\Phi(-v;y)\frac{1}{-u-v}\prod_{l=1}^{N}\frac{-u-a_{l}}{v-a_{l}}$$
(2.19)

considered as a function of v. Applying the residue theorem and changing v to -v, we thus arrive at the desired result. \Box

Remark 2.1. The case in which each $a_l = 0$ has been analysed previously [3,4,46], but using different working. Thus instead of computing the inverse matrix (2.8), functions

$$P_{j-1}(x) \in \text{Span} \{\eta_1(x), \dots, \eta_j(x)\}, \quad Q_{j-1}(x) \in \text{Span} \{\xi_1(x), \dots, \xi_j(x)\}$$

with the biorthogonality property $\int_0^\infty P_k(x)Q_l(x) dx = \delta_{k,l}$ were constructed. In terms of these functions (2.9) simplifies from a double sum to the single sum

$$K_N(x, y) = \sum_{j=1}^{N} P_{j-1}(x) Q_{j-1}(y).$$
(2.20)

Instead of (1.22) with each $a_l = 0$ (which strictly speaking is ill-defined due to the restriction on the contour C, but can be well understood in a limiting sense, cf. Remark 1.1), this leads to the double integral formula

$$K_N(x, y) = \frac{1}{(2\pi i)^2} \int_{-1/2 - i\infty}^{-1/2 + i\infty} du \oint_{\Sigma} dt \prod_{j=-1}^r \frac{\Gamma(v_j + u + 1)}{\Gamma(v_j + t + 1)} \times \frac{\Gamma(t - N + 1)}{\Gamma(u - N + 1)} \frac{x^t y^{-(u+1)}}{u - t},$$
(2.21)

where Σ is a simple closed contour encircling anti-clockwise t = 0, 1, ..., N-1 but not u; see [46, Prop. 5.1] for the detailed derivation, where similar integral representations for both multiple orthogonal functions P_{j-1} and Q_{j-1} are first derived and then the double integral follows from a particular combination.

Next, we further investigate Proposition 1.1 and establish a corollary under the assumption that all but a fixed number of source parameters are equal to *a*. Precisely, for $m \ge 0$ let $a_{m+1} = \cdots = a_N = a$. More definitions are also needed. For $k = 1, 2, 3, \ldots$ and $n = 0, 1, 2, \ldots$, set

$$\mathcal{L}_{n}^{(k)}(x;a,a_{1},\ldots,a_{k-1}) = \int_{0}^{\infty} du \, u^{\nu_{0}} e^{-u} \Psi(u;x)(u+a)^{n} \prod_{l=1}^{k-1} (u+a_{l}), \quad (2.22)$$

and

$$\widetilde{\mathcal{L}}_{n}^{(k)}(x; a, a_{1}, \dots, a_{k}) = \frac{1}{2\pi i} \int_{\mathcal{C}} dv \, e^{v} \Phi(v; x) (v+a)^{-n} \prod_{l=1}^{k} \frac{1}{v+a_{l}}, \qquad (2.23)$$

where C is a counterclockwise contour encircling $-a, -a_1, \ldots, -a_k$ but not any point on the positive real axis.

Corollary 2.3. Let K_N be the kernel (1.22), and for $m \ge 0$ let

$$a_{m+1} = \dots = a_N = a. \tag{2.24}$$

Then we have

$$K_N(x, y) = \frac{1}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \, u^{v_0} e^{-u+v} \Psi(u; x) \Phi(v; y) \frac{1}{u-v} \left(\frac{u+a}{v+a}\right)^{N-m} + \sum_{k=1}^m \mathcal{L}_{N-m}^{(k)}(x; a, a_1, \dots, a_{k-1}) \tilde{\mathcal{L}}_{N-m}^{(k)}(x; a, a_1, \dots, a_k),$$
(2.25)

where C is a counterclockwise contour encircling -a but not u.

Proof. We will use the identity

$$\frac{1}{u-v}\prod_{l=1}^{m}\frac{u+a_l}{v+a_l} = \frac{1}{u-v} + \sum_{k=1}^{m}\frac{\prod_{l=1}^{k-1}(u+a_l)}{\prod_{l=1}^{k}(v+a_l)}$$
(2.26)

which has been proved by induction in [23, see the equation (5.12) therein]. A direct proof can be given as follows. Rewriting $u - v = u + a_k - (v + a_k)$, we have

$$(u-v)\sum_{k=1}^{m}\frac{\prod_{l=1}^{k-1}(u+a_l)}{\prod_{l=1}^{k}(v+a_l)} = \sum_{k=1}^{m}\prod_{l=1}^{k}\frac{u+a_l}{v+a_l} - \sum_{k=1}^{m}\prod_{l=1}^{k-1}\frac{u+a_l}{v+a_l} = \prod_{l=1}^{m}\frac{u+a_l}{v+a_l} - 1,$$
(2.27)

and (2.26) follows.

Recalling (2.24), substituting (2.26) in (1.22), comparing the sought equation with (2.22) and (2.23) this completes the proof. \Box

2.2. Average of characteristic polynomials. Recall from Eq. (2.1) that a biorthogonal ensemble [15] refers to the joint density function

$$Q_N(x_1, \dots, x_N) = \frac{1}{Z_N} \det[\eta_i(x_j)]_{i,j=1}^N \det[\xi_i(x_j)]_{i,j=1}^N,$$
(2.28)

where all variables x_1, \ldots, x_N are assumed to lie in the same interval $I \subseteq \mathbb{R}$ for simplicity.

For the special case $\eta_i = x^{i-1}$, the average ratio of characteristic polynomials under the density (2.28) can be expressed in terms of the correlation kernel; thus as a minor variant of [24, Prop. 1] we have the following.

Proposition 2.4. With the same assumption and notation as in Proposition 2.2, let $\eta_j(x) = x^{j-1}$ for j = 1, 2, ... Then, for $z \in \mathbb{C} \setminus I$

$$\mathbb{E}\left[\prod_{l=1}^{N} \frac{x - x_{l}}{z - x_{l}}\right] = \int_{I} du \, \frac{x - u}{z - u} K_{N}(x, u).$$
(2.29)

Equivalently, if for $x \in \mathbb{R}$ we define the residue

$$\operatorname{Res}_{z=x} f(z) = \lim_{\varepsilon \to 0^+} \frac{1}{\pi} \operatorname{Im} f(x - i\varepsilon), \qquad (2.30)$$

then

$$K_N(x, y) = \frac{1}{x - y} \operatorname{Res}_{z = x} \mathbb{E} \Big[\prod_{l=1}^N \frac{x - x_l}{z - x_l} \Big].$$
(2.31)

In the case of the average of a single characteristic polynomial or its reciprocal, alternative expressions are also available; cf. Proposition 2 of [24].

Proposition 2.5. With the same assumption and notation as in Proposition 2.2, let $\eta_j(x) = x^{j-1}$ for j = 1, 2, ... Then,

$$\mathbb{E}\left[\prod_{l=1}^{N} \frac{1}{z - x_{l}}\right] = \frac{N!}{Z_{N}} \begin{vmatrix} g_{1,1} & g_{1,2} & \cdots & g_{1,N} \\ \vdots & \vdots & \ddots & \vdots \\ g_{N-1,1} & g_{N-1,2} & \cdots & g_{N-1,N} \\ \int_{I} du \frac{\xi_{1}(u)}{z - u} & \int_{I} du \frac{\xi_{2}(u)}{z - u} & \cdots & \int_{I} du \frac{\xi_{N}(u)}{z - u} \end{vmatrix}$$

$$= \int_{I} du \frac{1}{z - u} \sum_{j=1}^{N} c_{N,j} \eta_{j}(u), \qquad (2.33)$$

for $z \in \mathbb{C} \setminus I$ and

$$\mathbb{E}\left[\prod_{l=1}^{N} (x - x_l)\right] = \frac{N!}{Z_N} \begin{vmatrix} g_{1,1} & \cdots & g_{1,N} & \eta_1(x) \\ g_{2,1} & \cdots & g_{2,N} & \eta_2(x) \\ \vdots & \vdots & \vdots & \vdots \\ g_{N+1,1} & \cdots & g_{N+1,N} & \eta_{N+1}(x) \end{vmatrix}$$
(2.34)

$$= \frac{1}{\tilde{c}_{N+1,N+1}} \sum_{j=1}^{N+1} \tilde{c}_{j,N+1} x^{j-1}, \qquad (2.35)$$

where the normalization $Z_N = N! \det[g_{i,j}]_{i,j=1}^N$, and $\tilde{c}_{j,N+1}$ is the (N+1, j) entry of the inverse of $[g_{i,j}]_{i,j=1}^{N+1}$.

Proof. The formulas (2.32) and (2.34) have been proved in Proposition 2 of [24]. After noting the facts $[g_{i,j}]_N [c_{i,j}]_N^t = I_N$ and $N!/Z_N = \det[c_{i,j}]_N^t$, building on (2.32) simple manipulation gives (2.33). On the other hand, by $[\tilde{c}_{i,j}]_{N+1}^t [g_{i,j}]_{N+1} = I_{N+1}$ and

$$\frac{N!}{Z_N} = \frac{Z_{N+1}}{(N+1)Z_N} \frac{(N+1)!}{Z_{N+1}} = \frac{Z_{N+1}}{(N+1)Z_N} \det[\tilde{c}_{i,j}]_{N+1}^t,$$
(2.36)

we have from (2.34) that

$$\mathbb{E}\Big[\prod_{l=1}^{N} (x - x_l)\Big] = \frac{Z_{N+1}}{(N+1)Z_N} \sum_{j=1}^{N+1} \tilde{c}_{j,N+1} x^{j-1}, \qquad (2.37)$$

which further implies the sought Eq. (2.35) since it is a monic polynomial. \Box

Application of the previous two propositions gives us explicit evaluation of averages of characteristic polynomials for the product of random matrices (1.4).

Proposition 2.6. For the eigenvalue PDF (2.5), the following hold true.

(i) Let K_N be the kernel (1.22), then for $z \in \mathbb{C} \setminus \mathbb{R}$,

$$\mathbb{E}\Big[\prod_{l=1}^{N} \frac{x - x_l}{z - x_l}\Big] = \int_0^\infty du \frac{x - u}{z - u} K_N(x, u).$$
(2.38)

(ii) Let Φ be given by (1.16), then for $z \in \mathbb{C} \setminus \mathbb{R}$,

$$\mathbb{E}\Big[\prod_{l=1}^{N} \frac{1}{z - x_{l}}\Big] = \frac{1}{2\pi i} \frac{(-1)^{N-1}}{\prod_{l=1}^{r} \Gamma(v_{l} + N)} \int_{0}^{\infty} dt \\ \times \int_{\mathcal{C}} dv \, e^{v} \Phi(v; t) \frac{1}{z - t} \prod_{l=1}^{N} \frac{1}{v + a_{l}}$$
(2.39)

where C is a counterclockwise contour encircling $-a_1, \ldots, -a_N$ but not any point on the positive real axis.

(iii) Let Ψ be given by (1.14), then

$$\mathbb{E}\Big[\prod_{l=1}^{N} (x - x_l)\Big] = (-1)^N \prod_{l=1}^{r} \Gamma(\nu_l + N + 1) \\ \times \int_0^\infty du \, u^{\nu_0} e^{-u} \, \Psi(u; x) \prod_{l=1}^{N} (u + a_l).$$
(2.40)

Proof. It is immediate that Proposition 2.4 implies (i). For (ii), noting that the leading term of the Laguerre polynomial is

$$n!L_n^{\nu_0}(x) = (-x)^n + \cdots, \qquad (2.41)$$

dividing by $(-u)^{N-1}$ and taking the limit $u \to \infty$ in (2.15) we see that

$$c_{N,j} = \prod_{l=1}^{r} \frac{1}{\Gamma(\nu_l + N)} e^{-a_j} \prod_{l=1, l \neq j}^{N} \frac{1}{a_j - a_l}.$$
 (2.42)

Substituting $c_{N,j}$ in (2.33) and noting $\eta_j(u) = \Phi(-a_j; u)$, we obtain (2.39).

For (iii), we first introduce an auxiliary variable a_{N+1} and set $\eta_{N+1}(u) = \Phi(-a_{N+1}; u)$. The fact that $([g_{i,j}]_{N+1})^t [\tilde{c}_{i,j}]_{N+1} = I_{N+1}$ implies

$$\tilde{c}_{N+1,N+1} = \frac{\det[g_{i,j}]_N}{\det[g_{i,j}]_{N+1}} = \prod_{l=1}^r \frac{1}{\Gamma(\nu_l + N + 1)} e^{-a_{N+1}} \prod_{l=1}^N \frac{1}{a_{N+1} - a_l}.$$
 (2.43)

Changing N to N + 1 and using (2.15), as derived in (2.17) we obtain

$$\sum_{j=1}^{N+1} \tilde{c}_{j,N+1} x^{j-1} = \int_0^\infty du u^{\nu_0} e^{-u} \Psi(u;x) e^{-a_{N+1}} \prod_{l=1}^N \frac{-u-a_l}{a_{N+1}-a_l}.$$
 (2.44)

Combination of (2.43), (2.44) and (2.35) completes the proof of (iii).

Again, for the eigenvalue PDF (2.5), let

$$Q_{N-1}(x) = \operatorname{Res}_{z=x} \mathbb{E}\left[\prod_{l=1}^{N} \frac{1}{z - x_l}\right], \quad x \in (0, \infty),$$
 (2.45)

then use of Proposition 2.6 (ii) shows

$$Q_{N-1}(x) = \frac{1}{2\pi i} \frac{(-1)^{N-1}}{\prod_{l=1}^{r} \Gamma(v_l + N)} \int_{\mathcal{C}} dv \, e^v \Phi(v; x) \prod_{l=1}^{N} \frac{1}{v + a_l};$$
(2.46)

when a_1, \ldots, a_N are pairwise distinct it is a special case of [24, Proposition 2]. Also, let

$$P_N(x) = \mathbb{E}\Big[\prod_{l=1}^{N} (x - x_l)\Big],$$
(2.47)

combining Corollary 2.3 where *m* is taken to be zero and Proposition 2.6, the correlation kernel K_N given by (1.22) can be expressed as the single sum (2.20) in terms of $P_j(x)$ and $Q_j(x)$. Here, without loss of generality, it is assumed that $P_j(x)$ corresponds to the multi-parameters a_1, \ldots, a_j while $Q_j(x)$ corresponds to a_1, \ldots, a_{j+1} .

Remark 2.2. In the special case r = 0, use of (1.19) shows that (2.40) reduces to

$$P_N(x) = \frac{(-1)^N e^x}{\Gamma(v_0+1)} \int_0^\infty u^{v_0} e^{-u} {}_0F_1(v_0+1; -xu) \prod_{l=1}^N (u+a_l) \, du.$$

This same expression has been derived using combinatorial means in [28], and as the solution of a partial differential equation in [14]. Furthermore, in this case $Q_{N-1}(x)$ and $P_N(x)$ are so-called multiple functions of type I and II respectively, and (2.20) reduces to Corollary 7 in [24]; see [24] or [42] for more details, especially when the parameters a_j 's coalesce into D different values. For the case of $a_1 = \cdots = a_N = 0$ and general r, $Q_{N-1}(x)$ and $P_N(x)$ are also multiple functions of type I and II associated with r + 1 weights; see [46]. However, in the general case it remains as a challenge to identify a multiple orthogonal functions structure.

3. Hard Edge Limits

In this section we choose the source A such that all but possibly a fixed number m of the eigenvalues of A^*A are equal to bN. Three regimes are distinguished: subcritical regime 0 < b < 1, critical regime b = 1 and supercritical regime b > 1; as to the former two regimes, see [43] and [44] for a relevant discussion on non-intersecting Bessel paths which corresponds to the case r = 0. In the present paper we focus on the scaled hard edge limits in the three regimes and leave the bulk and soft-edge limits to a future work; for the case $a_1 = \cdots = a_N = 0$ the latter two limits have been established in [48]. The critical kernel results from a double scaling limit, and its functional form is our main result as stated in Sect. 1. As b increases from zero, we will describe a phase transition from the Meijer G-kernel (1.6) to the critical kernel (cf. Theorem 1.2), then to the shifted mean LUE kernel (1.22) (cf. Theorem 3.2 for the case r = 0). 3.1. Limiting kernels. We first suppose that 0 < b < 1. The hard edge scaling in this parameter range is in fact independent of b, and the hard edge correlation kernel (1.6) already known for the case b = 0 is reclaimed.

Theorem 3.1 (Subcritical regime). With the kernel (1.22), let

$$a_1 = \dots = a_N = bN. \tag{3.1}$$

Then for 0 < b < 1, we have

$$\lim_{N \to \infty} \frac{1}{(1-b)N} K_N\left(\frac{\xi}{(1-b)N}, \frac{\eta}{(1-b)N}\right) = K^{\mathbf{h},r}(\xi, \eta),$$
(3.2)

where $K^{h,r}$ is given by (1.6), valid uniformly for ξ , η in any compact set of $(0, \infty)$.

Proof. Introducing rescaled variables $x = \xi/((1-b)N)$, $y = \eta/((1-b)N)$ in (1.22) and substituting *u*, *v* by *uN*, *vN* respectively, we obtain

$$\frac{1}{(1-b)N} K_N\left(\frac{\xi}{(1-b)N}, \frac{\eta}{(1-b)N}\right) = \int_0^\infty du \int_{\mathcal{C}} dv \\ \times \frac{1}{2(1-b)\pi i} \frac{e^{-N(f(u)-f(v))}}{u-v} (Nu)^{\nu_0} \Psi\left(Nu; \frac{\xi}{(1-b)N}\right) \Phi\left(Nv; \frac{\eta}{(1-b)N}\right),$$
(3.3)

where

$$f(z) = z - \log(b + z).$$
 (3.4)

Although both the functions Ψ and Φ in the integrand of (3.3) depend on *N*, we will see that for the large *N* they do not enter the saddle point equation (e.g., cf. (3.10) and (3.11) below). So we may perform saddle-point approximations and this is what we will do next.

Consider now the exponent on the RHS of (3.3). Since

$$f'(z) = 1 - \frac{1}{b+z},$$
(3.5)

there is a saddle point $z_0 = 1 - b$. We hereby deform the contour C into the union of two closed contours $C_1 \cup C_2^-$ such that $C_1 = \{z \in \mathbb{C} : |z+b| = 1\}$ and C_2^- is a clockwise contour encircling the segment [0, 1-b] but not -b (C_2^- and C_2 refer to the same curve except that the former indicates the clockwise direction). For instance, we can choose C_2 as the union of two segments from -0.5b to $-b + e^{\pm i\epsilon}$ respectively and an arc $\{z : z = -b + e^{i\theta}, -\epsilon \le \theta \le \epsilon\}$ for some small positive ϵ . With such a choice, we divide the integration over C into two parts, and furthermore rewrite the double integral on the RHS of (3.3) as a sum of two integrals, that is,

$$\frac{1}{(1-b)N}K_N\left(\frac{\xi}{(1-b)N},\frac{\eta}{(1-b)N}\right) = \text{p.v.} \int_0^\infty du \int_{\mathcal{C}_1} dv(\cdot) + \text{p.v.} \int_0^\infty du \int_{\mathcal{C}_2^-} dv(\cdot) = I_1 + I_2.$$
(3.6)

Here the notation p.v. denotes the Cauchy principal value integral. It is worth stressing that, from (3.3), we can put some restrictions on the range of u, v in the above integrals

such that $u \neq 1 - b$ and $v \neq \pm i$. This is done for the convenience of subsequent asymptotic analysis only.

As $N \to \infty$, we claim that the leading contribution of the double integral on the RHS of (3.3) comes from the range of $u \in (0, 1-b)$ and $v \in C_2$. Actually, for I_2 , when u > 1 - b the *v*-integral vanishes by Cauchy's theorem since the integrand does not have any singularities inside C_2 , while for 0 < u < 1 application of the residue theorem gives

$$I_2 = \frac{1}{1-b} \int_0^{1-b} du \, (Nu)^{\nu_0} \Psi \Big(Nu; \frac{\xi}{(1-b)N} \Big) \Phi \Big(Nu; \frac{\eta}{(1-b)N} \Big). \tag{3.7}$$

Using the asymptotic expansion of the function $_1F_1$ for the large argument (cf. [7, Theorem 4.2.2 and Corollary 4.2.3]), for large *N* we have

$${}_{1}F_{1}(\nu_{0}+s;\nu_{0}+1;-N\nu) = \frac{\Gamma(\nu_{0}+1)}{\Gamma(1-s)}(N\nu)^{-\nu_{0}-s}\left(1+\mathcal{O}(\frac{1}{N})\right), \quad \operatorname{Re}\nu > 0, \quad (3.8)$$

and

$${}_{1}F_{1}(v_{0}+s;v_{0}+1;-Nv) = \frac{\Gamma(v_{0}+1)}{\Gamma(v_{0}+s)}(-Nv)^{s-1}e^{-Nv}\left(1+\mathcal{O}(\frac{1}{N})\right), \quad \text{Re } v < 0.$$
(3.9)

Keeping in mind (1.16) and (1.18), by definition of the Meijer G-function (1.7) we have from (3.8) that

$$(Nu)^{\nu_0} \Phi\left(Nu; \frac{\eta}{(1-b)N}\right) \sim G_{0,r+2}^{r+1,0} \left(\underbrace{-}_{\nu_0, \dots, \nu_r, 0} \middle| \frac{u\eta}{1-b} \right).$$
(3.10)

Here and below we use the notation $f_N \sim g_N$ to mean that $\lim_{N\to\infty} f_N/g_N = 1$. On the other hand, consideration of the definition (1.14) shows

$$\Psi\left(Nu; \frac{\xi}{(1-b)N}\right) \sim G_{0,r+2}^{1,0} \left(\underbrace{-}_{0,-\nu_0,\ldots,-\nu_r} \left| \frac{u\xi}{1-b} \right),$$
(3.11)

where use has been made of the identity (cf. Eq. (1.8))

$$G_{0,r+2}^{1,0}\left(\underbrace{-}_{0,-\nu_0,\ldots,-\nu_r}\left|z\right)=\prod_{l=0}^r\frac{1}{\Gamma(\nu_l+1)}\,_0F_{r+1}\left(\nu_0+1,\ldots,\nu_r+1;-z\right).$$
 (3.12)

Combining (3.7), (3.10) and (3.11), and changing variables we get

$$I_2 \to K^{\mathbf{h},r}(\xi,\eta). \tag{3.13}$$

Next, we deal with the integral I_1 and show that it is negligible. In this case because of different asymptotic forms of $_1F_1$ given in (3.8) and (3.9), we divide I_1 into two parts as

$$I_{1} = \text{p.v.} \int_{0}^{\infty} du \int_{\mathcal{C}_{1,+}} dv(\cdot) + \int_{0}^{\infty} du \int_{\mathcal{C}_{1,-}} dv(\cdot) := I_{11} + I_{12}, \qquad (3.14)$$

where $C_{1,+} = C_1 \bigcap \{v : \text{Re } v > 0\}$ and $C_{1,-} = C_1 \bigcap \{v : \text{Re } v < 0\}$. Notice that for 0 < b < 1 one can easily check that $\text{Re}\{f(u)\}$ attains its global minimum at u = 1 - b

over $(0, \infty)$, while Re{f(v)} attains its global maximum at v = 1 - b over C_1 . Therefore, for I_{11} combining (1.16), (1.18), (3.8) and (3.11) we have

$$I_{11} \sim \text{p.v.} \int_{0}^{\infty} du \int_{\mathcal{C}_{1,+}} dv \frac{1}{2(1-b)\pi i} \frac{e^{-N(f(u)-f(v))}}{u-v} \left(\frac{u}{v}\right)^{v_{0}} G_{0,r+2}^{1,0} \left(\frac{u\xi}{0,-v_{0},\ldots,-v_{r}} \left|\frac{u\xi}{1-b}\right) G_{0,r+2}^{r+1,0} \left(\frac{u\xi}{v_{0},\ldots,v_{r},0} \left|\frac{v\eta}{1-b}\right)\right).$$
(3.15)

For this, the standard steepest descent argument shows that the main contribution comes from the neighbourhood of the saddle point $z_0 = 1 - b$, namely,

$$I_{11} = \mathcal{O}(1/\sqrt{N}).$$
 (3.16)

Similarly, for I_{12} combination of (1.16), (1.18), (3.9) and (3.11) then gives us

$$I_{12} \sim \int_{0}^{\infty} du \int_{\mathcal{C}_{1,-}} dv \frac{1}{2(1-b)\pi i} \frac{e^{-N(f(u)-f(1-b))}}{u-v} G_{0,r+2}^{1,0} \Big(\underbrace{-}_{0,-\nu_{0},\ldots,-\nu_{r}} \Big| \frac{u\xi}{1-b} \Big) \\ \times u^{\nu_{0}} e^{-N(\log(b+\nu)+1-b)} \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \left(\frac{\eta}{1-b}\right)^{-s} (-\nu)^{s-1} N^{\nu_{0}+2s-1} \prod_{l=1}^{r} \Gamma(\nu_{l}+s),$$
(3.17)

for which the integrals of u and v respectively afford us the bounds $O(1/\sqrt{N})$ and $O(N^{\nu_0+2c-1}e^{-(1-b)N})$. Together, we obtain the exponential decay estimation

$$I_{12} = \mathcal{O}(N^{\nu_0 + 2c - 3/2} e^{-(1-b)N}).$$
(3.18)

Combining (3.13), (3.16) and (3.18), we arrive at the Eq. (3.2). Furthermore, it is clear that the previously derived estimates are valid uniformly for ξ , η in any given compact set of $(0, \infty)$. \Box

Remark 3.1. When all the parameters a_l 's are null, if we understand the double integral representation (1.22) as described in Remark 1.1, then the same argument as in the proof of Theorem 3.1 is also applicable. This gives another derivation of (1.6) different from that in [46].

We turn to proofs of Theorems 1.2 and 1.3.

Proof of Theorem 1.2. Rescaling variables in (1.22), we have

$$\frac{1}{\sqrt{N}} K_N\left(\frac{\xi}{\sqrt{N}}, \frac{\eta}{\sqrt{N}}\right) = \frac{\sqrt{N}}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \, \frac{e^{-N(f(u) - f(v))}}{u - v} (Nu)^{\nu_0} \Psi(Nu; \frac{\xi}{\sqrt{N}}) \Phi(Nv; \frac{\eta}{\sqrt{N}}), \quad (3.19)$$

where $f(z) = z - \log(1 + z/b)$ with $b = (1 - \tau/\sqrt{N})^{-1}$. If *b* is equal to the critical value 1, then the saddle point of f(z) is $z_0 = 0$. This time, for a fixed small number $\delta > 0$ we choose the contour as

$$\mathcal{C} = \{ z = -1 + (1 + 2\delta)e^{i\theta} : \theta_0 \le |\theta| \le \pi \} \cup \mathfrak{L}_{A_-OA_+},$$
(3.20)

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where

$$\theta_0 = \arccos \frac{1+\delta}{1+2\delta}, \qquad A_{\pm} = -1 + (1+2\delta)e^{\pm i\theta_0},$$
(3.21)

and $\mathcal{L}_{A_-OA_+}$ denotes the union of two line segments from the point A_- to the origin to the point A_+ . It is clear that $A_{\pm} = (\delta, \pm \sqrt{(2+3\delta)\delta})$, and the intersections of the *y*-axis and the contour C are $B_{\pm} = (0, \pm 2\sqrt{(1+\delta)\delta})$. Moreover, the four points come close to the origin as $\delta \to 0$, which permits us to use the Taylor series expansion of f(v) for any $v \in C_+$ defined below (3.22).

First, we divide the integral on the RHS of (3.19) into two parts

$$\frac{1}{\sqrt{N}}K_N\left(\frac{\xi}{\sqrt{N}},\frac{\eta}{\sqrt{N}}\right) = \int_0^\infty du \int_{\mathcal{C}_-} dv \left(\cdot\right) + \int_0^\infty du \int_{\mathcal{C}_+} dv \left(\cdot\right) := I_- + I_+, \quad (3.22)$$

where $C_{-} = \{v \in C : \text{Re } v < 0\}$ and $C_{+} = \{v \in C : \text{Re } v > 0\}$. We claim that the dominant contribution to (3.19) comes from the neighbourhoods of $u_0 = 0$ and $v_0 = 0$, so we need to expand the function f(z) at $z_0 = 0$. With the double scaling in mind, we obtain the Taylor series

$$f(z) = \frac{\tau z}{\sqrt{N}} + \frac{1}{2} (1 - \frac{\tau}{\sqrt{N}})^2 z^2 - \frac{1}{3} (1 - \frac{\tau}{\sqrt{N}})^3 z^3 + \cdots$$
 (3.23)

Therefore for I_+ , combining (1.14), (1.16), (1.18) and (3.8), together with the relation (3.12) and the definition of Meijer G-function (1.7) we see that

$$I_{+} \sim \frac{\sqrt{N}}{2\pi i} \int_{0}^{\infty} du \int_{\mathcal{C}_{+}} dv \frac{e^{-N(f(u) - f(v))}}{u - v} \left(\frac{u}{v}\right)^{v_{0}} \times G_{0,r+2}^{1,0} \left(\frac{-}{v_{0}, \dots, -v_{r}} \middle| \sqrt{N}u\xi \right) G_{0,r+2}^{r+1,0} \left(\frac{-}{v_{0}, \dots, v_{r}, 0} \middle| \sqrt{N}v\eta \right).$$
(3.24)

Fix the two endpoints B_{\pm} of C_{+} and deform it to the imaginary axis, then after substituting (3.23) into (3.24) and rescaling u, v by $u/\sqrt{N}, v/\sqrt{N}$, we conclude that I_{+} converges to the kernel defined by (1.24), uniformly for ξ, η in a compact set of $(0, \infty)$ and for τ in a compact set of \mathbb{R} .

Secondly, for the integral I_{-} , combination of (1.14), (1.16), (1.18) and (3.9) yields

$$I_{-} \sim \frac{1}{2\pi i} \int_{0}^{\infty} du \int_{\mathcal{C}_{-}} dv \frac{e^{-Nf(u) - N\log(1 + v/b)}}{u - v} u^{\nu_{0}} G_{0,r+2}^{1,0} \Big(\frac{-}{0, -\nu_{0}, \dots, -\nu_{r}} \Big| \sqrt{N} u\xi \Big) \\ \times \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \, \eta^{-s} (-v)^{s-1} N^{\nu_{0} + (3s-1)/2} \prod_{l=1}^{r} \Gamma(\nu_{l} + s). \quad (3.25)$$

Since for sufficiently large N,

$$\operatorname{Re}\{\log(1+\frac{v}{b})\} = \frac{1}{2}\log\left((1+2\delta)^2 + \mathcal{O}(\frac{\tau}{\sqrt{N}})\right) > \log(1+\delta)$$
(3.26)

holds true uniformly for τ in a compact set of \mathbb{R} and for $v \in C_{-}$, use of the steepest descent argument leads to an exponential decay

$$I_{-} = \mathcal{O}\left(N^{\nu_0 - 1 + 1.5c} e^{-N\log(1+\delta)}\right).$$
(3.27)

Lastly, by combining the foregoing results for I_{-} and I_{+} , we then complete the proof. \Box

Proof of Theorem 1.3. Rescaling variables in (1.22), we have

$$\frac{1}{\sqrt{N}}K_N\left(\frac{\xi}{\sqrt{N}},\frac{\eta}{\sqrt{N}}\right) = \frac{1}{2\pi i}\int_0^\infty du \int_{\mathcal{C}} dv N^{\nu_0+1/2}u^{\nu_0}\frac{e^{-N(f(u)-f(v))}}{u-v}$$
$$\times \prod_{j=1}^m \frac{(v+b)(u+\sigma_j/\sqrt{N})}{(u+b)(v+\sigma_j/\sqrt{N})}\Psi(Nu;\frac{\xi}{\sqrt{N}})\Phi(Nv;\frac{\eta}{\sqrt{N}}), \quad (3.28)$$

where $f(z) = z - \log(1 + z/b)$ with $b = (1 - \tau/\sqrt{N})^{-1}$.

Proceeding as in the proof of Theorem 1.2, Taylor expanding f(z) at z = 0, and rescaling u, v by $u/\sqrt{N}, v/\sqrt{N}$, we can complete the proof. \Box

We next consider the supercritical case, that is b > 1. For r = 0, the limiting eigenvalue density has support $[L_1, L_2]$ with $L_1 > 0$ (thus the left-most end changes from the hard to the soft edge as b increases beyond unity as already remarked in the Introduction); see e.g. [43,44]. However, for r > 0 and fixed $v_0, v_1, \ldots, v_r \ge 0$, considerations from free probability theory suggest that the support will include the origin for general b. Nonetheless, in the simplest case of r = 0 a particular tuning and scaling of the supercritical case can be given which, on an appropriate length scale, effectively separates a bunch of eigenvalues near the origin from the rescaled left-end support. A similar result is conjectured to be true for the general r > 0.

Theorem 3.2 (Supercritical regime for r = 0). With the kernel (1.22) where r = 0, for a fixed positive integer m let

$$a_j = \sigma_j b/(b-1), \ j = 1, \dots, m \text{ and } a_k = bN, \ k = m+1, \dots, N, \ (3.29)$$

where b > 1 and $\sigma_1, \ldots, \sigma_m > 0$. Then we have

$$\lim_{N \to \infty} e^{(1-\frac{1}{b})(\eta-\xi)} \left(1-\frac{1}{b}\right) K_N\left(\left(1-\frac{1}{b}\right)\xi, \left(1-\frac{1}{b}\right)\eta\right) = \frac{1}{2\pi i} \frac{\eta^{\nu_0}}{(\Gamma(\nu_0+1))^2} \\ \times \int_0^\infty du \int_{\mathcal{C}} dv \,_0 F_1(\nu_0+1; -u\xi) \,_0 F_1(\nu_0+1; -v\eta) u^{\nu_0} \frac{e^{-u+v}}{u-v} \prod_{j=1}^m \frac{u+\sigma_j}{v+\sigma_j}, \quad (3.30)$$

where C is a counterclockwise contour encircling $-\sigma_1, \ldots, -\sigma_m$ but not u.

Proof. Set $\kappa = (b-1)/b$. For the large N, we have from (1.22) with r = 0 that

$$(1-\frac{1}{b})K_N\left(\left(1-\frac{1}{b}\right)\xi,\left(1-\frac{1}{b}\right)\eta\right) = \frac{\kappa}{2\pi i}\int_0^\infty du \int_{\mathcal{C}} dv \ u^{v_0}\frac{e^{-u+v}}{u-v}$$
$$\times \Psi(u;\kappa\xi)\Phi(v;\kappa\eta)\prod_{j=1}^m \frac{u+\sigma_j/\kappa}{v+\sigma_j/\kappa}\left(\frac{1+u/(bN)}{1+v/(bN)}\right)^{N-m} := I_1+I_2, \quad (3.31)$$

where we have rewritten $C = C_1 \cup C_2$ and $I_j = \int_0^\infty du \int_{C_j} dv$ (·). The closed contour C_1 encircles $-\sigma_1/\kappa, \ldots, -\sigma_m/\kappa$ and on its left lies the path C_2 encircling -bN, beginning

at and returning to $-\infty$. Keeping (1.19) and (1.20) in mind, for the choice of the contour of I_2 we have used the asymptotic property of $_0F_1$ (cf. [49, Sect. 5.11.2])

$${}_{0}F_{1}(\nu_{0}+1;z) = \frac{\Gamma(\nu_{0}+1)}{2\sqrt{\pi}}(-z)^{-\frac{1+2\nu_{0}}{4}} \times \left(e^{-2i\sqrt{-z}}(1+\mathcal{O}(\frac{1}{\sqrt{z}})) + e^{2i\sqrt{-z}}(1+\mathcal{O}(\frac{1}{\sqrt{z}}))\right), \quad |z| \to \infty, \quad (3.32)$$

Again with (1.19) and (1.20) in mind, by taking the limit and changing variables, as $N \to \infty$ it is less difficult to know that $e^{\kappa(\eta-\xi)}I_1$ goes to the desired integral. For the part I_2 , by the fact (3.32), taking the limit in the integrand we see that the *v*-integral over the closed contour C_2 equals zero since the resulting integrand has no pole. The proof is thus completed. \Box

By comparison with (1.22) in the case r = 0, N = m, $\{a_l\} = \{\sigma_l\}$, substitution of (1.19) and (1.20) shows that the RHS of (3.30) is equal to $e^{\eta - \xi} K_m(\xi, \eta)|_{\{a_l\} = \{\sigma_l\}}$, which is equivalent to the kernel for the $m \times m$ Laguerre Unitary Ensemble with a source (see e.g. [24] or [27, Chapter 11]). For general $r \ge 1$, as to the supercritical case of b > 1 the following similar result is expected to be true

$$\lim_{N \to \infty} \left(1 - \frac{1}{b}\right) K_N\left(\left(1 - \frac{1}{b}\right)\xi, \left(1 - \frac{1}{b}\right)\eta\right) = \frac{1}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \left(\frac{u}{\kappa}\right)^{\nu_0} \frac{e^{-u+v}}{u-v}$$
$$\times \Psi(u/\kappa; \kappa\xi) \Phi(v/\kappa; \kappa\eta) \prod_{j=1}^m \frac{u+\sigma_j}{v+\sigma_j} =: \widetilde{\mathcal{K}}_m^{\mathrm{h},r}(\xi, \eta; \kappa; \sigma),$$
(3.33)

where $\kappa = (b-1)/b$, Ψ , Φ are given by (1.14), (1.16), and C is a counterclockwise contour encircling $-\sigma_1, \ldots, -\sigma_m$ but not u. To prove it, if we might control the behavior of $e^{-\kappa u}\Psi(u; \kappa\xi)$ and $e^{\kappa v}\Phi(v; \kappa\eta)$ (for instance, we can try to derive an estimate $e^{\kappa v}\Phi(v; \kappa\eta) = \mathcal{O}(v^{-2})$ as $v \to -\infty$ which should further be expected to vanish subexponentially), then $I_2 \to 0$ and I_1 goes to the desired integral as in the proof of Theorem 3.2. But such an estimate is yet to be found. Furthermore, we expect the correlations implied by (3.33) to be the same as those for $K_m(\xi, \eta)|_{\{a_l\}=\{\sigma_l\}}$ after being multiplied by the factor $g(\kappa; \eta)/g(\kappa; \xi)$ for some properly chosen function g. However, the mechanism which makes this true in the cases $r \ge 1$ remains to be clarified.

Remark 3.2. It is of interest to contrast the scalings of $\{a_j\}$ in Theorems 1.3 and 3.2 applying to the critical and supercritical cases respectively. Some insight as to the chosen values is possible by restricting attention to the case r = 0, for which the squared singular values have the interpretation as non-intersecting Brownian particles confined to a half line, as mentioned in the Introduction. In this interpretation, the initial position of particle *j* is a_j , and the particles evolve for time t = 1. We interpret the values in a_j in Theorem 1.3 as being such that the particles at the hard edge are all of the same order, with the *k* outlier particles appropriately merging with the spectrum edge of the N - k particles which started originally at $N(1 - \tau/\sqrt{N})$. On the other hand, in Theorem 3.2 only the *k* particles starting at order unity from the hard edge are at order unity from the hard edge when t = 1, with the remaining N - k particles never reaching the hard edge by this time.

Remark 3.3. If we strengthen the results in Theorems 1.2, 1.3, 3.1 and 3.2 from uniform convergence into the trace norm convergence of the integral operators with respect to the correlation kernels, then as a direct consequence we have the limiting gap probabilities after rescaling, especially including the smallest eigenvalue distribution; see [27, Chapters 8 & 9]. Since the proof of trace norm convergence is only a technical elaboration that confirms a well-expected result, we do not give the details.

3.2. Characteristic polynomials. In this subsection we want to evaluate scaling limits for the ratio of characteristic polynomials according to three different regimes.

Theorem 3.3. *With the eigenvalue PDF* (2.5), *fix* $m \in \{0, 1, 2, ..., \}$ *and let*

$$a_{m+1} = \cdots = a_N = Nb.$$

(i) Set $a_j = Nb_j$ with $b_j > 0$ for j = 1, ..., m, if 0 < b < 1, then for $\zeta \in \mathbb{C} \setminus \mathbb{R}$,

$$\lim_{N \to \infty} \frac{1}{(1-b)N} \mathbb{E} \Big[\prod_{l=1}^{N} \frac{x_l - \xi/((1-b)N)}{x_l - \zeta/((1-b)N)} \Big] = \int_0^\infty du \frac{\xi - u}{\zeta - u} K^{\mathbf{h}, r}(\xi, u).$$
(3.34)

(ii) Set $a_j = \sqrt{N}\sigma_j$ with $\sigma_j > 0$ for j = 1, ..., m, if $b = 1/(1 - \tau/\sqrt{N})$ with $\tau \in \mathbb{R}$, then for $\zeta \in \mathbb{C} \setminus \mathbb{R}$,

$$\lim_{N \to \infty} \frac{1}{\sqrt{N}} \mathbb{E} \Big[\prod_{l=1}^{N} \frac{x_l - \xi/\sqrt{N}}{x_l - \zeta/\sqrt{N}} \Big] = \int_0^\infty du \frac{\xi - u}{\zeta - u} \mathcal{K}_m^{\mathrm{h},r}(\xi, u; \tau, \sigma).$$
(3.35)

(iii) Set $a_j = \sigma_j b/(b-1)$ with $\sigma_j > 0$ for j = 1, ..., m, if b > 1 and $m \ge 1$, then for r = 0 and for $\zeta \in \mathbb{C} \setminus \mathbb{R}$,

$$\lim_{N \to \infty} (1 - \frac{1}{b}) \mathbb{E} \Big[\prod_{l=1}^{N} \frac{x_l - (1 - \frac{1}{b})\xi}{x_l - (1 - \frac{1}{b})\zeta} \Big] = \int_0^\infty du \frac{\xi - u}{\zeta - u} \widetilde{\mathcal{K}}_m^{\mathrm{h},0}(\xi, u; 1 - \frac{1}{b}; \sigma).$$
(3.36)

Proof. By Proposition 2.6, Theorems 3.1, 1.3 and 3.2 imply the sought results although a minor modification in the proof of Theorem 3.1 is required in relation to (3.34) (in the same circumstance the limiting subcritical kernel still holds true).

Likewise, based on Proposition 2.6, we can prove the following theorem concerning the average of one single characteristic polynomial or its inverse. For this purpose we introduce four sets of generalised multiple functions (we say generalised since only for r = 0 do we know the multiple polynomial system; recall Remark 2.2) of types II and I with *m* parameters $\sigma_1, \ldots, \sigma_m > 0$. For $k = 1, 2, \ldots, m$, we define two sets of generalised multiple functions by

$$\Gamma^{(k)}(x;\sigma_1,\ldots,\sigma_{k-1}) = \int_0^\infty du u^{\nu_0} e^{-\tau u - \frac{1}{2}u^2} \\ \times G_{0,r+2}^{1,0} \Big(\begin{array}{c} \\ 0, -\nu_0, -\nu_1,\ldots, -\nu_r \end{array} \middle| xu \Big) \prod_{j=1}^{k-1} (u+\sigma_j), \quad (3.37)$$

and

$$\widetilde{\Gamma}^{(k)}(x;\sigma_1,\ldots,\sigma_k) = \frac{1}{2\pi i} \int_{-i\infty}^{i\infty} dv v^{-\nu_0} e^{\tau v + \frac{1}{2}v^2} \times G_{0,r+2}^{r+1,0} {\binom{v}{\nu_0,\nu_1,\ldots,\nu_r,0}} |xv| \prod_{j=1}^k \frac{1}{v+\sigma_j}, \quad (3.38)$$

while for $0<\kappa\leq 1$ two sets of Laguerre-like generalised multiple functions are defined by

$$\mathcal{L}^{(k)}(x;\kappa;\sigma_1,\ldots,\sigma_{k-1}) = \int_0^\infty du \,(u/\kappa)^{\nu_0} e^{-u} \Psi(u/\kappa;\kappa x) \prod_{l=1}^{k-1} (u+\sigma_l), \quad (3.39)$$

and

$$\widetilde{\mathcal{L}}^{(k)}(x;\kappa;\sigma_1,\ldots,\sigma_k) = \frac{1}{2\pi i} \int_{\gamma} dv \, e^v \Phi(v/\kappa;\kappa x) \prod_{l=1}^k \frac{1}{v+\sigma_l}.$$
(3.40)

Here γ is a closed path which is encircling $-\sigma_1, \ldots, -\sigma_m$ once in the positive direction.

Theorem 3.4. With the eigenvalue PDF (2.5), fix $m \in \{0, 1, 2, \dots, \}$ and let

$$a_{m+1} = \cdots = a_N = Nb_N$$

(i) Set $a_j = Nb_j$ with $b_j > 0$ for j = 1, ..., m, if 0 < b < 1, then

$$\lim_{N \to \infty} \frac{-\sqrt{N}}{\gamma_N^{(\text{sub})}} \mathbb{E} \Big[\prod_{l=1}^N \frac{1}{x_l - \zeta/((1-b)N)} \Big] \\= \int_0^\infty du \frac{1}{\zeta - u} G_{0,r+2}^{r+1,0} \Big(\bigcup_{\nu_0, \nu_1, \dots, \nu_r, 0} \Big| u \Big),$$
(3.41)

$$\lim_{N \to \infty} \sqrt{N} \prod_{l=1}^{r} (\nu_l + N) \Upsilon_N^{(\text{sub})} \mathbb{E} \Big[\prod_{l=1}^{N} (x_l - \xi/((1-b)N)) \Big] \\= G_{0,r+2}^{1,0} \Big(\bigcup_{0, -\nu_0, -\nu_1, \dots, \nu_r} \Big| \xi \Big)$$
(3.42)

where

$$\Upsilon_N^{(\text{sub})} = (-1)^N \sqrt{2\pi} N^{\nu_0 + N} e^{-(1-b)N} \prod_{l=1}^r \Gamma(\nu_l + N) (1-b)^{\nu_0} \prod_{j=1}^m (1-b+b_j).$$
(3.43)

(ii) Set
$$a_j = \sqrt{N\sigma_j}$$
 with $\sigma_j > 0$ for $j = 1, ..., m$, if $b = 1/(1 - \tau/\sqrt{N})$, then

$$\lim_{N \to \infty} \frac{-\sqrt{N}}{\gamma_N^{(cri)}} \mathbb{E} \left[\prod_{l=1}^N \frac{1}{x_l - \zeta/\sqrt{N}} \right] = \int_0^\infty du \frac{1}{\zeta - u} \widetilde{\Gamma}^{(m)}(u; \sigma_1, ..., \sigma_m), \quad (3.44)$$

$$\lim_{N \to \infty} \sqrt{N} \prod_{l=1}^r (v_l + N) \, \Upsilon_N^{(cri)} \mathbb{E} \left[\prod_{l=1}^N \left(x_l - \xi/\sqrt{N} \right) \right] = \Gamma^{(m+1)}(\xi; \sigma_1, ..., \sigma_m) \quad (3.45)$$

where

$$\Upsilon_N^{(\text{cri})} = (-1)^N N^{N + (\nu_0 - m)/2} e^{\sqrt{N}\tau + \tau^2/2} \prod_{l=1}^r \Gamma(\nu_l + N).$$
(3.46)

(iii) Set $a_j = \sigma_j b/(b-1)$ with $\sigma_j > 0$ for j = 1, ..., m, if b > 1, then for r = 0,

$$\lim_{N \to \infty} \frac{-b}{(b-1)\Upsilon_{N}^{(\text{sup})}} \mathbb{E} \Big[\prod_{l=1}^{N} \frac{1}{x_{l} - (1 - \frac{1}{b})\zeta} \Big]
= \int_{0}^{\infty} du \frac{1}{\zeta - u} \widetilde{\mathcal{L}}^{(m)}(u; 1 - 1/b; \sigma_{1}, \dots, \sigma_{m}), \qquad (3.47)$$

$$\lim_{N \to \infty} \frac{b}{b-1} \prod_{l=1}^{r} (v_{l} + N) \Upsilon_{N}^{(\text{sup})} \mathbb{E} \Big[\prod_{l=1}^{N} (x_{l} - (1 - 1/b)\xi) \Big]
= \mathcal{L}^{(m+1)}(\xi; 1 - 1/b; \sigma_{1}, \dots, \sigma_{m}) \qquad (3.48)$$

where

$$\Upsilon_N^{(\text{sup})} = (-1)^N (bN)^{N-m} (b/(b-1))^m.$$
(3.49)

Proof. By Proposition 2.6, following almost the same procedure as that in Theorems 3.1, 1.3 and 3.2 we can evaluate the scaling limits. As a matter of fact, the proof will be simpler since it only involves a single variable integral. We omit the details. \Box

Let us conclude this section with two relationships between the limiting kernels (cf. (1.27) and (3.33)) and the generalised multiple functions defined by (3.37)–(3.40); cf. Corollary 2.3.

Proposition 3.5. We have

$$\mathcal{K}_{m}^{\mathbf{h},r}(x,y;\tau,\sigma) = \mathcal{K}^{\mathbf{h},r}(x,y;\tau) + \sum_{k=1}^{m} \Gamma^{(k)}(x;\sigma_{1},\ldots,\sigma_{k-1}) \,\widetilde{\Gamma}^{(k)}(y;\sigma_{1},\ldots,\sigma_{k}),$$
(3.50)

and

$$\widetilde{\mathcal{K}}_{m}^{\mathbf{h},r}(x,y;\kappa;\sigma) = \sum_{k=1}^{m} \mathcal{L}^{(k)}(x;\kappa;\sigma_{1},\ldots,\sigma_{k-1}) \,\widetilde{\mathcal{L}}^{(k)}(y;\kappa;\sigma_{1},\ldots,\sigma_{k}). \quad (3.51)$$

Proof. By use of the relation (2.26), noting the definition of involved functions (3.37)–(3.40), term-by-term integration immediately implies the above two formulas. Here use has been made of $\tilde{\mathcal{K}}_{0}^{\mathrm{h},r}(x, y; \kappa; \sigma) = 0$ for the second formula. \Box

4. Product with Truncated Unitary Matrices

The derivation of the double contour integral expression (1.22) for the correlation kernel is expected to be applicable to a wider class of biorthogonal ensembles, specifically to those characterized by the form of (2.1) with $\eta_i(x) = x^{i-1}$ and $\xi_i(x) = h(a_i, x)$ for some appropriate function of two variables h and N generic parameters a_1, \ldots, a_N . In this section we consider the specific case of the biorthogonal ensemble corresponding to the product of r truncated unitary matrices and one shifted mean Ginibre matrix, and derive a double integral representation of the correlation kernel and analyze the scaled limits at the hard edge. Other types of products $X_r \cdots X_1 Z$, where each X_j is a Ginibre or truncated unitary matrix while Z is a spiked Wishart matrix of the form $G_0\Sigma$ or a triangular random matrix (cf. [18,32]), are presently under consideration [47].

Explicitly, instead of (1.4), we now consider the matrix product

$$Y = T_r \cdots T_1(G_0 + A),$$
 (4.1)

where each T_j is an $(N + v_j) \times (N + v_{j-1})$ truncation of a Haar distributed unitary matrix of size $M_j \times M_j$ and G_0 is an $(N + v_0) \times N$ standard complex Gaussian matrix while *A* is of size $(N + v_0) \times N$ and fixed. Here $v_{-1} = 0, v_0, \ldots, v_r$ are the nonnegative integers and $\mu_j := M_j - N > v_j$ (for the general $v_j > -1$ the analysis below is also applicable). In the case that the matrix $(G_0 + A)$ is absent, this product has been studied in a recent paper [40].

An analogue of Proposition 1.1 for the correlation kernel can be given. As in Proposition 1.1, two auxiliary functions are required, and so as to stress the structural similarities, analogous notation is used. Specifically, with r = 1, 2, ..., and $0 \le q \le r$, the first is defined to be

$$\Psi_{q}(u;x) = \frac{1}{(2\pi i)^{r}} \frac{1}{\Gamma(\nu_{0}+1)} \int_{(0,\infty)^{q}} dt \int_{\Gamma} dw \prod_{l=1}^{q} t_{l}^{\mu_{l}} e^{-t_{l}} \\ \times \prod_{l=1}^{r} w_{l}^{-\nu_{l}-1} e^{w_{l}} \exp\left\{x \frac{t_{1} \cdots t_{q}}{w_{1} \cdots w_{r}}\right\} {}_{0}F_{1}\left(\nu_{0}+1; -xu \frac{t_{1} \cdots t_{q}}{w_{1} \cdots w_{r}}\right), \quad (4.2)$$

where $\Gamma = \gamma_1 \times \cdots \times \gamma_r$, and $\gamma_1, \ldots, \gamma_r$ are paths starting and ending at $-\infty$ and encircling the origin anticlockwise, while the other reads

$$\Phi_q(v; y) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \ y^{-s} \phi(v; s) \prod_{l=1}^r \Gamma(v_l + s) \prod_{l=1}^q \frac{1}{\Gamma(\mu_l + s)}, \tag{4.3}$$

where $\phi(v; s)$ is given in (1.18) and $c > -\min\{\nu_0, \nu_1, \dots, \nu_r\}$. It is worth stressing that $\Psi_q(u; x)$ can be expressed as a single contour integral

$$\Psi_{q}(u;x) = \frac{1}{\Gamma(\nu_{0}+1)} \frac{1}{2\pi i} \int_{\gamma} dw (-x)^{-w} {}_{1}F_{1}(w;\nu_{0}+1;u) \times \Gamma(w) \prod_{l=1}^{q} \Gamma(\mu_{l}+1-w) \prod_{l=1}^{r} \frac{1}{\Gamma(\nu_{l}+1-w)},$$
(4.4)

where γ encircles all non-positive integers such that $\operatorname{Re}\{w\} < \min\{\mu_1 + 1, \dots, \mu_q + 1\}$ for any $w \in \gamma$. This is a nice analogue of the definition of the function $\Phi_q(v; y)$ and

can be derived as follows. First, the power series expansions for the two functions e^x and ${}_0F_1(v_0 + 1; -xu)$ give us the relation

$$e^{x_{0}}F_{1}(v_{0}+1;-xu) = \sum_{k=0}^{\infty} \frac{1}{k!} x^{k_{1}}F_{1}(-k;v_{0}+1;u), \qquad (4.5)$$

from which, together with the definition of the function (4.2), by term-by-term integration we then read off

$$\Psi_{q}(u;x) = \frac{1}{\Gamma(\nu_{0}+1)} \sum_{k=0}^{\infty} \frac{1}{k!} x^{k} {}_{1}F_{1}(-k;\nu_{0}+1;u) \\ \times \prod_{l=1}^{q} \Gamma(\mu_{l}+1+k) \prod_{l=1}^{r} \frac{1}{\Gamma(\nu_{l}+1+k)}.$$
(4.6)

With this, noting that the integrand for the integral (4.4) has simple poles at 0, -1, -2, ..., we thereby apply the residue theorem to get the desired result.

Proposition 4.1. Let Y be defined by (4.1), and suppose that all eigenvalues a_1, \ldots, a_N of A^*A are positive. The eigenvalue PDF of Y^*Y can be written as

$$\mathcal{P}_N(x_1, \dots, x_N) = \frac{1}{N!} \det[K_N(x_i, x_j)]_{i,j=1}^N$$
(4.7)

with correlation kernel

$$K_N(x, y) = \frac{1}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \, u^{v_0} e^{-u+v} \Psi_r(u; x) \Phi_r(v; y) \frac{1}{u-v} \prod_{l=1}^N \frac{u+a_l}{v+a_l},$$
(4.8)

where C is a counterclockwise contour encircling $-a_1, \ldots, -a_N$ but not u.

Proof. Starting with the eigenvalue PDF (2.1) of $(G_0 + A)^*(G_0 + A)$, application of [40, Corollary 2.4] *r* times in succession shows that the eigenvalue PDF of Y^*Y is proportional to

$$\det[\eta_i(x_j)]_{i,j=1}^N \det[\xi_i(x_j)]_{i,j=1}^N,$$
(4.9)

where $\eta_i(x) = x^{i-1}$ and with $T = t_1 \cdots t_r$

$$\xi_{i}(x) = \frac{1}{\Gamma(\nu_{0}+1)} \int_{(0,1)^{r}} dt \prod_{l=1}^{r} t_{l}^{\nu_{l}-1} (1-t_{l})^{\mu_{l}-\nu_{l}-1} (\frac{y}{T})^{\nu_{0}} e^{-\frac{y}{T}} {}_{0}F_{1}(\nu_{0}+1;a_{i}\frac{x}{T}).$$
(4.10)

Next, we proceed as in the proof of Proposition 1.1. Our first task is to compute $g_{i,j} := \int_0^\infty \eta_i(x)\xi_j(x) dx$. For this purpose, we note that application of the Mellin transform gives

$$\int_0^\infty y^{s-1}\xi_j(y)\,dy = \phi(-a_j;s)\prod_{l=1}^r B(\nu_l + s, \mu_l - \nu_l),\tag{4.11}$$

where the notation B(a, b) refers to the gamma function evaluation of the beta integral and $\phi(v; s)$ is given in (1.18), while use of the inverse Mellin transform gives

$$\xi_j(y) = \Phi_r(-a_j; y) \prod_{l=1}^r \Gamma(\mu_l - \nu_l),$$
(4.12)

where Φ_r is defined in (4.3) with q = r. Combining (4.11), (1.17) and (2.11), we obtain

$$g_{i,j} = (i-1)! e^{a_j} L_{i-1}^{\nu_0}(-a_j) \prod_{l=1}^r B(\nu_l + i, \mu_l - \nu_l).$$
(4.13)

According to Proposition 2.2, with $G = [g_{i,j}]_{i,j=1}^N$ and $C = (G^{-1})^t$, the entries $c_{i,j}$ of *C* then satisfy

$$e^{a_k} \sum_{i=1}^{N} (i-1)! L_{i-1}^{\nu_0}(-a_k) \prod_{l=1}^{r} B(\nu_l + i, \mu_l - \nu_l) c_{i,j} = \delta_{j,k}.$$
(4.14)

Without loss of generality we assume that a_1, \ldots, a_N are pairwise distinct. The above equations imply

$$\sum_{i=1}^{N} (i-1)! L_{i-1}^{\nu_0}(u) \prod_{l=1}^{r} B(\nu_l + i, \mu_l - \nu_l) c_{i,j} = e^{-a_j} \prod_{l=1, l \neq j}^{N} \frac{-u - a_l}{a_j - a_l}, \quad (4.15)$$

which can be verified by noting that both sides are polynomials of degree N - 1 in u which are equal at N different points. Using this implicit formula for $\{c_{i,j}\}$ and the integral representations

$$\Gamma(z) = \int_0^\infty t^{z-1} e^{-t} dt, \qquad \frac{1}{\Gamma(z)} = \frac{1}{2\pi i} \int_\gamma w^{-z} e^w dw,$$
(4.16)

we have from (2.9) that with $T = t_1 \cdots t_r$ and $W = w_1 \cdots w_r$

$$\begin{split} K_{N}(x, y) &= \frac{1}{(2\pi i)^{r}} \sum_{i,j=1}^{N} \xi_{j}(y) \prod_{l=1}^{r} \frac{1}{\Gamma(\mu_{l} - \nu_{l})} \int_{(0,\infty)^{r}} dt \int_{\Gamma} dw \left(xT/W\right)^{i-1} \\ &\times \prod_{l=1}^{r} \left(t_{l}^{\mu_{l}} w_{l}^{-\nu_{l}-1} e^{w_{l}-t_{l}}\right) \prod_{l=1}^{r} B(\nu_{l} + i, \mu_{l} - \nu_{l}) c_{i,j} \\ &= \frac{1}{(2\pi i)^{r}} \sum_{j=1}^{N} \Phi_{r}(-a_{j}; y) \int_{(0,\infty)^{r}} dt \int_{\Gamma} dw \prod_{l=1}^{r} \left(t_{l}^{\mu_{l}} w_{l}^{-\nu_{l}-1} e^{w_{l}-t_{l}}\right) \frac{e^{xT/W}}{\Gamma(\nu_{0} + 1)} \\ &\times \sum_{i=1}^{N} (i-1)! \prod_{l=1}^{r} B(\nu_{l} + i, \mu_{l} - \nu_{l}) c_{i,j} \int_{0}^{\infty} du \, u^{\nu_{0}} L_{i-1}^{\nu_{0}}(u) e^{-u} {}_{0}F_{1}(\nu_{0} + 1; -uxT/W) \\ &= \frac{1}{(2\pi i)^{r}} \sum_{j=1}^{N} \Phi_{r}(-a_{j}; y) \int_{(0,\infty)^{r}} dt \int_{\Gamma} dw \prod_{l=1}^{r} \left(t_{l}^{\mu_{l}} w_{l}^{-\nu_{l}-1} e^{w_{l}-t_{l}}\right) \frac{e^{xT/W}}{\Gamma(\nu_{0} + 1)} \\ &\times \int_{0}^{\infty} du \, u^{\nu_{0}} e^{-u} {}_{0}F_{1}(\nu_{0} + 1; -uxT/W) e^{-a_{j}} \prod_{l\neq j} \frac{-u - a_{l}}{a_{j} - a_{l}}. \end{split}$$

$$(4.17)$$

Here the formulae (2.12) and (4.15) have been used in the second and third equalities respectively.

Finally, recalling (4.2) we can rewrite (4.17) as

$$K_N(x, y) = \int_0^\infty du \, u^{\nu_0} e^{-u} \Psi_r(u; x) \sum_{j=1}^N \Phi_r(-a_j; y) \, e^{-a_j} \prod_{l \neq j} \frac{-u - a_l}{a_j - a_l}.$$
 (4.18)

If we recognise the sum over j as the sum of the residues at $\{a_l\}$ of the v-function

$$e^{-v}\Phi_r(-v;y)\frac{1}{-u-v}\prod_{l=1}^N\frac{-u-a_l}{v-a_l},$$
(4.19)

by changing v to -v we then arrive at the desired result. \Box

At this stage it would be possible to develop the theory of the corresponding averaged characteristic polynomials and their reciprocals, and then proceed to analyse their hard edge limit; recall Sects. 2.2 and 3.2. However we pass on this, and instead analyse the hard edge phase transition analogous to the workings in Sect. 3.1. Specifically, taking $N \rightarrow \infty$, we keep all v_j fixed and simultaneously let some of μ_1, \ldots, μ_r go to ∞ . Without loss of generality, we suppose that for some $0 \le q \le r$ all $v_1, \ldots, v_r, \mu_1, \ldots, \mu_q$ are constants, and moreover

$$\mu_{q+1}, \dots, \mu_r \to \infty \quad \text{as} \quad N \to \infty;$$
(4.20)

see [40, Theorem 2.8] for the assumptions.

Theorem 4.2 (Subcritical kernel). *With the kernel* (4.8), *for a fixed nonnegative integer m let*

$$a_j = N\sigma_j, \ j = 1, \dots, m \text{ and } a_k = bN, \ k = m+1, \dots, N,$$
 (4.21)

where 0 < b < 1 and $\sigma_1, \ldots, \sigma_m > 0$. Set $c_N = (1 - b)N\mu_{q+1}\cdots\mu_r$. Under the assumption (4.20) we have

$$\lim_{N \to \infty} \frac{1}{c_N} K_N \left(\frac{\xi}{c_N}, \frac{\eta}{c_N} \right) = \int_0^1 G_{q,r+2}^{1,q} {-\mu_1, \dots, -\mu_q \choose 0, -\nu_0, \dots, -\nu_r} |u\xi| G_{q,r+2}^{r+1,0} {\mu_1, \dots, \mu_q \choose \nu_0, \dots, \nu_r, 0} |u\eta| du.$$
(4.22)

Proof. Substituting u, v by uN, vN respectively in (4.8), we obtain

$$\frac{1}{c_N} K_N\left(\frac{\xi}{c_N}, \frac{\eta}{c_N}\right) = \int_0^\infty du \int_{\mathcal{C}} dv \frac{N}{2c_N \pi i} \frac{e^{-N(f(u) - f(v))}}{u - v}$$
$$\times \prod_{j=1}^m \frac{(v+b)(u+\sigma_j)}{(u+b)(v+\sigma_j)} (Nu)^{\nu_0} \Psi_r\left(Nu; \frac{\xi}{c_N}\right) \Phi_r\left(Nv; \frac{\eta}{c_N}\right), \quad (4.23)$$

where $f(z) = z - \log(b + z)$.

We can complete the proof in much the same way as in that of Theorem 3.1. But this time we need to estimate the large N leading terms of the functions Ψ and Φ . That is, we have to rescale variables t_{q+1}, \ldots, t_r and rewrite them according to

$$\Psi_{r}\left(Nu;\frac{\xi}{c_{N}}\right) = \frac{1}{(2\pi i)^{r}} \frac{1}{\Gamma(v_{0}+1)} \int_{(0,\infty)^{r}} dt \int_{\Gamma} dw \prod_{j=q+1}^{r} \left(\mu_{j}^{\mu_{j}+1} e^{\mu_{j}(\log t_{j}-t_{j})}\right) \\ \times \prod_{l=1}^{q} t_{l}^{\mu_{l}} e^{-t_{l}} \prod_{l=1}^{r} w_{l}^{-\nu_{l}-1} e^{w_{l}} \exp\left\{\frac{\xi}{(1-b)N} \frac{t_{1}\cdots t_{q}}{w_{1}\cdots w_{r}}\right\} \\ \times {}_{0}F_{1}\left(v_{0}+1; -\frac{\xi u}{1-b} \frac{t_{1}\cdots t_{q}}{w_{1}\cdots w_{r}}\right),$$
(4.24)

and

$$\Phi_{r}(Nv; \frac{\eta}{c_{N}}) = \frac{1}{2\pi i} \int_{c-i\infty}^{c+i\infty} ds \left(\frac{\eta}{(1-b)N}\right)^{-s} \phi(Nv; s) \\ \times \prod_{l=q+1}^{r} \frac{\mu_{j}}{\Gamma(\mu_{j}+s)} \prod_{l=1}^{r} \Gamma(v_{l}+s) \prod_{l=1}^{q} \frac{1}{\Gamma(\mu_{l}+s)}, \qquad (4.25)$$

then apply the saddle point analysis (see e.g. [60]) to the integrals over t_{q+1}, \ldots, t_r in Ψ_r near the saddle point $t_0 = 1$, or expand the integrand in Φ_r by the Stirling approximation formula as $\mu_{q+1}, \ldots, \mu_r \to \infty$. Tracking the same contour deformations and following almost the same analysis as in Theorem 3.1, the proof will be done. We leave the details to the reader. \Box

The limiting kernel on the RHS of (4.22), with the parameter v_0 absent and r + 1 replaced by r first appeared in [40, Theorem 2.8] as the hard edge correlation kernel for a product of truncated unitary matrices. Clearly, it reduces to the Meijer G-kernel (1.6) in case q = 0. More generally, as remarked in [40] (cf. eqns (2.37) and (2.38) therein), it can be interpreted as a finite rank perturbation of (1.6).

For the critical regime, tracking the same contour deformations and following almost the same analysis as in Theorem 1.3, as for the proof of Theorem 4.2 the required working to establish the following theorem can be given.

Theorem 4.3 (Deformed critical kernel). *With the kernel* (4.8), *for a fixed nonnegative integer m let*

$$a_j = \sqrt{N}\sigma_j, \ j = 1, \dots, m \quad and \quad a_k = N(1 - \tau/\sqrt{N})^{-1}, \quad k = m + 1, \dots, N,$$
(4.26)

where $\tau \in \mathbb{R}$ and $\sigma_1, \ldots, \sigma_m > 0$. Set $c_N = \sqrt{N}\mu_{q+1}\cdots\mu_r$. Under the assumption (4.20) we have

$$\lim_{N \to \infty} \frac{1}{c_N} K_N\left(\frac{\xi}{c_N}, \frac{\eta}{c_N}\right) = \frac{1}{2\pi i} \int_0^\infty du \int_{-c-i\infty}^{-c+i\infty} dv \left(\frac{u}{v}\right)^{v_0} \frac{e^{-\tau u - \frac{1}{2}u^2 + \tau v + \frac{1}{2}v^2}}{u - v}$$
$$\times \prod_{j=1}^m \frac{u + \sigma_j}{v + \sigma_j} G_{q,r+2}^{1,q} \binom{-\mu_1, \dots, -\mu_q}{0, -v_0, \dots, -v_r} \left| u\xi \right) G_{q,r+2}^{r+1,0} \binom{\mu_1, \dots, \mu_q}{v_0, \dots, v_r, 0} \left| v\eta \right), \quad (4.27)$$

where $0 < c < \min\{\sigma_1, ..., \sigma_m\}$.

We remark that the kernels on the RHS of (4.27) reduce to the deformed critical kernels $\mathcal{K}_m^{h,r}$ in (1.27) in case q = 0. These are the most general form of critical kernels that we have derived in the present paper. Moreover, they are new except for the simplest case q = r = m = 0, which as previously remarked corresponds to non-intersecting squared Bessel paths and has been studied in [22, 24, 44].

As to the supercritical regime where b > 1, when $r \ge 1$ we have a similar expectation on the scaling limit (see Eq. (3.33) and relevant description below it), which can be stated as follows. With the kernel (4.8), for a fixed positive integer *m* let

$$a_j = \sigma_j b/(b-1), \ j = 1, \dots, m$$
 and $a_k = bN, \ k = m+1, \dots, N,$ (4.28)

where b > 1 and $\sigma_1, \ldots, \sigma_m > 0$. Set $c_N = \mu_{q+1} \cdots \mu_r b/(b-1)$, then under the assumption (4.20) we have

$$\lim_{N \to \infty} \frac{1}{c_N} K_N\left(\frac{\xi}{c_N}, \frac{\eta}{c_N}\right) = \frac{1}{2\pi i} \int_0^\infty du \int_{\mathcal{C}} dv \left(\frac{u}{\kappa}\right)^{\nu_0} \frac{e^{-u+v}}{u-v} \times \Psi_q(u/\kappa; \kappa\xi) \Phi_q(v/\kappa; \kappa\eta) \prod_{j=1}^m \frac{u+\sigma_j}{v+\sigma_j}, \qquad (4.29)$$

where $\kappa = (b-1)/b$, Ψ_q , Φ_q are given by (4.2), (4.3), and C is a counterclockwise contour encircling $-\sigma_1, \ldots, -\sigma_m$ but not u.

5. Asymptotics for Large Parameters and Variables

5.1. Limits for large parameters. The behavior of the critical kernel (1.24) for large values of the parameters will be discussed, one of which is the confluent relation between correlation kernels. The first to be considered is when some of v_1, \ldots, v_r , say v_{m+1}, \ldots, v_r , go to infinity.

Proposition 5.1. Let $\mathcal{K}^{h,r}(\xi, \eta; \tau)$ be the critical kernel (1.24). If $0 \le m < r$, then as $\nu_{m+1}, \ldots, \nu_r \to \infty$ we have

$$(\nu_{m+1}\cdots\nu_r)\mathcal{K}^{\mathbf{h},r}\big((\nu_{m+1}\cdots\nu_r)x,(\nu_{m+1}\cdots\nu_r)y;\tau\big)\longrightarrow\mathcal{K}^{\mathbf{h},m}(x,y;\tau).$$
 (5.1)

Proof. This immediately follows from the identity (3.12) for $G_{0,r+2}^{1,0}$ and the definition (1.7) for $G_{0,r+2}^{r+1,0}$. \Box

The above confluent relation allows for a natural interpretation, particularly in the original finite matrix dimension. Actually, in (1.4) substituting all G_j as square matrices being distributed according to the joint density proportional to $\det^{v_j}(G_j^*G_j) \exp\{-\operatorname{tr}(G_j^*G_j)\}$ (see the relevant description below (1.4)), then a saddle point approximation shows that all G_{m+1}, \ldots, G_r go to the identity matrix of order N as $v_m, \ldots, v_r \to \infty$. Thus these matrices do not contribute to the hard edge state.

A similar effect happens in relation to the parameter v_0 associated with G_0 , although now we find that a different rescaling is necessary, and furthermore that the limiting kernel is now subcritical. **Proposition 5.2.** Let $\mathcal{K}^{h,r}(\xi, \eta; \tau)$ be the critical kernel (1.24). For $r \ge 1$, we have

$$\lim_{\nu_0 \to \infty} \sqrt{\nu_0} \mathcal{K}^{\mathbf{h},r} \left(\sqrt{\nu_0} x, \sqrt{\nu_0} y; \tau \right) = K^{\mathbf{h},r-1}(x,y) \Big|_{\{\nu_0,\dots,\nu_{r-1}\} \to \{\nu_1,\dots,\nu_r\}},$$
(5.2)

where $K^{h,r-1}$ is given by (1.6).

Proof. Substituting u, v by $\sqrt{v_0}u$ and $\sqrt{v_0}v$ respectively in (1.24), we get

$$\sqrt{\nu_0} \mathcal{K}^{\mathbf{h},r} \left(\sqrt{\nu_0} x, \sqrt{\nu_0} y; \tau \right) = \frac{1}{2\pi i} \int_0^\infty du \int_{i\mathbb{R}} dv \, \frac{e^{-\nu_0 (f(u) - f(v))}}{u - v} e^{-\tau \sqrt{\nu_0} u + \tau \sqrt{\nu_0} v} \\ \times \nu_0 G_{0,r+2}^{1,0} \left(0, -\nu_0, -\nu_1, \dots, -\nu_r \left| \nu_0 u x \right) G_{0,r+2}^{r+1,0} \left(\nu_0, \nu_1, \dots, \nu_r, 0 \left| \nu_0 v y \right) \right) \right)$$
(5.3)

where $f(z) = -\log z + z^2/2$.

Choose one saddle point $z_0 = 1$ from f'(z) = 0 and deform $i\mathbb{R}$ as the union of one closed clockwise contour C encircling the interval [0, 1) and the vertical line x = 1. Note that as $\nu_0 \to \infty$

$$\nu_{0}G_{0,r+2}^{1,0}\Big(_{0,-\nu_{0},-\nu_{1},\ldots,-\nu_{r}}|\nu_{0}ux\Big)G_{0,r+2}^{r+1,0}\Big(_{\nu_{0},\nu_{1},\ldots,\nu_{r},0}|\nu_{0}vy\Big) \\ \sim G_{0,r+1}^{1,0}\Big(_{0,-\nu_{1},\ldots,-\nu_{r}}|ux\Big)G_{0,r+1}^{r,0}\Big(_{\nu_{1},\ldots,\nu_{r},0}|vy\Big), \quad (5.4)$$

proceeding as in the proof of Theorem 3.1, we can show that the dominant contribution comes from the range of $u \in [0, 1)$ and $v \in C$. Finally, application of the residue theorem gives the proof. \Box

Similarly, for large negative τ , we observe a transition from the critical kernel to the Meijer G-kernel. This is to be expected, as then the parameter *b* in (1.11) enters the subcritical regime b < 1, since effectively $b = (1 - \tau/N)^{-1}$.

Proposition 5.3. Let $\mathcal{K}^{h,r}(\xi,\eta;\tau)$ be the critical kernel (1.24). Then we have

$$\lim_{\tau \to -\infty} (-1/\tau) \mathcal{K}^{\mathbf{h},r} \left(-x/\tau, -y/\tau; \tau \right) = K^{\mathbf{h},r}(x, y).$$
(5.5)

Proof. Substituting u, v by $-\tau u$ and $-\tau v$ respectively in (1.24), we get

$$(-1/\tau)\mathcal{K}^{\mathbf{h},r}\left(-x/\tau,-y/\tau;\tau\right) = \frac{1}{2\pi i} \int_{0}^{\infty} du \int_{i\mathbb{R}} dv \left(\frac{u}{v}\right)^{v_{0}} \frac{e^{-\tau^{2}(f(u)-f(v))}}{u-v} \times G_{0,r+2}^{1,0}\left(_{0,-v_{0},-v_{1},\ldots,-v_{r}}\left|ux\right)G_{0,r+2}^{r+1,0}\left(_{v_{0},v_{1},\ldots,v_{r},0}\left|vy\right)\right)$$
(5.6)

where $f(z) = -z + z^2/2$. Proceeding as in the proof of Proposition 5.2, the sought result follows. \Box

Lastly, as to the critical kernel on the RHS of (4.27) with m = 0, the functions $\tilde{\Gamma}^{(1)}(x)$ and $\tilde{\Gamma}^{(0)}(x)$ defined in (3.37) and (3.38), there exists similar asymptotic behavior for large parameters as in the above three propositions, but we refrain from writing them down.

5.2. Conjectures and open problems. In the concluding section of [29] a number of questions, mostly relating to asymptotics, were posed in relation to the kernel (1.6). As we will indicate, these all carry over to the critical kernel (1.24). It is also the case that the conjectured behaviours are all closely related to analogous expected asymptotic properties of the finite N kernel (1.22). Two classes of asymptotic problems stand out.

The first is to establish the global scaling limit of the critical one-point function. For this we expect

$$\lim_{N \to \infty} N^r K_N(N^{r+1}x, N^{r+1}x) \Big|_{a_l = N} = \frac{1}{\pi} \operatorname{Im} G(x - i0),$$
(5.7)

where w(z) := zG(z), satisfies the algebraic equation

$$w^{r+3/2} - zw^{1/2} + z = 0. (5.8)$$

The latter is known to specify the Raney distribution with parameters (3 + 2r, 2), which according to free probability theory is the global density for the matrix (1.4) in the critical case (see e.g. [31, Remark 3.4]). In the case of the global limit (5.7) with $a_l = 0$ (l = 1, ..., N), a recent achievement [48] has been the use of the double contour integral formula (2.21) to deduce that (5.7) with w(z) := zG(z) satisfies the algebraic equation

$$w^{r+2} - zw + z = 0. (5.9)$$

The latter specifies the Raney distribution with parameters (r + 2, 1), also known as the Fuss–Catalan distribution with parameter r + 1 [56], and should give the asymptotic behavior of global density for small argument throughout the subcritical regime. In the supercritical regime, from a macroscopic viewpoint the number of random matrices in the product (1.4) is effectively r, since A dominates G_0 and moreover A is proportional to the identity. This implies that the corresponding asymptotic behaviour of the global density near the origin now corresponds to that of the Fuss–Catalan distribution with parameter r.

To see the relevance of (5.7) to the asymptotics of the density in the critical hard edge scaled state, $\mathcal{K}^{h,r}(x, x)$, we recall (cf. [31, Cor. 2.5]) that it can be deduced from (5.9) that for small *x* the global density has its leading asymptotics given by (1.12). In keeping with the discussion in the concluding section of [29], this should be the leading large *x* asymptotic form of $\mathcal{K}^{h,r}(x, x)$. Combining this with the small *x* asymptotic form (1.5) for the Fuss–Catalan density as applies to the subcritical and supercritical regimes (the latter with $r \mapsto r - 1$ as already commented), we therefore expect

$$\mathcal{K}^{h,r,\,(*)}(x,x) \underset{x \to \infty}{\sim} \begin{cases} \frac{1}{\pi} \sin \frac{\pi}{r+2} x^{-1+\frac{1}{r+2}}, & (*) = \text{subcritical} \\ \frac{1}{\pi} \sin \frac{2\pi}{2r+3} x^{-1+\frac{1}{r+3/2}}, & (*) = \text{critical} \\ \frac{1}{\pi} \sin \frac{\pi}{r+1} x^{-1+\frac{1}{r+1}}, & (*) = \text{supercritical.} \end{cases}$$
(5.10)

In general if the global density at the hard edge diverges as x^{-p} , then the expected number of eigenvalues in the interval (0, s) is proportional to Ns^{1-p} . For this to be of order unity we must scale $s \mapsto N^{1/(1-p)}s$. Taking into consideration the scaling

 $x \mapsto N^{r+1}x$ already present in (5.7), this suggests that the appropriate hard edge scalings are

$$x \mapsto \begin{cases} (1/N)x, & \text{subcritical} \\ (1/\sqrt{N})x, & \text{critical} \\ \text{no change, supercritical,} \end{cases}$$
(5.11)

in agreement with those used in the main body of the text.

The second class of asymptotic of the type identified in the concluding section of [29] is to compute the leading asymptotic form of the off diagonal analogue of the LHS of (5.7), namely

$$\hat{K}_N(x, y) := N^{r+1} K_N(N^{r+1}x, N^{r+1}y), \quad x \neq y,$$
(5.12)

where the dot above := indicates that terms which oscillate and average to zero are to be ignored. To see the interest in this quantity, note from (2.10) that the truncated (or connected) two-point correlation $\rho_{(2),N}^T(x_1, x_2) := \rho_{(2),N}(x_1, x_2) - \rho_{(1),N}(x_1)\rho_{(1),N}(x_2)$ is given by $\rho_{(2),N}^T(x_1, x_2) = -K_N(x_1, x_2)K_N(x_2, x_1)$, so knowledge of the asymptotics of $\hat{K}(x, y)$ tells us the asymptotics of

$$\hat{\rho}_{(2),N}^{T}(x,y) := N^{2(r+1)} \rho_{(2),N}^{T}(N^{r+1}x, N^{r+1}y), \qquad x \neq y,$$
(5.13)

With $G = \sum_{j=1}^{N} g(x_j)$ denoting a linear statistic in the bulk scaled system, in view of the formula (see e.g. [27, eqn. (14.38)])

$$\operatorname{Var} G = N^{2(r+1)} \int_0^\infty dx_1 \int_0^\infty dx_2 \, g(x_1) g(x_2) \rho_{(2),N}^T (N^{r+1} x_1, N^{r+1} x_2) + N^{r+1} \int_0^\infty g^2(x) \rho_{(1)}(N^{r+1} x) \, dx$$
(5.14)

one sees that (5.13) (sometimes referred to as a wide correlator; see e.g. [36]) essentially determines the large N form of this fluctuation, which is expected to be O(1) (see e.g. [27, Sect. 14.3]).

As a concrete example of this second type of asymptotics, consider the simplest case of (1.4), namely r = 0 and A = 0. The squared singular values correspond to the eigenvalues of $G_0^*G_0$, where G_0 is a $(N + \nu_0) \times N$ standard complex Gaussian matrix. This class of random matrices is referred to as the complex Wishart ensemble (see e.g. [27, Sect. 3.2]). For this ensemble it is a known result that [10]

$$N^{2}\rho_{(2),N}^{T}(Nx,Ny) \sim -\frac{1}{2\pi^{2}} \frac{1}{(x-y)^{2}} \frac{(L/2)(x+y) - xy}{(x(L-x)y(L-y))^{1/2}}, \qquad x \neq y, \quad (5.15)$$

with L = 4, and where the dot above the asymptotic sign denotes a restriction to non-oscillatory terms.

Suppose now that in the definition (5.12) of $\hat{K}_N(x, y)$ we introduce a scale factor L and compute instead the asymptotic form of $(N^{r+1}/L)K_N(N^{r+1}x/L, N^{r+1}y/L)$. For the complex Wishart ensemble the RHS of (5.15) with L a variable results. For general

r, if the original leading asymptotic form of $\hat{\rho}_{(2)}^T(x, y)$ was R(x, y), this will now equal $(1/L^2)R(x/L, y/L)$. Following [10] we expect that

$$\lim_{L \to \infty} \frac{1}{L^2} R\left(\frac{x}{L}, \frac{y}{L}\right) \to R^{\rm h}(x, y), \tag{5.16}$$

where $R^{h}(x, y)$ is the leading non-oscillatory large *x*, large *y* asymptotic form of the hard edge scaling of $\rho_{(2),N}^{T}(x, y)$. In the context of the present setting this corresponds to seeking the large *x*, large *y* form of $\mathcal{K}^{h,r}(x, y)$. In the case of the complex Wishart ensemble, (5.16) applied to (5.15) predicts that

$$\rho_{(2)}^{\mathbf{h},T}(x,y) \stackrel{\cdot}{\sim} -\frac{1}{4\pi^2} \frac{(x/y)^{1/2} + (y/x)^{1/2}}{(x-y)^2},\tag{5.17}$$

which is in fact a known exact result (see e.g. [27, eqn. (7.75)]). The analogue of (5.17) is known for the case r = 1, A = 0 of (1.4) [29, eqn. (5.28)], but the analogue of (5.15) is yet to be obtained. As discussed in [29], knowledge of an asymptotic form such as (5.17) is of interest for the computation of the variance of a scaled linear statistic at the hard edge, $G_{\alpha} = \sum_{j=1}^{\infty} g(x_j/\alpha)$ when $\alpha \to \infty$, which is given by

$$\lim_{\alpha \to \infty} \operatorname{Var} G_{\alpha} := \lim_{\alpha \to \infty} \left(\int_{0}^{\infty} d\lambda_{1} \int_{0}^{\infty} d\lambda_{2} \times g(\lambda_{1}/\alpha)g(\lambda_{2}/\alpha)\rho_{(2)}^{T,h,r}(\lambda_{1},\lambda_{2}) + \int_{0}^{\infty} d\lambda g^{2}(\lambda/\alpha)\rho_{(1)}^{h,r}(\lambda) \right).$$
(5.18)

A number of challenges for future research present themselves from the above discussion. We conclude this section with a list of a few more.

- Under the assumption of $a_1 = \cdots = a_N = bN$ with b > 0, verify the sine-kernel in the bulk and Airy-kernel at the soft edge for (1.22) and (4.8) (see recent monographs [6,21,27,58] for the sine and Airy kernels and [48] for recent progress on the random matrix products).
- Under the assumption of $a_{m+1} = \cdots = a_N = bN$ with b > 0, by tuning the parameters a_1, \ldots, a_m verify the BBP transition for (1.22) and (4.8) (cf. [9,55]).
- Verify the transitions from the critical kernels (1.24) and (4.27) to the sine-kernel and to the Airy-kernel (cf. [27, Exercise 7.2] and [29]).

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