Schur expansion of random-matrix reproducing kernels

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Abstract

We give expansions of reproducing kernels of the Christoffel–Darboux type in terms of Schur polynomials. For this, we use evaluations of averages of characteristic polynomials and Schur polynomials in random matrix ensembles. We explicitly compute new Schur averages, such as the Schur average in a $q$-Laguerre ensemble, and the ensuing expansions of random matrix kernels. In addition to classical and $q$-deformed cases on the real line, we use extensions of Dotsenko–Fateev integrals to obtain expressions for kernels on the complex plane. Moreover, a known interplay between Wronskians of Laguerre polynomials, Painlevé tau functions and conformal block expansions is discussed in relationship to the Schur expansion obtained.

Keywords: reproducing kernels, Schur polynomials, Schur expansion, random matrices, $q$-ensembles, $q$-Laguerre, Christoffel–Darboux kernel

(Some figures may appear in colour only in the online journal)

1. Introduction

The study of reproducing kernels [7, 59] is of significance across a remarkably transversal spectrum of areas and applications. Reproducing kernels appear in a crucial way in many different areas including random matrix theory [28], machine learning [43], quantization and the study of coherent states [39], Shannon sampling theorems, non-parametric density estimation in statistics and probability theory [11, 70].

Typical in random matrix theory are the so-called Christoffel–Darboux kernels. The most immediate meaning of these kernels is the description of the correlations between two

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eigenvalues of a random matrix [28]. The reproducing property allows to express higher-order correlation functions in terms of determinants of the two-point kernel, leading to the consideration of a multi-point kernel. In this work we will study the latter, and the method employed lends results for this more general kernel with no additional difficulty. In the simpler setting of the two-point kernel, we will show that the Schur expansion can also be written in terms of Chebyshev polynomials of the second kind.

Christoffel–Darboux kernels are a well-known type of delta sequences and, as such, had found since quite some time important applications in statistics [83]. These kernels are also used in modern machine learning and data-analysis contexts [71], including for example data-driven studies of approximations of the spectrum of the Koopman operator [54]. For these reasons, we will comment at the end how the very recently studied antisymmetric kernels [51] are related to the discussion here (and a complementary discussion in [77]) and to certain specific models discussed in this work, mainly the Stieltjes–Wigert model.

The paper is organized as follows. We start with the basic definitions and introduce notation and conventions. Section 2 presents the main idea of the work, summarized in theorem 2.1. Then, in section 3 we apply the Schur expansion to kernels associated to classical matrix ensembles. For the Laguerre ensemble, a relation with Painlevé V is revisited in section 4, where we also compare our expansion with the conformal block expansion of the kernel. In section 5 we focus on kernels associated to \(q\)-ensembles and in section 6 on kernels on \(C\).

Along the way, we compute the average of a Schur polynomial in the \(q\)-Laguerre ensemble, whose evaluation constitutes a novel result.

We conclude with an outlook on possible avenues for further research in section 7. The text is complemented with three appendices: appendix A contains the technical details of the computation of Schur averages, whereas appendices B and C exemplify how the Schur expansion method transcends the topic of this work and easily finds application in a broad variety of related problems.

1.1. Definitions and notation

1.1.1. Christoffel–Darboux kernels. For any given weight function \(w(z)\) we will denote \(P_n\) the corresponding monic orthogonal polynomial of degree \(n\) and adopt the usual notation \(h_n \equiv \|P_n\|^2 = \langle P_n, P_n \rangle_w\).

We denote by \(N\) the rank of the kernel and by \(n\) the number of pairs of variables \((x_1, \ldots, x_n; y_1, \ldots, y_n)\). Let \(K_N^n(x_1, \ldots, x_n; y_1, \ldots, y_n)\) be the 2\(n\)-point kernel, and \(K_N(x; y) \equiv K_N^1(x; y)\). We adopt the definition

\[
K_N(x; y) = \sum_{j=0}^{N-1} \frac{P_j(x)P_j(y)}{h_j},
\]

as for instance in [80, equation (3.1.9)] and [13, equation (2.2)], and differing by a factor \(\sqrt{w(x)w(y)}\) from [28, equation (5.6)]. The Christoffel–Darboux formula states that

\[
K_N(x, y) = \frac{1}{h_{N-1}} \frac{P_N(x)P_{N-1}(y) - P_{N-1}(x)P_N(y)}{x - y}.
\]

The multi-point kernel satisfies

\[
K_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \frac{1}{\Delta_n(x)\Delta_n(y)} \det K_N(x_i; y_j), \quad 1 \leq i, j \leq n
\]

(1.1)
where
\[ \Delta_n(x) \equiv \prod_{1 \leq i < j \leq n} (x_i - x_j) \]
is the Vandermonde determinant in \( n \) variables (and analogously for \( \Delta_n(y) \)).

We introduce the notation
\[ x^\vee \equiv \left( -\frac{1}{x_1}, \ldots, -\frac{1}{x_n} \right), \quad y^\vee \equiv \left( -\frac{1}{y_1}, \ldots, -\frac{1}{y_n} \right), \]
and often combine these parameters into the \( 2n \)-dimensional vector
\[ t \equiv \left( -\frac{1}{x_1}, \ldots, -\frac{1}{x_n}, -\frac{1}{y_1}, \ldots, -\frac{1}{y_n} \right). \tag{1.2} \]

Besides, to reduce clutter we define
\[ \hat{K}_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \prod_{j=N-n+1}^{N-1} j \prod_{i=1}^{n} (x_i y_i) \frac{N-n}{K_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n)}. \tag{1.3} \]

The kernel (1.1) admits an integral representation (see section 2) in which the number of integration variables is \( N - n \). We will denote \( M \equiv N - n \) the number of variables in a generic ensemble. Moreover, for any given ensemble of \( M \) variables with weight function \( w(z) \), \( Z_M \) denotes the partition function, that is, the \( M \)-fold integral
\[ Z_M \equiv \int \Delta_M(z)^2 \prod_{j=1}^{M} w(z_j) dz_j, \]
and we will use the shorthand notation
\[ N_M \equiv \frac{1}{M! Z_M} \tag{1.4} \]
for the ubiquitous normalization factor.

1.1.2. Partitions. Let \( \mathcal{Y} \) be the set of all partitions,
\[ \mathcal{Y} \equiv \{ \lambda = (\lambda_1, \lambda_2, \ldots) \mid \lambda_j \geq \lambda_{j+1} \geq 0, \forall j \geq 1 \}. \]
The length of a partition \( \lambda \) is \( \ell(\lambda) = \max \{ j : \lambda_j > 0 \} \), and \( |\lambda| = \sum_{j \geq 1} \lambda_j \) is its size. The transposed partition is denoted \( \lambda' \). Besides, for fixed \( L, M \in \mathbb{N} \) we define
\[ \mathcal{Y}_{LM} \equiv \{ \lambda \in \mathcal{Y} \mid \ell(\lambda) \leq L \text{ and } \lambda_1 \leq M \}, \tag{1.5} \]
the set of partitions that are contained in a rectangle of \( L \) rows and \( M \) columns (not to be confused with the partitions of \( L \times M \)).

\( s_\lambda(z_1, \ldots, z_M) \) is the Schur polynomial labelled by the partition \( \lambda \) [57]. For the argument of Schur polynomials, we will often use shorthand notations \( s_\lambda(z) \equiv s_\lambda(z_1, \ldots, z_M) \) and \( s_\lambda(1^M) \equiv s_\lambda(1, \ldots, 1) \).
2. General aspects of the Schur expansion

In this section we set up the method to obtain the Schur expansion of the kernel $K_N(x; y)$ and its multi-point generalization $K_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n)$. Our proof directly gives $\hat{K}_N^{(n)}$, as defined in (1.3), from which the kernel is immediately obtained.

The starting point is the integral representation \[ \hat{K}_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \frac{M!}{\prod_{j=1}^{n}(x_jy_j)}} \int \Delta w(z)^2 \prod_{j=1}^{n} \left( x_j - z_j \right) \left( y_j - z_j \right) w(z_j) dz_j \]

where in the second line we have used the definition (1.2) of $t$. The overall coefficient $N_M$ has been defined in (1.4) and $M = N - n$.

The next step is to use the dual Cauchy identity \[ \frac{2n}{\prod_{j=1}^{n}} \prod_{j=1}^{m} (1 + t_j z_j) = \sum_{\lambda \in \mathbb{V}_{2n,M}} s_\lambda(t)s_\lambda^*(z). \] (2.2)

We have used $t\lambda'_j = \lambda_1$, and $\mathbb{V}_{2n,M}$ is the finite set defined in (1.5).\(^3\) Plugging (2.2) in (2.1) yields the following central result.

**Theorem 2.1.** In the above notation,

\[ \hat{K}_N^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \sum_{\lambda \in \mathbb{V}_{2n,N-n}} s_\lambda(t)s_\lambda^*(w). \] (2.3)

Therefore, whenever the average of a Schur polynomial in the ensemble characterized by the weight function $w(z)$ is known, we get an expansion of $\hat{K}_N^{(n)}$, and thus of $K_N^{(n)}$, in the Schur basis with explicitly known coefficients. The symmetry enhancement \[ S_n \times S_n \times \mathbb{Z}_2 \twoheadrightarrow S_{2n} \] (2.4)

for the parameters $\lambda', y'$ is manifest in the expansion.

According to the Schur-reproducing property [61, 62], each summand in the Schur expansion will take the schematic form $c_{N-n}(\lambda')s_{\lambda'}(1^{N-n})s_\lambda(t)$, for some partition-dependent coefficient $c_{N-n}(\lambda')$. This property of the average is robust under various layers of deformations [62, 63, 66] and is inherited by the corresponding kernel. Concretely,

- In a classical ensemble, $c_{N-n}(\lambda')$ will be a rational function of the rows $\lambda'_j$, typically expressed in terms of $\Gamma$-functions.
- In a $q$-ensemble, $c_{N-n}(\lambda')$ will be a rational function of $q$ with exponential dependence on the rows $\lambda'_j$, possibly times a rational function of the $\lambda'_j$.

\(^3\) As argued in [76, section 5], the dual Cauchy formula (2.2) is an exact polynomial identity, as opposed to the Cauchy identity, that is only algebraic and should be understood in a perturbative sense.
This general observation will be manifest in the explicit results of the following sections.

**Remark 2.2.** The dual Cauchy identity (2.2) yields the expansion in Schur polynomials of the correlation function of characteristic polynomials in any beta-ensemble. However, denoting the beta-parameter by $2\gamma$, it is more convenient in the $\gamma \neq 1$ setup to use the alternative dual Cauchy identity \[58\]

$$\prod_{i=1}^{2n} \prod_{j=1}^{M} \left(1 + t_i z_j \right) = \sum_{\lambda \in \mathcal{Y}_{2n,M}} P^{(\gamma)}_{\lambda}(t) P^{\sqrt{\gamma}}_{\lambda}(z), \quad (2.5)$$

where $P^{(\gamma)}_{\lambda}$ are the Jack polynomials [28, 57]. This aspect is discussed explicitly in subsection 3.3 below.

It follows directly from theorem 2.1 that the two-point kernel $\hat{K}_N$ admits an expansion involving Chebyshev polynomials of the second kind, through the relation

$$\hat{K}_N(x; y) = \sum_{\lambda_1 = 0}^{N-2} \sum_{\lambda_2 = 0}^{\lambda_1} s_\lambda(t_1, t_2) \langle s_{\lambda'} \rangle_w \quad (2.6a)$$

$$= \sum_{\lambda_1 = 0}^{N-2} \sum_{\lambda_2 = 0}^{\lambda_1} \langle s_{\lambda'} \rangle_w \left( \prod_{j=0}^{\lambda_2} \left( t_1^{\lambda_1 - \lambda_2 - j} t_2^j \right) \right) \quad (2.6b)$$

$$= \sum_{\lambda_1 = 0}^{N-2} \sum_{\lambda_2 = 0}^{\lambda_1} \langle s_{\lambda'} \rangle_w (xy)^{-|\lambda|/2} U_{\lambda_1-\lambda_2} \left( \frac{x+y}{2\sqrt{xy}} \right) \quad (2.6c)$$

Equality (2.6a) follows from straightforward computation of the ratio of $2 \times 2$ determinants

$$s_\lambda(t_1, t_2) = \frac{1}{t_1 - t_2} \det \begin{pmatrix} t_1^{\lambda_1+1} & t_2^{\lambda_2+1} \\ t_2^{\lambda_1+1} & t_1^{\lambda_2+1} \end{pmatrix} = \left( t_1 t_2 \right)^{\lambda_1 + \lambda_2} \frac{t_1^{\lambda_1 - \lambda_2 + 1} - t_2^{\lambda_1 - \lambda_2 + 1}}{t_1 - t_2}$$

and (2.6b) by identification with the Chebyshev polynomial of the second kind (cf [47]). Eventually, in (2.6c) we have written the result in the variables $(x; y)$, related to $(t_1, t_2)$ via (1.2).

Note that the $(x; y)$-independent coefficient in the expansion is the same for the Schur and Chebyshev expansion, since the manipulations only involve the symmetric polynomials.

**Remark 2.3.** We emphasize that the two-point kernel admits an expansion in Chebyshev polynomials, whose orthogonality relations are on the interval $[-1, 1]$, regardless of the domain of the variables $x, y$. Indeed, we did not use the orthogonality property of these polynomials in the derivation. Notice however the important extra factor $(xy)^{-|\lambda|/2}$, that is, the dependence is not entirely captured by the Chebyshev polynomial.
From (2.1) we may equivalently expand for \((x_1, \ldots, x_n)\) and \((y_1, \ldots, y_n)\) separately, applying (2.2) twice.

**Theorem 2.4.** In the above notation,

\[
\hat{K}_N^{(\theta)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \sum_{\lambda, \mu \in \mathcal{Y}_n} s_{\lambda}(x) s_{\mu}(y) \langle s_{\lambda} s_{\mu} \rangle_w.
\]  

This latter expansion was first obtained by Rosengren [73, proposition 5]. The equivalence between (2.7) and Rosengren’s formula stems from \(\langle s_\lambda s_\mu \rangle_w\) being equal to the determinant of a Hankel minor, and rearranging the terms in the sum.

Thus, the result of theorem 2.4 is known, but the derivation here is different, only using the dual Cauchy identity. In addition, the identification with Rosengren’s previous result is not completely immediate and it involves the application of Andreief’s identity [5].

In the two-point case, theorem 2.4 implies the following classical result [19, 49] (see also [12, 13, 79]).

**Corollary 2.5.** Let \(H_N(w)\) the \(N \times N\) Hankel matrix of moments of the measure \(w(z)dz\),

\[
[H_N(w)]_{jk} = \int_{\mathbb{R}} z^{j+k} w(z)dz, \quad j, k \in \{0, 1, \ldots, N - 1\},
\]

and denote by \(H_N^{-1}(w)\) its inverse. The kernel \(K_N(x; y)\) is the generating function of \(H_N^{-1}(w)\).

\[
K_N(x; y) = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} x^j y^k [H^{-1}(w)]_{jk}.
\]

**Proof.** Set \(n = 1\) in theorem 2.4. We recognise in the coefficient \(\langle s_{1j} s_{1k} \rangle_w = \langle e_j e_k \rangle_w\) the \((j, k)\)-entry of the inverse matrix \(H^{-1}(w)\), up to a multiplicative factor \(h_{N-1}\). 

A Toeplitz analogue of corollary 2.5 is discussed in appendix B. Both results showing that the two-point kernel is a generating function are well known, and the Schur expansion here simply provides an immediate alternative proof.

3. Kernels corresponding to classical ensembles

Our master formula (2.3) implies that, to obtain the coefficient in the Schur expansion of the kernel, we need to evaluate the average \(\langle s_{\lambda} \rangle_w\) in the ensemble with weight function \(w(z)\). In this section we use known results for the classical ensembles to finalize the computation of the coefficients.

An ensemble is said to be **classical** if the weight function \(w(z)\) satisfies Pearson’s equation

\[
\frac{d}{dz} [\sigma(z) w(z)] = \tau(z) w(z) \quad \text{with} \quad \deg \sigma(z) \leq 2 \quad \text{and} \quad \deg \tau(z) = 1. 
\]

The list goes beyond the typical consideration as classical of many references, oftentimes limited to Gaussian, Laguerre and Jacobi ensembles. This is well-known in the orthogonal polynomials literature [53] and in the study of stationary solutions of stochastic processes [25, 85].
3.1. Gaussian ensemble

Consider the Gaussian unitary ensemble (GUE). The evaluation of the coefficient $\langle s_\lambda \rangle_{\text{GUE}}$ relies on the following result [21].

**Proposition 3.1 (Di Francesco–Itzykson [21]).** Let $\mu$ be a partition with $\ell(\mu)$ even, and denote

$$l_j = \mu_j + \ell(\mu) - j, \quad \forall j = 1, \ldots, \ell(\mu).$$

(3.1)

Consider the GUE ensemble of $M$ variables. Then

$$\langle s_\mu \rangle_{\text{GUE}} = (-1)^{\ell(\mu)} \left( \frac{\ell(\mu) - 2}{\ell(\mu)} \right) \prod_{j \text{ odd even}} \frac{l_o}{l_e} \prod_{\tilde{j} \text{ even}} \frac{\tilde{l}}{\tilde{l} - 1} s_\mu(1^M)$$

(3.2)

if $\ell(\mu)$ is even, or 0 otherwise.

The coefficient in the Schur expansion of the kernel follows from the specialization $\mu = \lambda'$.

3.2. Laguerre ensemble

Consider the Laguerre unitary ensemble (LUE), with weight function $w(z) = z^\alpha e^{-z} \mathbb{1}_{z > 0}$, $\alpha > -1$.

**Proposition 3.2.** Consider the LUE ensemble of $M$ variables and let $\mu$ be a partition with $\ell(\mu) \leq M$. Then,

$$\langle s_\mu \rangle_{\text{LUE}} = \prod_{j=1}^{M} \frac{\Gamma(\alpha + \mu_j + M + 1 - j) \Gamma(\alpha + M + 1 - j)}{\Gamma(\alpha + 1)} s_\mu(1^M).$$

(3.3)

The evaluation of the coefficient in the Schur expansion follows from the specialization $M = N - n$ and $\mu = \lambda'$. For completeness, we reproduce a proof of proposition 3.2 due to [35, section 4.2.2] in appendix A.1.

For the special case $\alpha \in \mathbb{N}_0$, (3.3) can be recast in equivalent forms:

$$\langle s_\mu \rangle_{\text{LUE}, \alpha \in \mathbb{N}_0} = \left[ \prod_{j=1}^{M} (\alpha + M - j)^{-1} \right] \langle s_{\mu + (\alpha M)} \rangle_{\text{LUE}, \alpha = 0}$$

(3.4a)

$$= s_\mu(1^{M+\alpha}) \prod_{j=1}^{M} \frac{\Gamma(\mu_j + M + 1) \Gamma(M + 1 - j)}{\Gamma(\mu_j - j + M + 1)}.$$

(3.4b)

The first identity (3.4a) is straightforward from (3.3), and can be alternatively derived from the integral representation absorbing the $\alpha$-dependent part of the Laguerre weight into the Schur polynomial, through the property

$$\left( \prod_{j=1}^{M} z_j^{\alpha} \right) s_\mu(z) = s_{\mu + (\alpha M)}(z).$$

The $\alpha$-dependent but $\mu$-independent coefficient in (3.4a) is entirely due to the denominator.
Using the dimension formula on \( s_{\mu}(1^{M+\alpha}) \) and splitting the product in the three regions: (i) \( j < k \leq M \), (ii) \( j \leq M \) with \( k > M \) and (iii) \( M < j < k \) gives

\[
s_{\mu}(1^{M+\alpha}) = s_{\mu}(1^M) \prod_{j=1}^{M} \prod_{k=M+1}^{M+\alpha} \frac{\mu_j - j + k}{k - j} \\
= s_{\mu}(1^M) \prod_{j=1}^{M} \frac{\Gamma(\mu_j - j + M + \alpha + 1) \Gamma(M + 1 - j)}{\Gamma(\mu_j - j + M + 1) \Gamma(M + \alpha + 1 - j)},
\]

using the hypothesis \( \ell(\mu) \leq M \). When plugged in (3.3), this proves (3.4b).

Expressions for Schur averages closely related to (3.3) at \( \alpha = M \) have been given in \[69\] (see also \[20, 68\] for related discussion), except that in \[69\] the argument of \( s_{\mu} \) in the average is inverted, \( s_{\mu}(z_1^{-1}, \ldots, z_M^{-1}) \). Recall that \( s_{\mu}(1^M) \) gives the dimension of a \( U(M) \) representation labelled by \( \mu \), but the same partition can label a representation of the symmetric group \( S_{|\mu|} \). Writing \( s_{\mu}(1^M) \) as a function of the dimension of the \( S_{|\mu|} \) representation recasts (3.3) is a form very similar to \[69\].

**Remark 3.3.** In \[17, equation (96)\] the average \( \langle s_{\mu} \rangle_{\text{JUE}} \) at \( \alpha = 0 \) is given as \( (s_{\mu}(1^M))^2 \), in conflict with our computation. A check in Mathematica for \( M = 2, 3, 4 \) and a small sample of partitions \( \mu \), however, shows agreement with formula (3.3), while \[17, equation (96)\] fails. As further confirmation, the computations in section 4, that rely on (3.3), are consistent with the existing literature.

### 3.3. Jacobi ensemble

Let us now take the (asymmetric) Jacobi ensemble with weight function \( w(z) = z^\alpha (1 - z)^\beta 1_{0 < z < 1} \), \( \alpha > -1, \beta > -1 \). The following holds, see e.g. \[33\].

**Proposition 3.4.** Consider the JUE ensemble of \( M \) variables and let \( \mu \) be a partition with \( \ell(\mu) \leq M \). Take, \( \alpha, \beta \in \mathbb{N}_0 \). Then,

\[
\langle s_{\mu} \rangle_{\text{JUE}} = \frac{s_{\mu}(1^M) s_{\mu}(1^{\alpha+M})}{s_{\mu}(1^{\alpha+\beta+2M})}.
\]

Formula (3.5) can be equivalently written as

\[
\langle s_{\mu} \rangle_{\text{JUE}} = \prod_{j=1}^{M} \frac{\Gamma(\mu_j - j + \alpha + M + 1) \Gamma(\alpha + \beta + 2M + 1 - j)}{\Gamma(\mu_j - j + \alpha + \beta + 2M + 1) \Gamma(\alpha + M + 1 - j)} s_{\mu}(1^M).
\]

The equality between (3.5) and (3.6) is shown using the dimension formula for the Schur polynomials. Many simplifications take place separating the double products, both in the numerator and in the denominator, in the three regions: (i) \( j < k \leq M \), (ii) \( j \leq M \) and \( k > M \), and (iii) \( M < j < k \).

The average of the Schur polynomial in the form (3.6) can be analytically continued to non-integer values of \( \alpha, \beta \). Since both sides of the equality (3.6) depend analytically on \( \alpha, \beta \) in a suitable region, we expect the result (3.6) to hold when the restriction \( \alpha, \beta \in \mathbb{N}_0 \) is lifted.
3.3.1. Jacobi beta-ensemble. As we have pointed out in remark 2.2, the strategy can be applied to any beta-ensemble, with beta-parameter denoted by $2\gamma$. It is convenient, however, to use the dual Cauchy identity (2.5) in terms of Jack polynomials. We are thus led to

$$\langle \prod_{\lambda=1}^{2n} \det(1 + tZ) \rangle_{\text{JE},2\gamma} = \sum_{\lambda \in \mathcal{V}_{2n,M}} P_{\lambda}^{(1/\gamma)}(t) < P_{\lambda}^{(1/\gamma)} >_{\text{JE},2\gamma},$$

where $\langle \cdot \rangle_{\text{JE},2\gamma}$ is the average taken in the Jacobi ($2\gamma$)-ensemble. The advantage of this expansion is that the Jack polynomial average is known [48]:

$$\langle P_{\lambda}^{(1/\gamma)} \rangle_{\text{JE},2\gamma} = \frac{2n}{(\alpha + \beta + 2 + \gamma(2M - j - 1))\lambda'} \prod_{1 \leq j < k \leq \lambda_1} \frac{(\gamma(k - j + 1))\lambda'_{j-k} \lambda_{j-k} \lambda_{j-k} (\gamma(M - j + 1))\lambda_{j-k}}{(\gamma(k - j))\lambda'_{j-k} \lambda_{j-k} \lambda_{j-k} (\gamma(M - 1))\lambda_{j-k}},$$

where $\lambda'$ is the Pochhammer symbol, and $\ell(\lambda') = \lambda_1$ has been used.

Jack polynomial averages have been extensively used in [34, 60] to compute (both positive and negative) moments in the Jacobi and Laguerre matrix beta-ensemble.

3.4. Other classical ensembles

The weight function $w(z) = \frac{\alpha}{(1 + z)^{\alpha+\beta}}$ also defines a classical ensemble since it satisfies Pearson’s equation (see for example [53]). A Brownian motion interpretation of this classical ensemble has been recently given in [10, remark 2.5]. The corresponding ensemble is related to a Jacobi ensemble, in the sense that both weight functions are related to Euler’s beta integral

$$B(a, b) \equiv \int_0^\infty \frac{z^a}{(1 + z)^{a+b}} \, dz = \int_0^1 u^{a-1}(1 - u)^{b-1} \, dz = \frac{\Gamma(a) \Gamma(b)}{\Gamma(a + b)},$$

(3.7)

with $\Re a, \Re b > 0$. We denote such ensemble by JUE.

However, as pointed out in [33], whereas a change of variables immediately relates the partition functions of the two ensembles, the same is not true if we have Schur insertions, since the argument of the Schur polynomial changes. Hence, the computation of the Schur average when $\alpha, \beta \in \mathbb{N}_0$ is similar to but intrinsically different from the one for the Jacobi ensemble of proposition 3.4 (cf [33]).

**Proposition 3.5.** Consider the JUE ensemble of $M$ variables and let $\mu$ be a partition with $\ell(\mu) \leq M$. Take, $\alpha, \beta \in \mathbb{N}_0$ with $\beta \equiv \beta - \alpha - M \geq 0$. Then,

$$\langle s_\mu \rangle_{\text{JUE}} = \frac{s_\mu(1^M)\lambda_0(1^{\alpha+M})}{s_\mu(1^\beta)}.$$  

(3.8)

A further classical ensemble to be considered is the one with weight function $w(z) = z^{-\alpha} e^{-1/z} \mathbb{1}_{0<z<\infty}$. It corresponds to the so-called inverse Wishart ensemble [38] and we schematically denote it as LUE. Probabilistically, its weight function is the probability density function (pdf) of the inverse Gamma distribution. Using the results for the Laguerre ensemble we can solve this case as well, but the relationship between both cases is non-trivial due to the presence of the Schur polynomial.
We rewrite
\[ \tilde{\alpha} = \alpha + 2M - 1 \] (3.9)
and ask \( \alpha > -1 \), which is sufficient for convergence at \( z \to \infty \).

A change of variables \( z_j' = z_j^{-1} \) yields
\[
\Delta M(z)^2 \prod_{j=1}^{M} \frac{e^{-z_j'}}{z_j'} dz_j = (-1)^M \Delta M(z_j')^2 \prod_{j=1}^{M} z_j' e^{-z_j'} dz_j.
\]
(We henceforth drop the primes). In other words, the joint pdf of the \( M \) eigenvalues in the \( \tilde{LUE} \) can be mapped to that of the LUE. The overall \((-1)^M\) will cancel when computing averages.

Things become more involved for the average of a Schur polynomial, because the change of variables inverts the argument:
\[
\langle s_{\lambda'} \rangle_{\tilde{LUE}, \tilde{\alpha}} = \mathcal{N}_{M, \text{LUE}} \int_{(0, \infty)^M} s_{\lambda'}(z^{-1}) \prod_{j=1}^{M} w_{\text{LUE}, \alpha}(z_j) dz_j.
\]
Besides, for the original integral to converge we must impose \( \alpha > \ell(\lambda) - 1 \). Recalling that this average appears in the sum (2.3), we impose \( \alpha > 2n - 1 \). We use the property \([57]\)
\[
s_{\lambda'} \left( \frac{1}{z_1}, \ldots, \frac{1}{z_M} \right) = s_{\lambda'}(z_1, \ldots, z_M) \prod_{j=1}^{M} z_j^{-(\ell(\lambda))},
\]
where we have employed \( \lambda'_j = \ell(\lambda) \), and the partition \( (\lambda')^* \) is
\[
(\lambda')^* = (\ell(\lambda) - \lambda'_M, \ell(\lambda) - \lambda'_{M-1}, \ldots, 0).
\]
We conclude that
\[
\langle s_{\lambda'} \rangle_{\tilde{LUE}, \tilde{\alpha}} = \langle s_{(\lambda')^*} \rangle_{\text{LUE}, \alpha - \ell(\lambda)}, \tag{3.10}
\]
with the relation between \( \tilde{\alpha} \) and \( \alpha \) as in (3.9). Since the Schur average in the LUE is known from (3.3), the Schur average in this other classical ensemble follows.

Precisely, the multi-point kernel of the inverse Wishart ensemble is studied in [38], where it is interpreted as the probability distribution of non-interacting fermions in a background potential. Theorem 2.1 combined with (3.10) and (3.3) yields a character expansion of such quantity.

**Remark 3.6.** We have collected Schur averages for classical ensembles from the literature, with the exception of eventual rephrasing of some expressions and our result (3.10). In section 5.2, we will compute the average in a \( q \)-Laguerre ensemble.

\(^4\) Analogous computations have been applied in [74, 76] to study loop operators in 3D Chern–Simons–matter theories.
4. Laguerre Wronskian, conformal block expansion and tau function of Painlevé V

As an application of our results, we will exploit now the fact that the diagonal of the kernels are directly related to Wronskians of orthogonal polynomials. In turn, these Wronskians appear in webs of relationships that oftentimes include Painlevé tau functions. We show here how the diagonal limit of the Laguerre expansion obtained before is related to the random matrix results in [9, 84].

The diagonal limit of the Christoffel–Darboux kernel, when all variables are the same,

\[ K_N^{(a)}(x; x, \ldots; x) \equiv \lim_{n \to \infty} K_N^{(a)}(x_1, \ldots, x_n; y_1, \ldots, y_n) \]

corresponds to the moment of order 2n of the characteristic polynomial average in the corresponding random matrix ensemble:

\[ \frac{1}{\prod_{j=N-n+1}^{N-1} x_j} K_N^{(a)}(x, \ldots, x) = \langle \det (x - Z)^{2n} \rangle_w, \quad (4.1) \]

where Z is a random matrix from the ensemble with weight function \( w(z) \) in \( N - n \) variables. The random matrix average (4.1) is a \( 2n \times 2n \) Wronskian determinant built with the orthogonal polynomials \( P_z(x) \) associated to \( w(z) \) [16, equation (15)]

\[ \langle \det (x - Z)^{2n} \rangle_w = \frac{1}{G(2n + 1)} \text{Wr} \left( P_{N-n}^{(a)}(x), P_{N+1-n}^{(a)}(x), \ldots, P_{N+(n-1)}^{(a)}(x) \right). \quad (4.2) \]

Typically, the Wronskian is equivalently written as a Hankel determinant involving different orthogonal polynomials [50, 55].

In the case of the Laguerre ensemble, the average (4.1) plays a prominent role and is related to a wide range of quantities [26]. The Hankel equivalent was studied in detail in [84], and later on in [9].

It is convenient for the rest of the subsection to adopt the shorthand

\[ \tilde{K}_N^{(a)}(x; x, \ldots; x) \equiv \frac{1}{\prod_{j=N-n+1}^{N-1} x_j} K_N^{(a)}(x, \ldots; x; x, \ldots; x) \big|_{\text{LUE}, \alpha = 2n} \]

Then

\[ \tilde{K}_N^{(a)}(x; x, \ldots; x) \equiv \langle \det (x - Z)^{2n} \rangle_{\text{LUE}, \alpha = 2n} \]

\[ \equiv (-1)^n \frac{G(M + 2n + 1)}{G(2n + 1)G(M + 2)} \text{Wr} \left( L_{N-n}^{(a=2n)}(x), L_{N+1-n}^{(a=2n)}(x), \ldots, L_{N+(n-1)}^{(a=2n)}(x) \right), \quad (4.2) \]

where \( L_z^{(a)} \) are the generalized Laguerre polynomials and the additional Barnes G-functions [28] on the second line come from passing from monic to orthonormal polynomials. Note that we are pegging the value of the parameter \( \alpha \) with the order of the moment of the characteristic polynomial, \( \alpha = 2n \).
Moreover, following [9], let us introduce the function
\[
f_{2n}(x) = e^{\frac{M}{2}x}(-1)^n \text{Wr}\left(L_{N-n}^{(\alpha=2n)}(-x), L_{N+1-n}^{(\alpha=2n)}(-x), \ldots, L_{N+n}^{(\alpha=2n)}(-x)\right)
\]
\[
= \frac{G(2n+1)G(M+2)}{G(M+2n+1)} e^{\frac{M}{2}x} \tilde{K}_N^{(\alpha)}(-x)_{\text{LUE}}.
\]

Then, combining the above discussion with [84, proposition 3] yields another expression for the Laguerre kernel.

**Proposition 4.1 (Winn [84]).** Let \(\zeta \in \mathbb{R}\) and set \(x = 2|\zeta|\). Consider the LUE of \(M = N - n\) variables with \(\alpha = 2n\). The associated kernel satisfies the identity
\[
f_{2n}(x) = \frac{2^{M^2+4Mn}}{(2\pi)^M M!} \int_{\mathbb{R}^M} \Delta_M(z)^2 \prod_{j=1}^M \frac{e^{i\zeta^j}}{(1 + z_j^2)^{M+2n}} \, dz_j,
\]
with \(f_{2n}\) related to the LUE kernel via (4.3).

Proposition 4.1 is just a rephrasing of [84, proposition 4], but here we are emphasizing the interpretation as a kernel for the LUE.

The Wronskian of Laguerre polynomials is in turn related to a solution to the \(\sigma\)-Painlevé V equation [9, theorem 1]. The result of interest for us is that the function \(f_{2n}(x)\) is essentially a tau function of Painlevé V. This implies a connection between the kernel of the LUE, with \(\alpha = 2n\), and tau functions of Painlevé V. The mapping between the two settings is non-linear, and indeed requires the term \(e^{\frac{M}{2}x}\).

Combining the present discussion with the Schur expansion, using the explicit knowledge of the coefficient from (3.4b) with \(\alpha = 2n\), we derive a new expansion for \(f_{2n}(x)\):
\[
f_{2n}(x) = \frac{G(2n+1)}{G(M+2n+1)} e^{\frac{M}{2}x} \sum_{\lambda \in \mathbb{Y}_{2n,M}} \sum_{j=1}^M \Gamma\left(\lambda_j - j + M + 1\right) \times \prod_{j=1}^M \Gamma\left(\lambda_j - j + M + 1\right).
\]

The sum runs over partitions contained in a \(2n \times M\) rectangle, thus the overall factor \(x^{2Mn}\) guarantees that only non-negative powers of \(x\) are included. Explicitly, expanding the exponential and rearranging the terms, we find
\[
f_{2n}(x) = \frac{G(2n+1)}{G(M+2n+1)} \sum_{k=0}^{\infty} x^k \sum_{m=0}^{k} \sum_{\lambda \in \mathbb{Y}_{2n,M}} \sum_{|\mu| = m} \left(\frac{-M}{2}\right)^{k-m} \frac{1}{(k-m)!} 
\]
\[
\times s_{(M^2n),\mu}(1^{2n}) s_{(2n)^M,\mu}(1^{2n+M}) \prod_{j=1}^M \Gamma\left(2n + M + 1 - j - \mu_{M-j+1}\right).
\]

We have replaced the sum over \(\lambda\) by a sum over partitions \(\mu\) such that \(\lambda\) is realised subtracting \(\mu\), rotated by 180°, from the bottom-right corner of the \(2n \times M\) rectangular partition \((M^{2n})\).
Then, $\lambda'$ is realised by subtracting $\mu'$ to the rectangular partition $((2n)^M)$ in the same way. For example, for $M = 6$ and $2n = 4$,

\[
\lambda = (6^2, 5, 3) \quad \mu = (3, 1)
\]

\[\rightarrow\]

\[
\lambda' = (4^3, 3^2, 2) \quad \mu' = (2, 1^2)
\]

In particular, writing as in [9]

\[f_{2n}(x) = f_{2n}(0) \left[ 1 + b_1 x + b_2 x^2 + O(x^3) \right],\]

from (4.4) we find

\[f_{2n}(0) = s_{(2n)^M}(1^{M+2n}) \]

\[= G(M + 1) \frac{G(4n + M + 1)G(2n + 1)^2}{G(4n + 1)G(2n + M + 1)^2}\]

and

\[b_1 = \frac{\mu = 0}{2} + \frac{1}{2n} \frac{s_{(M^{2n})\setminus(1)}(1^{2n})s_{((2n)^M\setminus(1))(1^{M+2n})}}{s_{(2n)^M}(1^{M+2n})} = 0,\]

\[b_2 = \frac{M^2}{8} - \frac{M}{2n} \frac{s_{(M^{2n})\setminus(2)}(1^{2n})s_{((2n)^M\setminus(2))(1^{M+2n})}}{s_{(2n)^M}(1^{M+2n})} = MN\]

\[+ \frac{1}{2n(2n - 1)} \frac{s_{(M^{2n})\setminus(1)}(1^{2n})s_{((2n)^M\setminus(2))(1^{M+2n})}}{s_{(2n)^M}(1^{M+2n})} = - \frac{M(M + 4n)}{8(4n + 1)(4n - 1)} \]
in perfect agreement with [9].

The tau function of Painlevé V, and thus \( f_{2n} \), admits an expansion in conformal blocks [56], involving a differently organized sum over partitions. Comparing our formula with the conformal block expansion, we make the following observations.

- In the spirit of theorem 2.1, we get a sum over a single partition, as opposed to the sum over pairs of partitions in the conformal block expansion.
- The Wronskian presentation (4.2) makes it manifest that the dependence on \( x \) must be a polynomial times the exponential factor \( e^{-\frac{M}{2}x} \). This property is explicit in the Schur expansion (4.4), as opposed to the conformal block expansion. Indeed, the sum over partitions in (4.4) terminates.
- The conformal block expansion makes the dependence on the parameters of the Painlevé equation more manifest.

In summary, our expansion is meant for the pure Wronskian part, whereas the conformal block and related expansions [9, 56] are suitable for Painlevé transcendents and tau functions.

5. Kernels corresponding to \( q \)-ensembles

Throughout the present section we compute the coefficients in the Schur expansion of the kernel for a selected class of \( q \)-ensembles. The definition of \( q \)-ensemble is the one put forward in [67], namely, a standard random matrix ensemble whose weight function is such that its associated orthogonal polynomials are \( q \)-deformed. We will focus on the Stieltjes–Wigert ensemble, which has multitude of physical applications [27, 30, 74, 81, 82], and its one-parameter generalization, the \( q \)-Laguerre ensemble, which of course is also a one-parameter generalization of the Laguerre ensemble studied above.

5.1. Stieltjes–Wigert ensemble

Consider the Stieltjes–Wigert ensemble, whose weight function is

\[
w(z) = \frac{1}{\sqrt{2\pi g}} e^{-\frac{1}{2} g (\ln z)^2} \mathbf{1}_{0 < z < \infty}, \quad g > 0.\]

It is customary to introduce the \( q \)-parameter \( q = e^{-g} \). To set the notation, we define the symmetric \( q \)-number as

\[
[z]_q = \frac{q^{z/2} - q^{-z/2}}{q^{1/2} - q^{-1/2}}.
\]  

(5.1)

The \( q \)-dimension of a \( U(M) \) representation labelled by a partition \( \mu \) is

\[
\dim_q \mu = \prod_{1 \leq j < k \leq M} \frac{[\mu_j - j - \mu_k + k]_q}{[j - k]_q}.
\]

(5.2)

**Proposition 5.1.** Consider the Stieltjes–Wigert ensemble of \( M \) variables and let \( \mu \) be a partition with \( \ell(\mu) \leq M \). Then

\[
\langle s_\mu \rangle_{SW} = q^{\frac{1}{2} \sum_{j=1}^{M} \mu_j (3M+1-2j)} \dim_q \mu.
\]

(5.3)

This result already appears in [22], and we give an alternative proof in appendix A.2.

It is worth mentioning that, as for the Laguerre ensemble discussed in section 4, the diagonal limit of the multidimensional Stieltjes–Wigert kernel can be related to other objects previously studied in the literature. In this case, with the partition function of a model of fermions with a
large non-abelian symmetry [6, 37, 82]. Indeed, the diagonal limit of the kernel is proportional to the Wronskian of Stieltjes–Wigert polynomials, which was shown in [82] to compute the partition function $Z_{\text{fermion}}(x)$ of the fermion model. Therefore, we have the character expansion

$$Z_{\text{fermion}}^{\text{M},2n}(x) = \frac{1}{M!} \int_{(0,\infty)^M} \Delta(z)^2 \prod_{j=1}^{M} (x - z_j)^{2n} e^{-\frac{1}{4\pi i} \log z_j^2} \frac{dz_j}{\sqrt{2\pi i}}$$

$$= \prod_{j=0}^{N+n-1} h_j \prod_{j=0}^{N-n-L} h_j K_{N}^{(n)}(x; \ldots, x; x, \ldots, x)_{\text{SW}}$$

$$= Z_M^{\text{SW}}(q) \sum_{\lambda \in \mathcal{Y}_{2n,M}} x^{2nM-|\lambda|} q^{\frac{M+1}{2}(|\lambda| + \sum_{j=1}^{M} \left( \frac{4}{3}|\lambda| + j \right))} \dim \lambda \dim_q \lambda',$n

where $x$ is a spectral parameter in the theory [37, 82], $\dim \lambda = s_{\lambda}(1^{2n})$ is the dimension of the $U(2n)$ representation labelled by $\lambda$ and

$$Z_M^{\text{SW}}(q) = \prod_{j=1}^{M-1} \Gamma_q(1 + j) \big( 1 - q^{(M+1)/2} \big) q^{\frac{M+1}{2} \sum_{j=1}^{M} \left( \frac{4}{3}|\lambda| + j \right)}$$

is the Stieltjes–Wigert partition function. In this way, the setup of the present work provides a new interpretation of the physically meaningful quantity $Z_{\text{fermion}}(x)$, together with its character expansion, in the spirit of [76]. Being the Schur expansion of $Z_{\text{fermion}}(x)$ a generating polynomial in the fugacity $x$, the expression is especially well-suited for comparison with analogous observables in distinct theories, for instance with the aim of testing dualities.

Next we discuss the case of the $q$-Laguerre ensemble, which has not been studied before in this context and has the interesting feature that it generalizes the Stieltjes–Wigert ensemble (a $q$-ensemble) while also being a $q$-deformation of the classical Laguerre ensemble. We will be checking out these generalizing features of the ensemble, by taking the appropriate limits.

5.2. $q$-Laguerre ensemble

Consider the $q$-Laguerre ensemble, with weight function $w(z) = \frac{z^n}{(1-z)_{\infty}^q}$, with $(::)_{\infty}$ the $q$-Pochhammer symbol and $0 < q < 1$. The corresponding orthogonal polynomials have been constructed by Moak [64], thus the kernel can be obtained explicitly from the Christoffel–Darboux formula.

Here we instead compute the average of a Schur polynomial in the $q$-Laguerre ensemble, providing a $q$-deformation of the result in proposition 3.2. Note that our definition of $q$-Laguerre weight is as in [64], and differs from [52, section 14.2] and [18] by a normalization factor, as in these references the denominator is $(-z; q)_{\infty}$.

**Proposition 5.2.** Consider the $q$-Laguerre ensemble of $M$ variables and let $\mu$ be a partition with $\ell(\mu) \leq M$. Then

$$\langle s_{\mu} \rangle_{\text{LUE}} = q^{\frac{1}{2} \sum_{j=1}^{M} \mu_j (M+1)} \dim_q \mu \times \prod_{j=1}^{M} \frac{\Gamma(\alpha + 1 + \mu_j + M - j) \Gamma(-\alpha - \mu_j - M + j)}{\Gamma(\alpha + j) \Gamma(-\alpha - j + 1)} \frac{\Gamma_q(-\alpha + j + 1) \Gamma_q(-\alpha - \mu_j - M + j)}{\Gamma_q(-\alpha - j + 1) \Gamma_q(-\alpha - \mu_j - M + j)}.$$  

(5.4)
The result \((\mathcal{S}_M)\) is new, to our knowledge, and we give a proof in appendix A.3. It is straightforward to see that \((\mathcal{S}_M)\) converges to \((\mathcal{S}_M)\) in the \(q \to 1\) limit. For \(\alpha \notin \mathbb{Z}\) the expression \((\mathcal{S}_M)\) may be simplified using Euler’s reflection property \(\Gamma(1-z)\Gamma(z) = \frac{\pi}{\sin(\pi z)}\), \(z \notin \mathbb{Z}\). In turn, for \(\alpha \in \mathbb{N}_0\), property \((\mathcal{S}_M)\) carries over directly to the \(q\)-deformed setting.

It has been shown by Askey [8] that the \(q\)-Laguerre polynomials, upon changing the weight to \(w(\alpha) = \frac{q^\alpha}{\Gamma(q^\alpha)}\) and scaling the variable \(z \mapsto q^{-\alpha}z\), converge to the Stieltjes–Wigert polynomials in the \(\alpha \to \infty\) limit. We now discuss the implications for the kernel as seen from the Schur expansion.

(a) First, we pass from Moak’s [64] to Askey’s [8] normalization. We can either

1. redo the computations in appendix A.3 with modified moments, or

2. we can simply notice that, when computing the average \(\langle \mathcal{S}_M \rangle\) all prefactors will cancel against the normalization except for the one coming from a rescaling of the variables in \(s_j(z)\).

In both ways we find that we must include a term \((1 - q)^{[\mu]}\) in the average of the Schur polynomial.

(b) We then take the large \(\alpha\) limit of the \(\alpha\)-dependent but \(q\)-independent part in \((\mathcal{S}_M)\). It is convenient to take the limit \(\alpha \to \infty\) with \(\alpha \notin \mathbb{Z}\), in which case

\[
\lim_{\alpha \to \infty} \prod_{j=1}^M \frac{\Gamma(\alpha + 1 + \mu_j + M - j)}{\Gamma(\alpha + j)} = \frac{(-1)^{\mu} \sin(\pi \alpha)}{\sin(\pi (\alpha + j + M))} = \prod_{j=1}^M (-1)^{\mu_j + M - 1} = (-1)^{[\mu]}. \tag{5.5}
\]

(c) We use basic properties of the \(\Gamma_q\) function to rewrite \((\mathcal{S}_M)\) in a more suitable form:

\[
\prod_{j=1}^M \frac{\Gamma_q(-\alpha - j + 1)}{\Gamma_q(-\alpha - \mu_j - j + M)} = (-q)^{(-1)^{[\mu]}} \prod_{j=1}^M \frac{\Gamma_q(\alpha + \mu_j - j + M)}{\Gamma_q(\alpha + j - 1)}. \tag{5.6}
\]

(d) To take \(\alpha \to \infty\) in the product in \((\mathcal{S}_M)\), we use Moak’s \(q\)-analogue of Stirling’s formula [65]. It gives

\[
\prod_{j=1}^M \frac{\Gamma_q(\alpha + \mu_j - j + M)}{\Gamma_q(\alpha + j - 1)} \approx (1 - q^{-1})^{-[\mu]} \prod_{j=1}^M \frac{(-q)^{[\mu]}}{(-q)^{[\mu]} \Gamma_q(\alpha + j - 1) \Gamma_q(\alpha + j - 1 + M - 1)} = (-1)^{[\mu]} (1 - q^{-1})^{-[\mu]} q^{-\sum_{j=1}^M (\mu_j + 2M - 1)} q^{-\alpha[\mu]}. \tag{5.7}
\]

The overall sign \((-1)^{[\mu]}\) cancels against \((\mathcal{S}_M)\), while combining the factor \((1 - q^{-1})^{-[\mu]}\) with \((-q)^{[\mu]}\) from \((\mathcal{S}_M)\) cancels the factor from the change of weight (cf step (a)).

(e) Putting all the pieces together and comparing with \((\mathcal{S}_M)\), we arrive at

\[
\lim_{\alpha \to \infty} q^{(\alpha - 1)[\mu]} \langle s_j \rangle_{q \text{LUE}} = \langle s_j \rangle_{\text{SW}}. \tag{5.8}
\]
(f) So far we have discussed the $\alpha \to \infty$ limit of the Schur average alone. We additionally impose the scaling $(x_1, \ldots, x_n; y_1, \ldots, y_n) \mapsto (q^{-1/\alpha}x_1, \ldots, q^{-1/\alpha}x_n; q^{-1/\alpha}y_1, \ldots, q^{-1/\alpha}y_n)$, so that a factor $q^{\alpha(n-1)/N}$ comes from the Schur polynomial, cancelling the prefactor in (5.8)\(^5\).

6. Kernels on $\mathbb{C}$ and generalizations of Dotsenko–Fateev integrals

The Schur expansion can be applied to reproducing kernels on $\mathbb{C}$, such as Bergman kernels and kernels on Bargmann–Fock spaces. Because kernels in a complex space setting are, in quite a few instances, of a very simple nature, in those cases the type of expansion discussed above will not yield an alternative expression. We show this explicitly with the straightforward case of the Ginibre ensemble. Even in more complicated scenarios, as for example with polyanalytic Bargmann–Fock spaces [1, 41], in some cases (but certainly not all) the kernel itself may be as simple as a single Laguerre polynomial [1, 41].

For non-trivial results, we focus here on other models, and we discuss thoroughly a Dotsenko–Fateev ensemble as our main example.

The starting point of our analysis is the representation of the kernel as the average of a product of characteristic polynomials, that generalizes the real case. We will need a particular Dotsenko–Fateev ensemble as our main example.

To set the stage, let $(x_1, \ldots, x_n) \in \mathbb{C}^n$ and $(y_1, \ldots, y_n) \in \mathbb{C}^n$, $n \geq 1$. The $x_i$ will be related to the holomorphic sector and the $y_j$ to the anti-holomorphic sector. As a consequence, the kernel will depend holomorphically on the $x_i$ and anti-holomorphically on $y_j$. The notation (1.1) is then extended to the complex case as

$$K_N^{(\alpha)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \frac{1}{\Delta_n(x)\Delta_n(y)} \det_{1 \leq i, j \leq n} [K_N(x_i, y_j)].$$

The extension of the integral representation (2.1) to complex ensembles is as follows [3].

**Proposition 6.1 (Akemann–Vernizzi [3]).** With the notation as in section 2,

$$K_N^{(\alpha)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \frac{N_M}{\prod_{j=0}^{N(M)}} \int \Delta_M(z)\Delta_M(\bar{z}) \prod_{j=1}^{M} \prod_{j=1}^{n} (x_i - z_j) (y_i - \bar{z}_j) w(z_j, \bar{z}_j) dz_j d\bar{z}_j.$$

As opposed to the real case, the Schur expansion of the kernel on $\mathbb{C}$ will require two Schur polynomials: one for the holomorphic and one for the anti-holomorphic sector. Therefore, the Schur expansion is akin to theorem 2.4. Applying the dual Cauchy identity (2.2) twice we get

$$\hat{K}_N^{(\alpha)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \sum_{\lambda, \mu \in Y_{n,M}} s_{\lambda}(x^\vee) s_{\mu}(y^\vee) \langle s_{\lambda} s_{\mu} \rangle_w,$$

where

$$\langle s_{\lambda} s_{\mu} \rangle_w \equiv N_M \int \Delta_M(z)\Delta_M(\bar{z}) |z^\lambda(\bar{z})|^2 w(z_j, \bar{z}_j) dz_j d\bar{z}_j.$$

\(^5\)We need to take the scaling $x \mapsto q^{-1/\alpha}x$, instead of Askey’s $x \mapsto q^\alpha x$. This seems to be due to the fact that Askey showed convergence to a different weight with same associated system of orthogonal polynomials.
Notice the lack of symmetry enhancement (2.4) in the complex case, due to one set of variables appearing holomorphically and the other anti-holomorphically.

### 6.1. Ginibre ensemble

The first example we consider is the complex Ginibre ensemble. The corresponding monic orthogonal polynomials on \( \mathbb{C} \) are the monomials, \( P_j(z) = z^j \) with \( h_j = j! \). Therefore, with the notation (1.3),

\[
\hat{K}_N(x;\bar{y})|_{\text{Gin}} = (N - 1)! \sum_{j=0}^{N-1} \frac{(xy)^{j-N+1}}{j!}. \tag{6.1}
\]

To compare with its expansion in the basis of Schur polynomials, we use [31, equation (3.16)]

\[
\langle s_{\lambda'} \bar{s}_{\mu'} \rangle_{\text{Gin}} = \delta_{\lambda'} \prod_{j=1}^{M} \Gamma \left( M - j + 1 + \lambda'_j \right) \Gamma \left( M - j + 1 \right).
\]

The double sum reduces to a single sum,

\[
\hat{K}^{(n)}(x_1, \ldots, x_n; \bar{y}_1, \ldots, \bar{y}_n)|_{\text{Gin}} = \sum_{\lambda \in \mathcal{Y}_{n,\bar{N}-n}} s_{\lambda}(x) s_{\lambda}(\bar{y}) \langle s_{\lambda'} \bar{s}_{\lambda'} \rangle_{\text{Gin}}.
\]

For the particular case \( n = 1 \), using

\[
\lambda'_j = \begin{cases} 
1 & 1 \leq j \leq \lambda_1 \\
0 & \text{otherwise}
\end{cases} \quad \Rightarrow \quad \prod_{j=1}^{M} \frac{\Gamma \left( N - j + \lambda'_j \right)}{\Gamma \left( N - j \right)} = \prod_{j=1}^{\lambda_1} (N - j),
\]

the agreement with (6.1) is immediately checked.

Based on remark 2.2, we can study the kernel in the real Ginibre ensemble. We use

\[
\hat{K}_N(x; y)|_{\text{R-Gin}} = (x - y) \langle \det(x - M) \det(y - M^T) \rangle_{\text{R-Gin}},
\]

where \( M \) is a random matrix taken from the real Ginibre ensemble. Proceeding as for the complex Ginibre ensemble, but this time using the expansion from theorem 2.1 instead of theorem 2.4, we independently reproduce the result of [2]:

\[
\hat{K}_N(x; y)|_{\text{R-Gin}} = (x - y)(N - 1) \sum_{j=0}^{N-1} \frac{(xy)^j}{j!}.
\]

### 6.2. Dotsenko–Fateev ensemble

The coefficient in the Schur expansion can be evaluated explicitly for the weight function

\[
w_{\text{DF}}(z, \bar{z}) = w_{\text{JUE}}(z) w_{\text{JUE}}(\bar{z}) = |z|^{2\alpha} |1 - z|^{2\beta}.
\]
The corresponding partition function is a Dotsenko–Fateev integral \([23]\), which is a complex version of the Selberg integral,

\[
Z_M^{\text{DF}} = \frac{1}{M!} \int \left[ \Delta_M(z) \Delta_M(\bar{z}) \right]^M \prod_{j=1}^{M} \omega_{\text{DF}}(z_j, \bar{z}_j) dz_j d\bar{z}_j
\]

\[
= \left( Z_M^{\text{JE,} \gamma} \right)^{2M} \prod_{j=1}^{M} \frac{\sin \pi (\alpha + 1 + \gamma (M-j)) \sin \pi (\beta + 1 + \gamma (M-j))}{\sin \pi (\alpha + \beta + \gamma (2M-j-1))} \frac{\sin \frac{\gamma j}{2}}{\sin \frac{\gamma j}{2}}.
\]

\(\Re \gamma \geq 0\). In the second line, we have introduced the Jacobi beta-ensemble with beta-parameter \(\gamma\), in which the Vandermonde factor appears as \(|\Delta_M(z)|^2\). The corresponding partition function is the celebrated Selberg integral \([32, 78]\):

\[
Z_M^{\text{JE,} \gamma} = \frac{1}{M!} \prod_{j=1}^{M} \Gamma (\alpha + 1 + \gamma (M-j)) \Gamma (\beta + 1 + \gamma (M-j)) \Gamma (1 + \gamma j). \]

As a corollary of \([44, \text{ theorem 14}]\) we get

\[
\langle s_\lambda \bar{s}_\mu \rangle_{\text{DF}} = \langle s_\lambda \rangle_{\text{JE,} \gamma} \langle s_\mu \rangle_{\text{JE,} \gamma}.
\]

Thus, the following factorization theorem holds.

**Proposition 6.2.** The kernel on \(\mathbb{C}\) associated to the Dotsenko–Fateev ensemble is completely factorized into holomorphic and anti-holomorphic sector:

\[
\tilde{K}_N^{(n)}(x_1, \ldots, x_n, \bar{y}_1, \ldots, \bar{y}_n) |_{\text{DF}} = K_N^{(n)}(x) \cdot K_N^{(n)}(\bar{y}),
\]

where

\[
K_N^{(n)}(x) = \sum_{\lambda \subseteq S_{n,N-n}} s_\lambda (x^\nu) \langle s_\lambda \rangle_{\text{JE,} \gamma}.
\]

The DF kernels we are considering are tightly related to conformal blocks in the \(SL(2, \mathbb{R})\) WZW conformal field theory \([45, 46]\). In particular, for the two-point kernel, it follows from \([45]\) (see also \([44]\)) that

\[
K_1^{(1)}(z) = \sum_{k=0}^{N-1} h_k(-z^{-1}) \langle e_k \rangle_{\text{JE,} \gamma}
\]

\[
= \sum_{k=0}^{N-1} (-z)^{-k} \binom{N-1}{k} \prod_{j=1}^{k} \frac{\alpha + 1 + \gamma j}{\alpha + \beta + 2 + \gamma (2N-j-3)},
\]

where \(e_k\) are elementary symmetric polynomials. The two-point DF kernel is thus reinterpreted as a generating function of certain correlation functions in the WZW model, and the factorization follows from the ‘chiral’ factorization property studied in \([45, 46]\).

\[\text{In particular, for } \gamma = 1 \text{ in the Dotsenko–Fateev integral we need the JOE.}\]
7. Outlook

While we have studied a considerable number of cases explicitly, including novel analytical evaluations, it is manifest that a more exhaustive study of the expansions obtained can be carried out. It is worth mentioning that the kernels discussed here appear in different contexts, sometimes without any reference to random matrix theory. For example, in the very recent [51], a family of antisymmetric kernels is constructed. Due to the reproducing property, these kernels are essentially of the type (1.1) if the seed kernel is a Christoffel–Darboux kernel. Therefore, the expansions obtained here could conceivably be applied in this other context. It is then a natural question to ask if such an expansion could provide any type of, say computational, advantage.

If the seed kernel is a simpler one, such as a Gaussian kernel, an instance studied in detail in [51], we still have a multi-faceted connection with random matrix theory, as explained in [77, section 4]. The corresponding kernel is then the Karlin–McGregor kernel for non-intersecting diffusion processes. In general, the antisymmetrization procedure parallels the Karlin–McGregor construction of non-intersecting diffusion processes. This highlights yet another universality and interdisciplinarity aspect, inherent to such kernels. It would be interesting if some of the analytical results here have applications along these lines. A seemingly simpler open problem would be to eventually interpret the various Schur averages as transition probabilities of different systems of non-intersecting walkers.

To conclude, it is worth to mention that expanding the multi-point kernel in the Schur basis is certainly not the only option. One may expand in any basis of symmetric function as, for instance, in powers sums in \( t \). Such expansions follow from theorem 2.1 with a change of basis. Let \( \{ u_\mu \}_\mu \) be any basis in the ring of symmetric functions in \( 2n \) variables and let \( C^\lambda_\mu \) denote the entries of the transition matrix to the Schur basis, \( s_\lambda = \sum_\mu C^\lambda_\mu u_\mu \). Then, theorem 2.1 implies that

\[
\tilde{K}^{(n)}(x_1, \ldots, x_n; y_1, \ldots, y_n) = \sum_{\mu \in \mathcal{Y}} u_\mu(t) \left[ \sum_{\nu \in \mathcal{Y}_{2n-n}} \langle s_\nu \rangle u_{\nu'} C^\nu_\mu \right] .
\]

The sum in square brackets is assembled into the average of a symmetric function (labelled by \( \mu \)), distinct from \( u_\nu \) unless \( u_\mu = s_\mu \). In general, these averages are not known in closed form.

Alternatively, one could also expand the determinant \( \det \prod_{i=1}^{2n} (1 + t Z) \) in the matrix average into products of traces of powers of the \( M \times M \) matrix \( Z \). Then, the coefficients in the expansion of the multi-point kernels would be given by (intricate combinations of) correlators of moments of random matrices, that appear in a broad variety of areas.

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\[ ^7 \text{Likewise, if the base space is discretized, i.e. } \mathbb{R}^d \text{ is replaced with a lattice } a\mathbb{Z}^d, \text{ the antisymmetric Gaussian kernel in [51] can be written as the average of two Schur polynomials in a Stieltjes–Wigert or Rogers–Szegő ensemble. The evaluation of such average is known and is given by a certain topological knot invariant in } \mathbb{S}^3 \text{ [37].} \]
A.1. Schur polynomial average in the Laguerre ensemble

In this appendix we give a proof of proposition 3.2, following [35, section 4.2.2]. Along the way, we keep the discussion general to show how the method is suitable to the averages of a Schur polynomial in classical ensembles.

Starting with the definitions

$$\langle s_\mu \rangle = \frac{\int \Delta(z)^2 s_\mu(z) \prod_{j=1}^M w(z_j) dz_j}{\int \Delta(z)^2 \prod_{j=1}^M w(z_j) dz_j},$$

and

$$\Delta(z)^2 s_\mu(z) = \det_{1 \leq i,j \leq M} [x_i^{M-j}] \det_{1 \leq i,j \leq M} [x_i^{j+M-j}],$$

we can apply Andreief’s identity [5, 29] to write

$$\langle s_\mu \rangle = \frac{\det_{1 \leq i,j \leq M} [m_{\mu_j+M-j-1+k}]}{\det_{1 \leq i,j \leq M} [m_{j+k-1}]}.$$

where we have adopted the shorthand notation

$$m_p = \int z^p w(z) dz$$

for the moments of the measure $w(z)dz$. They can be evaluated exactly for the classical ensembles of section 3:

$$m_p^{\text{GUE}} = \int_{-\infty}^{+\infty} \frac{dz}{\sqrt{2\pi}} z^p e^{-z^2/2} = \frac{2^{p/2}}{\sqrt{\pi}} (1 + (-1)^p) \Gamma \left( \frac{p+1}{2} \right),$$

$$m_p^{\text{LUE}} = \int_0^{+\infty} dz z^{p+\alpha} e^{-z} = \Gamma (1 + \alpha + p),$$

$$m_p^{\text{JUE}} = \int_0^1 dz z^{p+\alpha} (1-z)^\beta = \frac{\beta}{(1+\alpha+p)(2+\alpha+p)},$$

$$m_p^{\text{JUE}} = \int_0^{+\infty} dz \frac{z^{p+\alpha}}{(1+z)^{M+\beta}} = \frac{\Gamma(p+\alpha+1)\Gamma(M-p-\alpha+\beta-1)}{\Gamma(M+\beta)},$$

$$m_p^{\text{LUE}} = \int_0^{+\infty} dz e^{-z/\sqrt{\alpha}} = \Gamma(\sqrt{\alpha} - p - 1).$$

The property $\Gamma(z+1) = z\Gamma(z)$ can be used to simplify the determinants. We have the relations

$$m_{\mu_j+M-j+k+1}^{\text{GUE}} = (\mu_j + M - j + k) m_{\mu_j+M-j+k}^{\text{GUE}},$$

$$m_{\mu_j+M-j+k}^{\text{LUE}} = (\alpha + \mu_j + M - j + k) m_{\mu_j+M-j+k-1}^{\text{LUE}},$$

$$m_{j+k}^{\text{GUE}} = (j + k - 1) m_{j+k-2}^{\text{GUE}},$$

$$m_{j+k-1}^{\text{LUE}} = (\alpha + j + k) m_{j+k-2}^{\text{LUE}}.$$
This appendix contains a proof of formula (5.3) for \( J. \) Phys. A: Math. Theor.
the determinant in the numerator, and the factor where we recall that
In the Stieltjes–Wigert case the moments are
proposition 3.2[35].

Remark A.1. Identifying the integral with the insertion of a Schur polynomial with a
determinant of a minor of a Hankel matrix makes manifest that, whenever a weight function
measure \( m \) is even and with even support, the average \( s_\mu \) will be subject to parity constraints on the
partition \( \mu \), precisely as in the GUE. This stems from the vanishing of the odd moments of the
measure \( w(z)dz \), as for instance in the Chebyshev weights of first and second kind.

A.2. Schur polynomial average in q-ensembles: Stieltjes–Wigert

This appendix contains a proof of formula (5.3) for \( s_\mu \) in the Stieltjes–Wigert ensemble which
is along the lines of appendix A.1. See [22] for a different proof.

We henceforth focus on the LUE. We bring out the factor in the (1, j)-entry, \( \forall j = 1, \ldots, M \)
both in the numerator and the denominator. This leaves the numerator as the determinant of the
matrix whose entry \( (j, k) \) is \( \prod_{\nu=1}^{j-1} (\alpha + \mu_j + M - j + \nu) \), with \( \prod_{\nu=1}^{j-1} 1 \) understood. A
similar expression is obtained in the denominator.

Then, we successively subtract lower columns to the \( k \)th column, so to reorganize the
expression as

\[
\langle s_\mu \rangle_{\text{LUE}} = \frac{1}{\det_{1 \leq j, k \leq M} (j + 1)^{k-1}} \cdot \det_{1 \leq j, k \leq M} (\mu_j + M - j + 1)^{k-1} / \Gamma (\alpha + j) \cdot \Gamma (\alpha + j + 1) / \Gamma (\alpha + j + 1) \cdot \Gamma (\alpha + j + 2).
\]

The determinant in the denominator is simply a Vandermonde on the integers, that gives \( G(M+1) \),
while the numerator is a Vandermonde on the lattice \( \mu_j + M - j + 1 \), which gives \( G(M+1) s_\mu(1, \ldots, 1) \). The ratio of determinants leaves behind \( s_\mu(1^M) \). This proves
proposition 3.2[35].

Remark A.1. Identifying the integral with the insertion of a Schur polynomial with a
determinant of a minor of a Hankel matrix makes manifest that, whenever a weight function \( w(z) \)
is even and with even support, the average \( s_\mu \) will be subject to parity constraints on the
partition \( \mu \), precisely as in the GUE. This stems from the vanishing of the odd moments of the
measure \( w(z)dz \), as for instance in the Chebyshev weights of first and second kind.

A.2. Schur polynomial average in q-ensembles: Stieltjes–Wigert

This appendix contains a proof of formula (5.3) for \( s_\mu \) in the Stieltjes–Wigert ensemble which
is along the lines of appendix A.1. See [22] for a different proof.

Reasoning as in appendix A.1, we write

\[
\langle s_\mu \rangle_{\text{SW}} = \frac{\det_{1 \leq j, k \leq M} \left[ m_{\mu_j + M - j + k - 1}^{\text{SW}} \right]}{\det_{1 \leq j, k \leq M} \left[ m_{\mu_j + M - j + k - 2}^{\text{SW}} \right]}.
\]

In the Stieltjes–Wigert case the moments are

\[
m_{\mu_j + M - j + k}^{\text{SW}} = \int_0^{\infty} z^\mu e^{-\frac{1}{2} \left( \ln z \right)^2} \frac{dz}{\sqrt{2\pi}} = q^{\frac{(\mu_j + M - j + 1)^2}{2}},
\]

where we recall that \( q = e^{-g} \). We then extract the factor \( q^{\frac{1}{2} (\mu_j + M - j + 1)^2} \) from the \( j \)th row in
the determinant in the numerator, and the factor \( q^{\frac{1}{2} j^2} \) from the \( j \)th row in the determinant in
the denominator. We are left with
Consider now the Schur polynomial average in $q$-ensembles: $q$-Laguerre

\[ \langle s_\mu \rangle_{SW} = \prod_{j=1}^{M} q^{-\frac{(\mu_j + M - j + 1)^2}{2}} \cdot \frac{\det_{1 \leq j, k \leq M} \left[ q^{-\mu_j - M - j + k + 1} q^{-j} \right]}{\det_{1 \leq j, k \leq M} \left[ q^{-M - j + k} \right]} \]

After bringing out the common factor from the $k$th row in both the determinants, we recognize the ratio of a Vandermonde determinant evaluated at the exponential lattice $q^{-\mu_j + M - j + 1}$ in the numerator, and a Vandermonde determinant on the exponential lattice $q^{-j}$ in the denominator. Simplifying also the prefactor, we get

\[ \langle s_\mu \rangle_{SW} = q^{-\sum_{j=1}^{M} \left( \frac{\mu_j + M - j + 1}{2} \right)} \prod_{1 \leq j, k \leq M} \frac{q^{-\mu_j - M - j + k + 1} - q^{-\mu_j - M - j + k + 1}}{q^{-j} - q^{-k}} \]

By direct computation, the last product can be simplified into

\[ \prod_{1 \leq j, k \leq M} \frac{q^{-\mu_j - M - j + k + 1} - q^{-\mu_j - M - j + k + 1}}{q^{-j} - q^{-k}} = \prod_{j=1}^{M} q^{-\mu_j - M - j + 1} \prod_{1 \leq j, k \leq M} \frac{[\mu_j - j - \mu_k + k]}{[j - k]} q^{1}, \] (A.1)

where in the last line we have used the symmetric $q$-number $[1]_q$, see (5.1). Recognizing the $q$-dimension (5.2), proposition 5.1 follows.

A.3. Schur polynomial average in $q$-ensembles: $q$-Laguerre

Consider now the $q$-Laguerre weight. In this case the moments are [64]

\[ w^\text{qLUE}_p = \int_0^\infty dz \frac{z^{p+\alpha}}{(-1-q)^{-1} z; q)_\infty} = \frac{\Gamma(-p - \alpha) \Gamma(p + \alpha + 1)}{\Gamma(-p - \alpha)} \]

Under a shift $p \mapsto p + 1$ they transform simply as

\[ w^\text{qLUE}_{p+1} = [- \alpha - \alpha + 1]_q w^\text{qLUE}_p, \]

where $[z]_q$ is the asymmetric $q$-number

\[ [z]_q \equiv \frac{1 - q^z}{1 - q} \]

This closely resembles the behaviour of the LUE moments in appendix A.1, with ordinary numbers $(p + \alpha + 1)$ replaced by their $q$-analogue $[-(p + \alpha + 1)]_q$.

We can therefore proceed as in appendix A.1, writing $\langle s_\mu \rangle_{q\text{LUE}}$ as a ratio of determinants and extract the common factor in the $j$th row, $\forall j = 1, \ldots, M$. We arrive at

\[ \langle s_\mu \rangle_{q\text{LUE}} = \prod_{j=1}^{M} \frac{\Gamma(\alpha + 1 + \mu_j + M - j) \Gamma(-\alpha - \mu_j - M + j)}{\Gamma(\alpha + j) \Gamma(-\alpha - j + 1)} \frac{\Gamma(\alpha - j + 1)}{\Gamma(\alpha - \mu_j - M + j)} \]

\[ \times \frac{\det_{1 \leq j, k \leq M} \left[ \prod_{s=1}^{k-1} [-(\alpha + \mu_j + M - j + s)]_q \right]}{\det_{1 \leq j, k \leq M} \left[ \prod_{s=1}^{k-1} [-\alpha + j - 1 + s]_q \right]} \]
(with $\prod_{s=1}^{k} (\ldots) \equiv 1$ understood). Using the simple recursion $[-z]_q = q^{-1} (1 - [-z - 1])_q$ and taking linear combinations of the columns, we can rearrange the determinant in the numerator such that the $(j, k)$-entry is \(^8\)

$$-(-q)^{\frac{jk-1}{2}} \frac{[\alpha + j(M - j + 1)]_q^{-1}}{\Gamma(\alpha + j + M - j + 1)}$$

and similarly in the denominator with the usual replacement $\mu_j + M - j \mapsto j - 1$. Bringing out the common factor $-(-q)^{\frac{jk-1}{2}}$ from the 4th column, $k = 2, \ldots, M$, both in the numerator and denominator, we are left with a ratio of Vandermonde determinants, over the exponential lattices $[-(\alpha + \mu_j + M - j + 1)]_q$ and $[-(\alpha + j)]_q$ respectively.

Putting all together we have

$$\langle \mathcal{S}_q \rangle_{\text{LUE}} = \prod_{j=1}^{M} \frac{\Gamma(\alpha + 1 + \mu_j + M - j) \Gamma(-\alpha - \mu_j - M + j)}{\Gamma(\alpha + j) \Gamma(-\alpha - j + 1)} \frac{\Gamma_q(-\alpha - j + 1)}{\Gamma_q(-\alpha - \mu_j - M + j)}$$

$$\times \prod_{1 \leq j < k \leq M} \frac{[-(\alpha + \mu_j - j + M + 1)]_q}{[\alpha + j]_q} \frac{[-(\alpha + \mu_k - k + M + 1)]_q}{[\alpha + k]_q}.$$  

Writing

$$\prod_{1 \leq j < k \leq M} \frac{[-(\alpha + \mu_j - j + M + 1)]_q}{[\alpha + j]_q} \frac{[-(\alpha + \mu_k - k + M + 1)]_q}{[\alpha + k]_q} = \prod_{1 \leq j < k \leq M} \frac{q^{-\mu_j + M - j + 1}}{q^{-\mu_k + M - k + 1}} \frac{q^{-j} - q^{-k}}{q^{-j} - q^{-k}}$$

and using again (A.1) to identify the $q$-dimension, we arrive at proposition 5.2.

**Appendix B. Duduchava–Roch formula for Toeplitz inverses and kernels**

In this appendix we do not focus on matrix models defined on (subsets of) $\mathbb{R}$, associated to Hankel determinants, and discuss instead a result concerning Toeplitz matrices and their inverses, showing a relationship with random matrix kernels. In this way, we give a different expansion of such kernels, in a monomial basis, instead of in a Schur basis, as in the main text.

Let $\mathcal{T}(f)$ be the Toeplitz matrix with symbol $f : \mathbb{S}^1 \to \mathbb{C}$. Besides, let $\mathcal{M}_z$ be the diagonal matrix

$$(\mathcal{M}_z)_{jk} = \frac{\Gamma(z + k)}{\Gamma(k) \Gamma(z + 1)} \delta_{jk}, \quad z \in \mathbb{C}.$$  

A pure Fisher–Hartwig singularity is the weight function on the circle defined as

$$u_{\text{FH}}(z) = \omega_\gamma(z) \tilde{\omega}_\delta(z), \quad \omega_\gamma(z) = (1 - z/z_0)^\gamma, \quad \tilde{\omega}_\delta(z) = (1 - z_0/z)^\delta, \quad (B.1a)$$

$$\omega_\gamma(z) = (1 - z/z_0)^\gamma, \quad \tilde{\omega}_\delta(z) = (1 - z_0/z)^\delta, \quad (B.1b)$$

\(^8\)The sign in front of each entry requires care. There is a product $\prod_{s=1}^{k-1}$ and each entry of the product includes a factor $(1 - [-(-\ldots)]_q)$ if $s$ is odd and $[-(-\ldots)]_q$ if $s$ is even, $2 \leq s \leq k - 1$, while $s = 1$ always contributes $-[-(-\ldots)]_q$. 

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It was pointed out in [36, equation (2.34)] that given a proof based only on the Duduchava–Roch formula (B.2), the partition function from (B.4) has cancelled against the change in normalization from

\[ T \left( \omega, \beta \right) \mathcal{M}_{\gamma, \delta} T \left( \omega, \beta \right) = \frac{\Gamma(1 + \gamma) \Gamma(1 + \delta)}{\Gamma(1 + \gamma + \delta)} \mathcal{M}_S T \left( \omega, \beta \right) \mathcal{M}_\gamma, \tag{B.2} \]

where \( \omega, \beta \) are defined in (B.1).

The identity (B.2) is known as the Duduchava–Roch formula. See [15] for extensive discussion and proofs.

Let us assume \( \gamma, \delta \in \mathbb{Z}_{>0} \), and also set \( z_0 = 1 \) to lighten the formulae. Our interest is in the inverse of the \( M \times M \) Toeplitz matrix \( T_M \left( \omega, \beta \right) \). It can be computed using (B.2), obtaining [15, 36]

\[ \left[ T^{-1}_M \left( \omega, \beta \right) \right]_{jk} = (-1)^{j+k} \frac{\Gamma(\gamma+j)\Gamma(\delta+k)}{\Gamma(j)\Gamma(k)} \times \sum_{r=\max(j,k)}^{M-1} \frac{\Gamma(r)}{\gamma(\gamma+\delta+r)} \left( \frac{\gamma+r-k-1}{r-k} \right) \left( \frac{\delta+r-j-1}{r-j} \right). \tag{B.3} \]

It was pointed out in [36, equation (2.34)] that

\[ (-1)^{j+k} \left[ T^{-1}_M \left( \omega, \beta \right) \right]_{jk} = \frac{Z_M^{\gamma, \delta}}{Z_{M+1}^{\gamma, \delta}} \langle e_{j-1} \bar{e}_{k-1} \rangle_{\text{FH}} |_{M}, \tag{B.4} \]

with \( Z_M^{\gamma, \delta} \) the partition function of the Fisher–Hartwig ensemble of \( M \) variables. Therefore, (B.3) implies the evaluation of the Selberg–Morris integral

\[ Z_M^{\gamma, \delta} \langle e_{j-1} \bar{e}_{k-1} \rangle_{\text{FH}} |_{M} = \frac{1}{M!} \int_{(\mathbb{C}^N)^M} e_{j-1}(z) e_{k-1}(\bar{z}) \prod_{j=1}^{M} \omega_j(z_j) \omega_{\beta j}(z_j) \frac{dz_j}{2\pi i z_j}, \]

where \( e_{j-1}(z) \equiv e_{j-1}(z_1, \ldots, z_M) \) and likewise for \( e_{k-1}(\bar{z}) \).

At this point, we consider the two-point kernel \( K_N(x, y)_{\text{FH}} \) associated to the Fisher–Hartwig weight function (B.1). With the conventions of section 6 and using theorem 2.4 we get

\[ K_N(x, y)_{\text{FH}} = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} x^{N-j-1} y^{N-k-1} (-1)^{j+k} \langle e_{j} \bar{e}_{k} \rangle_{\text{FH}} \]

\[ = \sum_{j=0}^{N-1} \sum_{k=0}^{N-1} x^{N-j-1} y^{N-k-1} \left[ T^{-1}_N \left( \omega, \beta \right) \right]_{jk}. \tag{B.5} \]

In passing to the second line we have used (B.4) with \( M = N - 1 \). Notice that the ratio of partition functions from (B.4) has cancelled against the change in normalization from \( \tilde{K}_N \) to \( K_N \).

Therefore, (B.5) shows that the kernel associated to the Fisher–Hartwig weight is the generating function of the inverse of the Toeplitz matrix with symbol \( w_{\text{FH}} \). This fact is known since long ago and is a particular case of [19], see also [49, 79] for further discussion. Here we have given a proof based only on the Duduchava–Roch formula (B.2).

with \( \Re(\gamma + \delta) > -1 \) and \( z \in S^1 \), and for a fixed reference point \( z_0 \in S^1 \) that is usually set to 1.

**Theorem B.1 (Duduchava–Roch [15, 24, 72]).** If \( \gamma, \delta, \gamma + \delta \notin \mathbb{Z}_{<0} \), it holds that

\[ T \left( \omega, \beta \right) \mathcal{M}_{\gamma, \delta} T \left( \omega, \beta \right) = \frac{\Gamma(1 + \gamma) \Gamma(1 + \delta)}{\Gamma(1 + \gamma + \delta)} \mathcal{M}_S T \left( \omega, \beta \right) \mathcal{M}_\gamma, \tag{B.2} \]
In conclusion, formulas (B.5)–(B.3) give explicitly the expansion of the Fisher–Hartwig kernel. In turn, \( K_{\varepsilon}(x, y) \) is related via Gessel’s identity [40] to the Meixner kernel (see [75] and references therein). The latter is a limit of the hypergeometric kernel [14], that has many connections with various objects in random matrix theory.

**Appendix C. Alternative derivation of the Chebyshev heat kernel**

In this work we have restricted our attention to Christoffel–Darboux kernels, which have no temporal dependence and hence can be understood as an ‘initial time’ version of heat kernels (with no scaling, this is equivalent to the delta sequence point of view of the Christoffel–Darboux kernels). However, we can employ the little-used map between Schur polynomials of two variables and Chebyshev polynomials of the second kind, to evaluate the heat kernel associated to such polynomials using Schur-related techniques. This emphasizes the correspondence.

Consider then the heat kernel [4, section 3]

\[
K_q(\xi, \eta) = \sum_{j=0}^{\infty} q^j \left( U_j \left( \frac{\xi}{2} \right) U_j \left( \frac{\eta}{2} \right) \right)
\]

\[
= \frac{1 - q^2}{1 - q\xi\eta + q^2 (\xi^2 + \eta^2 - 2) - q^3 \xi\eta + q^4},
\]

for \(-2 \leq \xi, \eta \leq 2\) and \(0 < q < 1\). The ‘time’ is \(-2 \log q\). The heat kernel (C.1) is genuinely different from the kernels we have considered in the main text, whence the change in notation.

As in (2.6b), the relation

\[
U_j \left( \frac{\xi}{2} \right) = (x_1 x_2)^{-k-j/2} s_{k+j, k}(x_1, x_2), \quad \xi \equiv \frac{x_1 + x_2}{\sqrt{x_1 x_2}}
\]

between Schur and Chebyshev polynomials of second kind can be plugged in (C.1a) to show that

\[
K_q(\xi, \eta) = \sum_{j=0}^{\infty} q^j s_{k+j, k}(x, x^{-1}) s_{k+j, k}(y, y^{-1})
\]

\[
= \sum_{j=0}^{\infty} q^j h_j(x, x^{-1}) h_j(y, y^{-1}).
\]

We have used the freedom in doubling the variables to set \(x_1 = x = x_2^{-1}\) and \(y_1 = y = y_2^{-1}\). At this point, we use the equality

\[
\sum_{j=0}^{\infty} q^j h_j(x, x^{-1}) h_j(y, y^{-1}) = \oint \frac{dz}{2\pi i z (1 - qz/x) (1 - qz/y) (1 - qz/x) (1 - qz/y)},
\]

which is the simplest case of Gessel’s identity [40]. Computing the integral by residues reproduces (C.1b).

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