One-matrix differential reformulation of two-matrix models

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Differential reformulations of field theories are often used for explicit computations. We derive a one-matrix differential formulation of two-matrix models, with the help of which it is possible to diagonalize the one- and two-matrix models using a formula by Itzykson and Zuber that allows diagonalizing differential operators with respect to matrix elements of Hermitian matrices. We detail the equivalence between the expressions obtained by diagonalizing the partition function in differential or integral formulation, which is not manifest at first glance. For one-matrix models, this requires transforming certain derivatives to variables. In the case of two-matrix models, the same computation leads to a new determinant formulation of the partition function, and we discuss potential applications to new orthogonal polynomials methods.

I. Introduction

Two-matrix models have been widely studied in the literature since their introduction by Itzykson and Zuber [1], see for instance [2–7]. Their Feynman expansions count ribbon graphs with two kinds of vertices. They match, for different choices of potentials, the partition functions of some statistical models such as the Ising model [8], colored triangulations [9], hard particles [10] etc. on random surfaces (thereby relating to 2D quantum gravity).

In quantum field theory, differential formulations such as

\[
\frac{\mathcal{Z}}{\mathcal{Z}_0} = \frac{1}{\mathcal{Z}_0} \int \mathcal{D}\phi e^{-\int d^d x d^d y \frac{1}{2} \phi(x)K(x,y)\phi(y) - \int d^d x V(\phi(x))}
= \left[ e^{-\int d^d w V\left(\frac{\partial}{\partial J(w)}\right)} e^{\int d^d x d^d y J(x)K^{-1}(x,y)J(y)} \right]_{J=0},
\]

are widely used (Zee coins it the “central identity of quantum field theory” in his famous book [11]).

Here the \(\phi\) are fields on a \(D\)-dimensional spacetime, \(\int \mathcal{D}\phi\) is a functional integration, the interaction potential \(V\) is a polynomial in the fields, \(K\) is the propagator, and \(\mathcal{Z}_0\) is a normalization. Such formulations allow expressing Gaussian expectation values and Wick’s formula, the key ingredient of perturbative expansions, as well-defined algebraic expressions that do not suffer the potential divergences of the combinatorially equivalent integrals (see for instance [12–15]). They moreover provide a practical tool for computations of Feynman diagrams.

In particular, the differential formulation provides well-defined expressions for the Gaussian expectation values of two-matrix models, while the integrals are not well-defined and only understood formally through Wick’s theorem.

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Remarkably, as we will show in the paper, two-matrix models can be expressed in terms of a single matrix in a differential formulation: the two matrices necessary to control the interaction between the two kinds of vertices (or equivalently impose bipartiteness of the ribbon graphs) are no longer needed in the differential formulations. Since the necessity of using two matrices complicates the resolution of two-matrix models, the existence of a one-matrix differential formulation raises the question whether this allows for a simpler resolution.

With the help of the one-matrix differential reformulation, it is possible to diagonalize the two-matrix models directly in the differential formulation. The steps leading to the expression of the partition function as a Slater determinant and thereby to the resolution by biorthogonal polynomials involve elegant and interesting equations whose equivalence is not always manifest. We detail their relations and take the opportunity to review and clarify some aspects of the differential formulation that can be found in the matrix-model literature, as well as the formula needed to diagonalize the differential operators with respect to matrix elements for Hermitian matrices, implicitly shown by Itzykson and Zuber in [1] and stated by Zuber in [16]. These computations do not seem to indicate any simplification in the resolution of two-matrix models in differential formulation as initially hoped: the role played by the two matrices is now played by the matrix and the derivatives with respect to the matrix, leading to the same resolution by biorthogonal polynomials.

On the other hand, the analysis of the equivalence between the different formulas for one-matrix models does provide new insights on two-matrix models. Indeed, the proof of equivalence between two of the diagonalized differential formulations of the one-matrix models involves transforming certain derivatives to variables in the Slater determinant formulation. The computation is not trivial and is detailed in the last section. The same computation can be applied with little modifications to two-matrix models, and leads to a new determinant formulation of their partition functions. Building on this, we comment on a potential new resolution using orthogonal polynomials instead of biorthogonal polynomials.

In order to show that the different expressions of the partition functions indeed coincide, we verify that all the prefactors and normalizations match. These constants are often omitted in the literature as they are sometimes tedious to compute and do not contribute to the critical behaviors or the combinatorial interpretations. An advantage of the computations in differential formulations is that they most often do not need an overall normalization, as seen for instance in [11].

Differential formulations are often used in the literature with little or without justifications or references, and in addition to the developments presented here, the present paper is also intended to provide a text of reference by giving an exhaustive review in the context of random matrix models.

For the benefit of the reader, we first present a diagrammatic overview of all the different representations of the one- and two-matrix models that we consider in this article, with lines connecting two representations when we have a direct computation for translating between them. The expressions are shown in a schematic way, omitting normalization terms, traces, factors of \( N \), and the like. The full form of each expression can be found in the main text of the article. Function arguments and integration variables are omitted where possible. If confusion may arise, the arguments of the potential functions \( V \) (for one-matrix models) and \( V_1, V_2 \) (for two-matrix models) are indicated by superscripts. The action is understood to be \( S = N \text{Tr}(\frac{1}{2}M^2 - V(M)) \) for
one-matrix models, and \( S = N \text{Tr}(AB - V_1(A) - V_2(B)) \) for two-matrix models. Elements of the matrices appearing inside the determinants are labeled \( i, j \). In the external expressions (77) and (78), the \( r, s, p \) are related to the \( i \) of (56) and (19) as \( i = (p - 1)r + s \) for a polynomial potential \( V_1(x) = \sum_k \frac{\alpha_k}{k!}x^k \) of degree \( p \).

The structure is the same in both diagrams. In the bottom layer, we have all the integral representations of the matrix model under consideration. In the top layer, we show the corresponding expressions in differential form. \( \Delta \) denotes the Vandermonde determinant. The front layers gather the expressions obtained after diagonalizing in integral form, while the back layers gather the expressions obtained after diagonalizing in differential form.

Several lines connecting parts within a diagram share common features, indicated by a common greek letter. The lines \( \alpha \) symbolize the direct equivalence between integral and differential formulations. Lines \( \beta \) symbolize the equivalence between the one- and two-matrix differential formulations. Lines \( \delta \) involve a diagonalization procedure in integral formulation, leading to the appearance of Vandermonde determinants. Similarly, lines \( \eta \) refer to a diagonalization in the differential formulation. When moving from left to right along lines marked \( \phi \), the common strategy is to expand the Vandermonde determinants, thereby moving the integration or differentiation inside the matrix elements of the determinant. The appearance of \( \kappa \) indicates that a heat kernel method was used to transform between an integral and differential expression. The steps marked \( \sigma \) absorb the \( 1/\Delta \) prefactor by extracting the symmetric part of the determinant. Finally, for vertical lines labeled \( \lambda \), it is shown how to transform some of the derivatives to variables and vice-versa, while \( \tilde{\lambda} \) corresponds to the equivalent computation in integral formulation. These are the computations leading to our newly found determinantal expressions for two-matrix models, that could potentially lead to a new resolution in terms of orthogonal polynomials.

![Figure 1. Diagrammatic overview of the various expressions for one-matrix models.](image-url)
Figure 2. Diagrammatic overview of the various expressions for two-matrix models.

II. Brief presentation of the models

We start with a reminder on the two models presenting all relevant details, in particular their perturbative expansions and diagonalizations as well as the reformulation of two-matrix models in terms of biorthogonal polynomials.

II. 1. One-matrix models

(a) The model. The partition function of the Hermitian one-matrix model with potential $V$ is

$$Z_V = \int \frac{dM}{a_N} e^{-N\text{Tr}(\frac{1}{2}M^2 - V(M))},$$

(2)

where $\mathbb{H}_N$ is the set of Hermitian matrices of size $N \times N$, the potential is $V(x) = \sum_{k \geq 1} \lambda_k x^k$, the measure is $dM = \prod_{i=1}^N dM_{ii} \prod_{i<j} d\Re M_{ij} d\Im M_{ij}$, and the normalization $a_N = 2^{-N^2/2} (\pi/N)^{-N^2/2}$ is such that $Z_{V=0} = 1$ (by explicit computation of the Gaussian integral). The free energy of the model is $F_V = \log Z_V$. 
(b) **Diagrammatic expansion.** By expanding the exponential of the potential $V$, exchanging the summation and the integration, and using Wick’s theorem, one obtains a Feynman expansion of the one-matrix model over ribbon graphs, which can be seen as graphs with an additional cyclic ordering of the edges around each vertex. Given a ribbon graph, a cyclic sequence of edges such that every edge in the sequence also precedes the edge that follows in the ordering around a vertex of the graph is called a *face* (for more details see [17–21]). In the expansion of the partition function, the vertices of valency $k$ (that is, with $k$ edges attached) are counted with a weight $N^{\lambda_k}$ and the edges with a weight $1/N$, as they correspond to propagators

$$\int_{\mathbb{H}_N} \frac{dM}{a_N} e^{-\frac{N}{2} \text{Tr}(M^2)} M_{ij} M_{kl} = \frac{1}{N} \delta_{i,l} \delta_{j,k}. \quad (3)$$

To each face corresponds a trace of these Kronecker symbols, resulting in an additional factor $N$ per face. The Feynman expansion of the free energy involves only connected graphs, which is not true for the partition function. Denoting by $V_k(\Gamma)$ the number of vertices of valency $k$ of a ribbon graph $\Gamma$, the weight (or amplitude) of a graph in the perturbative expansion is therefore

$$A(\Gamma) = \frac{N^{\chi(\Gamma)}}{|\text{Aut}(\Gamma)|} \prod_{k \geq 1} \lambda_k^{V_k(\Gamma)}, \quad (4)$$

where $\chi(\Gamma)$ is the Euler characteristic

$$\chi = V - E + F = 2K - 2g, \quad (5)$$

in which $V = \sum_k V_k$ is the total number of vertices, $E$ is the total number of edges, $F$ the total number of faces, and $K$ the number of connected components of the ribbon graph ($K = 1$ for the graph expansion of the free energy), and $g$, the genus of the ribbon graph, is a non-negative integer. The combinatorial factor $|\text{Aut}(\Gamma)|$ is the number of automorphisms of the ribbon graph $\Gamma$ (see [17–19]).

(c) **Clarification regarding matrix integrals.** All matrix integrals in the paper are understood as formal integrals: They are perturbative expansions labelled by Feynman diagrams, that is, formal generating functions for ribbon graphs (see [19, 21, 22] and references therein). Every equality between partition functions is understood as equality between their perturbative expansions, term by term, regardless of the analytic properties of these series. The questions of whether the matrix integrals are well-defined, whether the series are summable when the coupling constants are in some domain and in some certain limits (e.g. large $N$), and whether an equality between these different quantities holds non-perturbatively are not addressed in this paper.

(d) **Diagonalization.** One of the standard ways of solving this model is to first diagonalize the Hermitian matrices as $M = U^* X U$, where $U \in U(N)$ and $X = \text{diag}(x) = (x_1, \ldots, x_N) \in \mathbb{R}^N$. Due to the unitary invariance, the integration over unitary matrices factors out the volume of the unitary group, and the partition function is expressed as an integral over the $N$ real eigenvalues,

$$Z_V = \int_{\mathbb{R}^N} \frac{dx}{b_N} \Delta^2(x) e^{-N(\frac{1}{2}x^2 - V(x))}, \quad (6)$$

where the measure is $dx = \prod_{i=1}^N dx_i$, the normalization is $b_N = (2\pi)^{N/2} \prod_{j=1}^N j! N^{-N^2/2}$ (this can be computed from Mehta’s integral Eq. 3.3.10 in [23]), $\Delta(x) = \prod_{i<j}(x_j-x_i)$ is the Vandermonde
determinant, \( x^2 = \sum_{i=1}^{N} x_i^2 \), and we use the notation
\[
V(x) \equiv \sum_{i=1}^{N} V(x_i).
\] (7)
To solve the model one may apply for example the saddle point method, loop or Schwinger-Dyson equations, or the method of orthogonal polynomials [19]. Here we focus on the steps leading to the resolution by orthogonal polynomials relying on the diagonalization.

II. 2. Two-matrix models

(a) Partition function. The models we are interested in have the following form:
\[
Z_{V_1,V_2} = \int_{\mathbb{H}_N \times \mathbb{H}_N} \frac{dA dB}{c_N} e^{-N \text{Tr}(AB - V_1(A) - V_2(B))}
\] (8)
where \( V_1(x) = \sum_{k \geq 1} \frac{\alpha_k}{k} x^k \) and \( V_2(x) = \sum_{k \geq 1} \frac{\beta_k}{k} x^k \), and the normalization \( c_N \) is chosen such that for \( V_1 = V_2 = 0 \), \( Z_{0,0} = 1 \). The Gaussian part is formal and should be understood as follows. For two real variables, we formally have for \( \alpha \in \mathbb{R} \):
\[
\int_{\mathbb{R}^2} dx dy e^{-\alpha N xy} = \frac{2\pi}{\alpha N},
\] (9)
and more generally for \( n, m \geq 0 \):
\[
\int_{\mathbb{R}^2} dx dy e^{-\alpha N xy} x^n y^m = \frac{\delta_{n,m}}{(-N)^n} \frac{\partial^n}{\partial \alpha^n} \int_{\mathbb{R}^2} dx dy e^{-\alpha N xy} = \frac{\delta_{n,m}}{(-N)^n} \frac{\partial^n}{\partial \alpha^n} \frac{2\pi}{\alpha N},
\] (10)
so that
\[
\int_{\mathbb{R}^2} dx dy e^{-\alpha N xy} x^n y^m = \delta_{n,m} \frac{2\pi}{\alpha N} \left( \frac{1}{\alpha N} \right)^{n+1} n!.
\] (11)
This leads to the normalization:
\[
c_N = \int_{\mathbb{H}_N \times \mathbb{H}_N} dA dB e^{-N \text{Tr}(AB)} = 2^N \left( \frac{\pi}{4N} \right)^N, \quad (12)
\]
and to the propagator
\[
\int_{\mathbb{H}_N \times \mathbb{H}_N} \frac{dA dB}{c_N} e^{-N \text{Tr}(AB)} A_{ab} B_{cd} = \frac{1}{N} \delta_{a,d} \delta_{b,c}.
\] (13)
The integrals in Eq. (9), Eq. (11), Eq. (12) and Eq. (13) are not well-defined. On the other hand, the values we set formally here are the ones needed in order to recover the correct combinatorial expansions with the correct overall normalizations of the partition functions. In the differential formulation, the equivalent expressions are well-defined and automatically normalized, as will be detailed in the rest of the text.

1 The formal propagator Eq. (11) can be found in [19], Eq. (4.2), but we have added the factor \( 2\pi/\alpha \) to agree with the value of Eq. (8) when \( V_1(x) = V_2(x) = -\varepsilon x^2/2 \), in the limit where \( \varepsilon \to 0 \) see e.g. [24], App. A.
(b) **Diagrammatic expansion.** The Feynman graph expansion of a two-matrix model involves a summation over ribbon graphs that have two kinds of vertices, respectively associated to the traces of the matrices $A$ and $B$, so that the edges only link vertices of different kinds (the ribbon graphs are bipartite). Denoting by $V^A_k(\Gamma)$ and $V^B_k(\Gamma)$ the number of vertices of valency $k$ of each type of vertices of a ribbon graph $\Gamma$, the weight of a Feynman graph is given by

$$A(\Gamma) = \frac{N^\chi(\Gamma)}{|\text{Aut}(\Gamma)|} \prod_{k \geq 1} \frac{\alpha^V_{A_k}(\Gamma) \beta^V_{B_k}(\Gamma)}{k!},$$

the graphs contributing to the free-energy being connected.

(c) **Diagonalization.** The first step in the resolution is to diagonalize $A$ and then use the Harish-Chandra–Itzykson–Zuber formula \[1, 25\] to diagonalize $B$:

$$\int_{U(N)} \frac{dU}{N!} e^{\alpha \text{Tr}(AUBU^\dagger)} = \frac{1}{\sqrt{2\pi i N}} \prod_{k=1}^{N-1} \frac{k! \det \{ e^{\alpha \alpha \beta \beta} \}_{1 \leq i,j \leq N}}{\Delta(a)\Delta(b)},$$

where $dU$ is the normalized Haar measure on the unitary group $U(N)$ and $\alpha$ is a complex coefficient. As shown in \[2\], for $\alpha = N$ this leads to

$$Z_{V_1, V_2} = \int_{\mathbb{R}^{2N}} \frac{da \, db}{\sqrt{N!}} \Delta(a)\Delta(b) e^{-N(a \cdot b - V_1(a) - V_2(b))},$$

where $d_N = (2\pi i)^N \prod_{j=1}^{N-1} j! N^{-N(N+1)/2}$.

(d) **Determinant form.** Let us detail the steps leading to the usual resolution using biorthogonal polynomials \[2\]. The Leibniz determinant formula

$$\det \{ f_{i,j} \}_{1 \leq i,j \leq N} = \sum_{\sigma \in S_N} (-1)^\sigma \prod_{i=1}^N f_{i,\sigma(i)}$$

where $(-1)^\sigma$ is the sign of the permutation $\sigma$, is then used to develop the Vandermonde determinants $\Delta(x) = \det \{ x_i^{j-1} \}_{1 \leq i,j \leq N}$, so that Eq. (16) reads:

$$Z_{V_1, V_2} = \frac{1}{d_N} \sum_{\sigma, \sigma^\prime \in S_N} (-1)^{\sigma+\sigma^\prime} \prod_{i=1}^N \int dx \int dy x^{\sigma(i)-1} y^{\sigma^\prime(i)-1} e^{-N(x-y-V_1(x) - V_2(y))}$$

$$= \frac{1}{d_N} \sum_{\sigma \in S_N} (-1)^\sigma \prod_{i=1}^N \int dx \int dy x^{\sigma(i)-1} y^{i-1} e^{-N(x-y-V_1(x) - V_2(y))},$$

leading to the following expression of the partition function as a Hankel determinant (again by Eq. (17):

$$Z_{V_1, V_2} = \frac{1}{d_N} \det \left\{ \int dx \int dy x^{0} y^{0} e^{NV_1(x)} y^{0} e^{NV_2(y)} \right\}_{0 \leq i,j \leq N-1}.$$
(e) Biorthogonal polynomials. Using the properties of the determinant, one can subtract from each row or column a linear combination of rows or columns of lower indices without affecting the result, so that one can replace $x^i$ and $y^j$ by any polynomials $P_i(x)$, $Q_j(y)$ of degrees $i$ and $j$ respectively with unit leading coefficients (such polynomials are said to be monic):

$$Z_{V_1,V_2} = \frac{1}{d_N} \det \left( \{P_i Q_j\}_{0 \leq i, j \leq N-1} \right), \quad \langle P_i | Q_j \rangle = \int dx \, dy \, e^{-Nxy} P_i(x) e^{NV_1(x)} Q_j(y) e^{NV_2(y)}.$$  \hspace{1cm} (20)

In particular, the monic polynomials $P_i, Q_j$ can be chosen to be orthogonal for the formal symmetric bilinear form on the right-hand side of Eq. (20), in which case the determinant Eq. (19) is the product of the diagonal terms:

$$\langle P_i | Q_j \rangle = h_i \delta_{i,j}, \quad Z_{V_1,V_2} = \frac{1}{d_N} \prod_{i=1}^{N} h_i.$$  \hspace{1cm} (21)

Two sequences of polynomials satisfying this orthogonality relation are said to be biorthogonal. They are determined recursively for specific choices of the potentials $V_1$ and $V_2$ [2, 3, 6, 9, 10].

III. Differential reformulation of matrix integrals

In this section we show how to reformulate two-matrix models in terms of a single matrix and its derivatives. We first review the usual differential formulations of one- and two-matrix models. Then we present the new one-matrix differential formulation of the latter. We finally show how to diagonalize the models in this formulation.

III. 1. The differential formulation

It is known (see e.g. [14]) that a Hermitian one-matrix model admits the following reformulation:

$$Z_V = \int_{\mathbb{H}_N} \frac{dM}{a_N} e^{-N\text{Tr}(\frac{1}{2}M^2 - V(M))} = \left[ e^{\frac{1}{2N} \text{Tr} (\frac{\partial}{\partial M})^2} e^{N\text{Tr} V(M)} \right]_{M=0},$$  \hspace{1cm} (22)

where $\text{Tr} (\frac{\partial}{\partial M})^2 = \sum_{a,b=1}^{N} \frac{\partial}{\partial M_{ab}} \frac{\partial}{\partial M_{ba}}$. The role of the matrices and derivatives can be exchanged:

$$Z_V = \left[ e^{N\text{Tr} V(\frac{\partial}{\partial M})} e^{\frac{1}{2N} \text{Tr} (M^2)} \right]_{M=0},$$  \hspace{1cm} (23)

which is the formulation analogous to Eq. (11) for a one-matrix model. Note that the differential expressions are clearly well-normalized, as for $V = 0$, the only contributing term in the series-expansion of the exponential is 1.

As mentioned in Sec. II.1.(c), the sign “=” means that the equality holds at the perturbative level: the Feynman expansions coincide term by term. While the perturbative expansion of the matrix model is obtained by expanding the exponential of $V$ in series and exchanging the summation and the integration (Sec. II.1.(b)), in the differential formulation, the perturbative expansion is obtained by exchanging the summation and the evaluation $M = 0$.

\[^3\] In the case where both sides of Eq. (22) are well-defined non-perturbatively, one may wonder whether the two coincide.
More precisely, recalling that $V(x) = \sum_{k \geq 1} \frac{\lambda_k}{k} x^k$, the perturbative expansion of the right-hand side of Eq. (22) reads:

$$\left[ e^{\frac{1}{2N} \text{Tr} \left( \frac{\partial^2}{\partial M^2} \right) e^{iNtV(M)} } \right]_{M=0} = \sum_{\{n_k \geq 0\}} \prod_k \frac{(N\lambda_k/k)^{n_k}}{n_k!} \left[ e^{\frac{1}{2N} \text{Tr} \left( \frac{\partial^2}{\partial M^2} \right)^2} \prod_k \text{Tr}(M^k)^{n_k} \right]_{M=0}$$

$$= \sum_{\{n_k \geq 0\}} \prod_k \frac{(N\lambda_k/k)^{n_k}}{n_k!} \sum_{i=0}^1 \frac{1}{i!} \left[ \left( \frac{1}{2N} \text{Tr} \left( \frac{\partial}{\partial M} \right)^2 \right)^i \prod_k \text{Tr}(M^k)^{n_k} \right]_{M=0}$$

$$= \sum_{\{n_k \}} \prod_k \frac{(N\lambda_k/k)^{n_k}}{n_k!} \left[ \left( \frac{1}{2N} \text{Tr} \left( \frac{\partial}{\partial M} \right)^2 \right)^{\ell(n)/2} \prod_k \text{Tr}(M^k)^{n_k} \right],$$

(24)

where $\ell(n) = \sum_k k n_k$. Developing the term originating from the propagator:

$$\left( \text{Tr} \frac{\partial}{\partial M} \frac{\partial}{\partial M} \right)^{\ell(n)/2} = \sum_{p=1}^\ell \sum_{i_p,j_p=1}^N \prod_{p=1}^\ell \frac{\partial}{\partial M_{i_p,j_p}} \frac{\partial}{\partial M_{j_p,i_p}},$$

(25)

we see that the expression between brackets in the last line of Eq. (24) is a sum over all possible ways to pair the derivatives $\frac{\partial}{\partial M_{ij}}$ and the matrix elements $M_{kl}$ originating from the interaction potential, with:

$$\frac{1}{2N} \text{Tr} \frac{\partial}{\partial M_{i_p,j_p}} \frac{\partial}{\partial M_{j_p,i_p}} M_{ab} M_{cd} = \frac{1}{2N} (\delta_{i_p,a} \delta_{j_p,b} \delta_{j_p,c} \delta_{i_p,d} + \delta_{i_p,c} \delta_{j_p,d} \delta_{j_p,a} \delta_{i_p,b}).$$

(26)

Since each $i_p$ and $j_p$ appear only in one term of the sort, the sums over $i_p,j_p$ can be carried out with the following simplifications:

$$\sum_{i_p,j_p} \frac{1}{2N} \text{Tr} \frac{\partial}{\partial M_{i_p,j_p}} \frac{\partial}{\partial M_{j_p,i_p}} M_{ab} M_{cd} = \frac{1}{N} \delta_{a,d} \delta_{b,c},$$

(27)

where we recognize the propagator Eq. (3). The combinatorial factor $\frac{1}{\ell(n)/2}$ compensates for the number of permutations of the traces of pairs of derivatives (the index $p$), as all permutations give the same result. In other words, the term in square brackets in the last line of Eq. (24) coincides with a Gaussian expectation through Wick’s theorem:

$$\frac{1}{(\ell(n)/2)!} \left( \frac{1}{2N} \text{Tr} \left( \frac{\partial}{\partial M} \right)^2 \right)^{\ell(n)/2} \prod_k \text{Tr}(M^k)^{n_k} = \int_{\mathbb{H}_N} \frac{dM}{a_N} e^{-\frac{N}{2} \text{Tr}(M^2)} \prod_k \text{Tr}(M^k)^{n_k},$$

(28)

which is expressed as a sum over ribbon graphs with $n_k$ vertices of valency $k$ and $\ell(n)/2 = \frac{1}{2} \sum_k k n_k$ edges. We thus recover the usual perturbative expansion of Eq. (2), which proves Eq. (22) at the perturbative level. Eq. (23) is shown the same way, with:

$$\frac{1}{(\ell(n)/2)!} \prod_k \left( \text{Tr} \left( \frac{\partial}{\partial M} \right)^k \right)^{n_k} \left( \frac{1}{2N} \text{Tr} (M^2) \right)^{\ell(n)/2} = \int_{\mathbb{H}_N} \frac{dM}{a_N} e^{-\frac{N}{2} \text{Tr}(M^2)} \prod_k \text{Tr}(M^k)^{n_k}.$$
III. 2. Differential formulation of two-matrix models

For a two-matrix model the standard differential formulation is simply

\[ Z_{V_1,V_2} = \int_{\mathbb{H}_N^2} \frac{dA dB}{cN} e^{-N\text{Tr}(AB-V_1(A)-V_2(B))} = \left[ e^{\frac{N}{2} \text{Tr} \left( \frac{\partial}{\partial A} \frac{\partial}{\partial B} \right) e^{N\text{Tr}(V_1(A)+V_2(B))}} \right]_{A,B=0}. \]  (30)

Again, the differential expression is clearly well-normalized. Eq. (30) is shown perturbatively using the fact that now the propagator corresponds to:

\[ \sum_{i_p,j_p} \frac{1}{N} \text{Tr} \frac{\partial}{\partial A_{i_p,j_p}} \frac{\partial}{\partial B_{j_p,i_p}} A_{ab} B_{cd} = \frac{1}{N} \delta_{a,d} \delta_{b,c}, \]  (31)

so that through Wick’s theorem, denoting by \( \ell(a+b) = \sum_k k(a_k + b_k) \), we have

\[ \frac{1}{((a+b)!)} \left( \frac{1}{N} \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} \right)^{\ell(a+b)} \prod_k \text{Tr}(A^k)^{a_k} \text{Tr}(B^k)^{b_k} = \int \frac{dA dB}{cN} e^{-N\text{Tr}(AB)} \prod_k \text{Tr}(A^k)^{a_k} \text{Tr}(B^k)^{b_k}, \]  (32)

the other steps being the same as for one-matrix models. This last quantity is expressed as a sum over bipartite ribbon graphs with \( a_k \) and \( b_k \) vertices of valency \( k \) associated to the matrix \( A \) respectively \( B \) and \( \sum_k ka_k = \sum_k kb_k \) edges.

While the right-hand side of Eq. (32) is formal and understood via Wick’s theorem and the formal propagator Eq. (13), the left-hand side is well-defined.

In the differential formulation however, it is no longer necessary to use two matrices: this is important in the integral formulation to impose bipartiteness of the Feynman ribbon graphs. This bipartiteness can also be implemented in a differential formulation using the fact that derivatives only act on matrices. The resulting differential formulation thereby only involves a single matrix:

\[ Z_{V_1,V_2} = \left[ e^{\frac{N}{2} \text{Tr} V_1(\frac{\partial}{\partial A} \frac{\partial}{\partial B})} e^{\frac{N}{2} \text{Tr} V_2(\frac{\partial}{\partial A} \frac{\partial}{\partial B})} \right]_{AB=0}. \]  (33)

This can be proven by showing that this differential formula generates the same ribbon graphs, together with the same combinatorial weights Eq. (13). It can also be proven in the differential formulation starting from the right-hand side of Eq. (30):

\[ Z_{V_1,V_2} = \left[ e^{\frac{N}{2} \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} e^{N\text{Tr} V_1(A)} e^{N\text{Tr} V_2(B)}} \right]_{A,B=0} = \left[ \left[ e^{\frac{N}{2} \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} e^{N\text{Tr} V_1(A)}} \right]_{A=0} e^{N\text{Tr} V_2(B)} \right]_{B=0}, \]  (34)

where, recalling that \( \ell(a) = \sum_k ka_k \):

\[ \left[ e^{N\text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} e^{N\text{Tr} V_1(A)}} \right]_{A=0} = \sum_{\{a_k \geq 0\}} \sum_{i \geq 0} \prod_k \left( \frac{(N\alpha_k/k)^{a_k} 1}{a_k!} \right)^i \left( \frac{1}{N} \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} \right)^{\ell(a)} \prod_k (\text{Tr} A^k)^{a_k}. \]  (35)

Assuming that we have proven that:

\[ \left( \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} \right)^{\ell(a)} \prod_k (\text{Tr} A^k)^{a_k} = \ell(a)! \prod_k \left( \text{Tr} \left( \frac{\partial}{\partial B} \right)^k \right)^{\alpha_k}, \]  (36)
we obtain
\[ e^{N \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} e^{N \text{Tr} V_1(A)}} \bigg|_{A=0} = e^{N V_1 \left( \frac{\partial}{\partial A} \right)}, \tag{37} \]
and thereby
\[ Z_{V_1, V_2} = e^{N \text{Tr} V_1 \left( \frac{\partial}{\partial A} \right) e^{N \text{Tr} V_2(B)}} \bigg|_{B=0}. \tag{38} \]
The sought formula Eq. (33) is obtained by change of variable \( M = \sqrt{N} B \).²

Let us now prove Eq. (36). We write
\[ \left( \text{Tr} \frac{\partial}{\partial A} \frac{\partial}{\partial B} \right)^{\ell(a)} \prod_k \text{Tr}(A^k)^{a_k} = \sum_{\text{possible assignments}} \prod_k \left( \text{Tr} \left( \left( \frac{\partial}{\partial B} \right)^k \right)^{a_k} \right) = \ell(a)! \prod_k \left( \text{Tr} \left( \frac{\partial}{\partial B} \right)^k \right)^{a_k}, \tag{39} \]
where the last equality holds because there are \((\sum_k k a_k)\)! ways of assigning the derivatives \( \frac{\partial}{\partial B_{ij}} \) to the matrix elements \( A_{kl} \) in the product of traces, and the result for each assignment is independent of the specific assignment chosen. This concludes the proof.

### III. 3. Diagonalization in the differential formulation

It is known [1, 16] that for Hermitian matrices \( A \), the differential operator \( D_k \left( \frac{\partial}{\partial A} \right) = \text{Tr} \left( \frac{\partial}{\partial A} \right)^k \) acting on \( U(N) \)-invariant functions₄ diagonalizes as
\[ D_k \left( \frac{\partial}{\partial a_i} \right) = \frac{1}{\Delta(a)} \sum_{i=1}^{N} \frac{\partial}{\partial a_i^k} \Delta(a). \tag{40} \]
This implies in particular that the differential operator in Eq. (33) has the diagonal form
\[ \text{Tr} V_1 \left( \frac{1}{\sqrt{N}} \frac{\partial}{\partial M} \right) = \sum_{k \geq 1} \alpha_k k \sqrt{N} \sum_{i=1}^{N} \frac{\partial}{\partial m_i} D_k \left( \frac{\partial}{\partial M} \right) = \frac{1}{\Delta(m)} \sum_{i=1}^{N} V_1 \left( \frac{1}{\sqrt{N}} \frac{\partial}{\partial m_i} \right) \Delta(m), \tag{41} \]

---

² Note that in this differential formulation, changes of variables are made without “Jacobian”, since both sides are clearly normalized.

⁵ One may of course wonder if the developments of the paper still hold for matrix models involving real symmetric matrices. The diagonalization would require a formula analogous to Eq. (10), but for the derivatives of a real symmetric matrix-valued \( O(N) \)-invariant function. Such a formula is not known, because the equivalent of the HCIZ formula Eq. (15) with the integration over \( U(N) \) matrices replaced with an integration over \( O(N) \) matrices is known for skew-symmetric matrices \( A, B \), whereas the case of symmetric matrices is much more involved [16].
since it acts on $\exp(V_2(M))$, which depends on traces of $M$ and is thus $U(N)$-invariant.

Since this is only briefly mentioned in the final remarks of [16], let us make the argument more precise here. One defines the operator $\hat{D}_k$ for some positive integer $k$ as the operator which acts on the exponential trace of the product of two Hermitian matrices $A$ and $B$ as

$$\hat{D}_k e^{\alpha \text{Tr} AB} \equiv \alpha^k \text{Tr}(B^k) e^{\alpha \text{Tr} AB} \quad (42)$$

and $\alpha$ is a complex coefficient. From this definition it follows that this operator has the explicit representation as a functional of matrix derivatives

$$D_k \left( \frac{\partial}{\partial A} \right) = \text{Tr} \left( \frac{\partial}{\partial A} \right)^k = \sum_{i_1, \ldots, i_k} \frac{\partial}{\partial A_{i_1 i_2}} \frac{\partial}{\partial A_{i_2 i_3}} \cdots \frac{\partial}{\partial A_{i_k i_1}} \quad (43)$$

on functions of Hermitian matrices $f(A)$ whose function space is spanned by the basis $\{ e^{\alpha \text{Tr} AB} \}_{B \in \mathbb{H}_N}$. For $\alpha = i$, any Hermitian matrix-valued function belongs to that space by Fourier transformation on Hermitian matrices (see e.g. [20])

$$f(A) = \frac{1}{2N_\pi N^2} \int_{\mathbb{H}_N} dB \tilde{f}(B) e^{i \text{Tr} AB}. \quad (44)$$

Consider now the representation of the operator $\hat{D}_k$ on the subspace of $U(N)$-invariant functions $f(A) = f(U A U^{-1})$, or equivalently

$$f(A) = \int_{U(N)} dU f(U A U^{-1}) = \frac{1}{2N_\pi N^2} \int_{\mathbb{H}_N} dB \tilde{f}(B) \int_{U(N)} dU e^{i \text{Tr}(U A U^{-1} B)}. \quad (45)$$

The domain of such invariant functions is $N$-dimensional, and the functions can be represented on diagonal matrices. The defining equation Eq. (42) on the invariant subspace thus has the form

$$\hat{D}_k \int_{U(N)} dU e^{i \text{Tr}(U A U^{-1} B)} \equiv \alpha^k \text{Tr}(B^k) \int dU e^{i \text{Tr}(U A U^{-1} B)} \quad (46)$$

Using the Itzykson-Zuber integral formula Eq. (15), it follows directly from a Laplace expansion of the determinant,

$$\sum_l \left( \frac{\partial}{\partial a_{1l}} \right)^k \delta(e^{a_{m_1} b_{n_1}})_{m,n} = \sum_l \left( \frac{\partial}{\partial a_{1l}} \right)^k \sum_j (-1)^{l+j} e^{a_{l_1} b_{j}} \delta(e^{a_{m_1} b_{n_1}})_{m \neq l, n \neq j} \quad (47)$$

$$= \sum_j (\alpha b_j)^k \sum_l (-1)^{l+j} e^{a_{l_1} b_{j}} \delta(e^{a_{m_1} b_{n_1}})_{m \neq l, n \neq j} = \alpha^k \text{Tr}(B^k) \delta(e^{a_{m_1} b_{n_1}})_{m,n},$$

that the functional of eigenvalue derivatives $D_k(\frac{\partial}{\partial a_{1l}})$ Eq. (40) is the representation of $\hat{D}_k$ on $U(N)$-invariant functions in diagonalized variables, that is on eigenvalues.

Another equivalent way to prove this formula is to see the function $f$ as a function of the matrix elements $\{ A_{i,j} \}$ in the canonical basis and consider the eigenvalues $\{ a_1, \ldots, a_N \}$ of $A \in \mathbb{H}_N$, together with $N(N-1)$ variables that determine uniquely $U \in U(N)$ such that $A = U \text{Diag} \{ \{ a_i \} \} U^\dagger$, as new variables after a change of variables (see the detailed discussion in Chapter 3 of [23]). Differentiating $\sum_{l,k} U_{il}^\dagger A_{lk} U_{kj}$ with respect to the $a_p$, we find $\sum_{l,k} U_{il}^\dagger \frac{\partial A_{lk}}{\partial a_p} U_{kj} = \delta_{ij} \delta_{jp}$, which is...
inverted as $\frac{\partial A_{ij}}{\partial a_p} = U_{ip}U^\dagger_{pj}$. From this, the derivative of any function of $A$, $\frac{\partial}{\partial a_p} f(\{A_{ij}\})$ is well-defined, and we may verify explicitly that both the right-hand-side $D^{(1)}$ of Eq. (43) acting on a unitary invariant function expressed as Eq. (44), and the right-hand-side $D^{(2)}$ of Eq. (40) acting on Eq. (45) expressed in eigenvalue-variables using Eq. (15) give the same expression:

$$D^{(1)} f(A) = D^{(2)} f(A) = \frac{i}{2N!^2} \int_{\mathbb{H}_N} dB \tilde{f}(B) \text{Tr}(B^k)e^{k \text{Tr}(AB)}.$$  

(48)

This requires using the fact that the Fourier transform $\tilde{f}(B)$ is also unitary invariant.

We use this formula to diagonalize the differential formulations. For a one-matrix model, applying Eq. (40) to Eq. (22), we obtain

$$Z_{V_1, V_2} = e^N \left[ e^{\frac{1}{\sqrt{N}} \text{Tr} \frac{\partial^2}{\partial x^2}} e^{NV_1(M)} \right]_{M=0} = e^N \left[ \frac{1}{\Delta(x)} e^{\frac{1}{\sqrt{N}} \left( \frac{\partial}{\partial x} \right)^2 \Delta(x)} e^{NV_1(x)} \right]_{x=0}$$  

(49)

where $(\frac{\partial}{\partial x})^2 = \sum_i \frac{\partial^2}{\partial x^2}$. The formula is clearly well-normalized, since $(\frac{\partial}{\partial x})^2 \Delta(x) = 0$. Note that the differential reformulation of the usual integral over eigenvalues Eq. (6) reads

$$Z_V = \int_{\mathbb{R}^N} \frac{dx}{b_N} e^{-\frac{N}{2} x^2} \Delta^2(x) e^{NV(x)} = e_N \left[ e^{\frac{1}{\sqrt{N}} \left( \frac{\partial}{\partial x} \right)^2 \Delta^2(x)} e^{NV(x)} \right]_{x=0},$$  

(50)

where the normalization $e_N = N^{(N-1)/2} / \prod_{j=1}^N j!$ is computed in Appendix A.

For the diagonalization of two-matrix models it is crucial that we have a differential formulation using only one matrix such as Eq. (33). It is not possible to simultaneously diagonalize differential operators involving two matrices such as $\text{Tr}(\frac{\partial}{\partial A} \frac{\partial}{\partial B})$ in Eq. (30), at least not by the arguments used here. However, the differential formulation in terms of a single matrix allows us to diagonalize two-matrix models.

Applying Eq. (41) to Eq. (33) yields

$$Z_{V_1, V_2} = e^N \left[ e^{\frac{1}{\sqrt{N}} \text{Tr} \frac{\partial^2}{\partial x^2}} \Delta(x) e^{NV_1(x)} \right]_{x=0}.$$  

(51)

For comparison, the differential formulation of the usual integral over eigenvalues Eq. (15) is

$$Z_{V_1, V_2} = e_N \left[ e^{\frac{1}{\sqrt{N}} \sum_{i=1}^{N} \frac{\partial}{\partial a_i} \frac{\partial}{\partial b_i}} \Delta(a) \Delta(b) e^{NV_1(a)} e^{NV_2(b)} \right]_{a=b=0}$$  

(52)

where the normalization is the same as for the one-matrix model (see Appendix A).

The equivalence between Eq. (49) and Eq. (50) on one hand, and between Eq. (51) and Eq. (52) on the other hand is not manifest in this form, however it becomes clear when formulated in terms of Slater determinants, as detailed later in Sec. IV. 1. (c) for the two-matrix case and in Sec. IV. 3. (a) for the one-matrix model.
IV. Expressions as Slater determinants and applications

IV. 1. Equivalent formulations of a two-matrix model

In this first subsection, we detail the steps leading to the Slater determinant formulation of the partition functions of two-matrix models, in differential formulation. The same holds for the one-matrix models, as the special case $V_1(x) = \frac{x^2}{2}$ (cf. Sec. IV. 3. (a)).

(a) Determinant form. Starting from Eq. (51):

\[
Z_{V_1, V_2} = \left[ \frac{1}{\Delta(x)} \left( \prod_{i=1}^{N} e^{NV_1(\frac{1}{\sqrt{N}} \frac{d}{dx} x_i)} \right) \Delta(x) \left( \prod_{i=1}^{N} e^{NV_2(\frac{1}{\sqrt{N}} x_i)} \right) \right]_{x=0} \quad ,
\]

we incorporate the products on the left and right of the Vandermonde determinant into a Slater determinant:

\[
Z_{V_1, V_2} = \left[ \frac{1}{\Delta(x)} \det \left\{ e^{NV_1(\frac{1}{\sqrt{N}} \frac{d}{dx} x_i) x_j^{j-1}} e^{NV_2(\frac{1}{\sqrt{N}} x_i)} \right\}_{1 \leq i,j \leq N} \right]_{x=0} ,
\]

We can extract the symmetric part of the determinant by using the following identity (27, Thm. 1.2.4 p. 24):

\[
\left[ \frac{1}{\Delta(x)} \det (f_{j}(x_i)) \right]_{\forall i, x_i = x} = f_N \det (f_j^{(i-1)}(x)) ,
\]

with $f_N = 1/\prod_{j=1}^{N-1} j!$. This leads to the following Slater determinant:

\[
Z_{V_1, V_2} = f_N \det \left\{ \left[ \frac{d^i}{dx^i} e^{NV_1(\frac{1}{\sqrt{N}} \frac{d}{dx} x_i) x_j^{j-1}} e^{NV_2(\frac{1}{\sqrt{N}} x_i)} \right]_{x=0} \right\} ,
\]

This is again well-normalized, since for $V_1 = V_2 = 0$ the determinant is $\prod_{i=1}^{N-1} i!$.

(b) Biorthogonal polynomials. As in Sec. II. 2. (e), $\frac{d^i}{dx^i}$ and $x^j$ can be replaced in Eq. (56) by any monic polynomials $P_i(\frac{d}{dx})$, $Q_j(x)$ respectively of degrees $i$ and $j$:

\[
Z_{V_1, V_2} = f_N \det \left\{ \left[ P_i \left( \frac{d}{dx} \right) e^{NV_1(\frac{1}{\sqrt{N}} \frac{d}{dx} x_i) x_j^{j-1}} e^{NV_2(\frac{1}{\sqrt{N}} x_i)} \right]_{x=0} \right\} ,
\]

the differential formulation of the bilinear form of Eq. (20) being:

\[
\langle f | g \rangle = c_1 \left[ f \left( \frac{1}{\sqrt{N}} \frac{d}{dx} \right) e^{NV_1(\frac{1}{\sqrt{N}} \frac{d}{dx} x_i)} g \left( \frac{1}{\sqrt{N}} \frac{d}{dx} \right) e^{NV_2(\frac{1}{\sqrt{N}} x_i)} \right]_{x=0} .
\]

From this, one can apply the method of biorthogonal polynomials.

\[\text{6 The factor } (-1)^{\frac{N(N-1)}{2}} \text{ in the reference is due to a different convention in the definition of the Vandermonde determinant.}\]
(c) Equivalence between the diagonalized formulations for two-matrix models. Expanding the Vandermonde determinants in Eq. (52) as in Sec. II. 2. (d) we obtain:

$$Z_{V_1, V_2} = e_N N! \det \left\{ \left[ e^{\frac{1}{N} \frac{\partial}{\partial a} \frac{\partial}{\partial b} a^i e^{N V_1(a)} b^j e^{N V_2(b)}} \right]_{a=b=0} \right\}_{0 \leq i,j \leq N-1}. \quad (59)$$

This is seen to be equivalent to Eq. (59), since, from Eq. (37) which also holds when $A, B$ are real variables (and seeing $N$ as a real coefficient):

$$\left[ \frac{d^i}{dx^i} e^{N V_1(\sqrt{N} x_i^2) \sqrt{N} x_i^2} \right]_{x=0} = N^{i/2} \left[ e^{\frac{1}{N} \frac{\partial}{\partial a} \frac{\partial}{\partial b} a^i e^{N V_1(a)} b^j e^{N V_2(b)}} \right]_{a=b=0}, \quad (60)$$

which in turn by the change of variable $b = \sqrt{N} b'$ is equal to $N^{i+j} \left[ e^{\frac{1}{N} \frac{\partial}{\partial a} \frac{\partial}{\partial b} a^i e^{N V_1(a)} b^j e^{N V_2(b)}} \right]_{a=b=0}$. The normalizations do agree since $f_N = e_N N! / (N^N(1-N)!/2$.

Eq. (59) is in turn seen to be equivalent to the Hankel determinant Eq. (19), as from Eq. (30) for $N = 1:

$$\left[ e^{\frac{1}{N} \frac{\partial}{\partial a} \frac{\partial}{\partial b} a^i e^{N V_1(a)} b^j e^{N V_2(b)}} \right]_{a=b=0} = \int_{\mathbb{R}^2} \frac{dx dy}{c_1} e^{-N x y^2} e^{N V_1(x)} y^j e^{N V_2(y)}, \quad (61)$$

where we recall that $c_1 = \frac{2^N}{N}$, so that $e_N N! / (c_1)^N = N^{N(N-1)/2} f_N / (c_1)^N = 1/d_N$.

(d) Expansion over Schur functions. Performing steps similar to Sec. II. 2. (d) but in the differential formulation and the opposite direction, we may re-express Eq. (56) as:

$$Z_{V_1, V_2} = \frac{f_N}{N!} \left[ \Delta \left( \frac{\partial}{\partial x} \right) e^{N V_1(\sqrt{N} x / \sqrt{N})} \Delta(x) e^{N V_2(\sqrt{N} x / \sqrt{N})} \right]_{x=0}. \quad (62)$$

As explained in (28), Eq. (2.2.4)), the expression Eq. (62) is easily seen — by its action on Schur functions — to be a scalar product on symmetric functions, with

$$Z_{V_1, V_2} = f_N \left\langle \exp \left( N V_1 \left( \frac{1}{\sqrt{N}} \cdot \right) \right), \exp \left( N V_2 \left( \frac{1}{\sqrt{N}} \cdot \right) \right) \right\rangle. \quad (63)$$

Using the orthogonality of Schur functions for this scalar product and the Cauchy-Littlewood formula, one obtains directly an explicit double-series expansion over Schur functions (Eq. (2.2.13) in 28), which guarantees the fact that $Z_{V_1, V_2}$ is a tau function of the Toda hierarchy.

IV. 2. Integration over one variable in the determinant form of two-matrix models

In order to verify the equivalence — in differential formulation — between Eq. (49) and Eq. (50), it is required to go to Slater determinant form, and then transform certain derivatives to variables. We do this for general $V_1$, as the computation is the same in differential formulation for general $V_1$ or for $V_1(x) = x^2/2$ (corresponding to the one-matrix models). We also detail the computation in the integral formulation for general $V_1$, which is naturally more involved than for $V_1(x) = x^2/2$, in which case it is a simple Gaussian integration. This leads to a new determinant formulation of the partition functions of two-matrix models, and we comment on its potential use for resolutions involving orthogonal polynomials instead of biorthogonal polynomials.
(a) General computation. We start from Eq. (56):

\[ Z = f_N \det_{i,j} \left\{ \left[ \frac{d^i}{dx^i} e^{NV_1(\frac{d}{dx})} x^i e^{NV_2(\frac{d}{dx})} \right]_{x=0} \right\} = f_N \det_{i,j} \left\{ \left[ \frac{d^i}{dx^i} e^{NV_1(\frac{d}{dy})} x^i e^{NV_2(x)} \right]_{y=0} \right\} , \]  

(64)

where 0 ≤ i, j ≤ N − 1, by change of variable\(^7\). In this subsection, we assume \( V_1 \) to be a polynomial:

\[ V_1(x) = \sum_{k=1}^{p} \alpha_k x^k, \quad p \geq 1. \]  

(65)

We write, as in Eq. (37) but for real variables:

\[ \left[ \frac{d^i}{dx^i} e^{NV_1(\frac{d}{dy})} x^i e^{NV_2(x)} \right]_{x=0} = \left[ \frac{d^i}{dy^i} e^{NV_1(\frac{d}{dy})} y^i e^{NV_2(x)} \right]_{y=0} . \]  

(66)

We first notice that:

\[ \frac{d^r}{dy^r} e^{NV_1(\frac{d}{dy})} = (\sqrt[N]{y}^r \alpha_p)^r S_{(p-1)r}(y) e^{NV_1(y)} , \]  

(67)

where \( S_{(p-1)r}(y) \) is a monic polynomial of degree \((p-1)r\).

Let us consider the row \( i = (p-1)r + s \), where \( s \in \{0, 1, \ldots, p-1\} \). We can replace in the determinant the right-hand side of Eq. (66) by \( \left[ \frac{d^r}{dy^r} y^s S_{(p-1)r}(y) e^{NV_1(\frac{d}{dy})} x^i e^{NV_2(x)} \right]_{y=0} \) and therefore by

\[ Z = f_N \det_{i,j} \left\{ (\sqrt[N]{y}^{-p} \alpha_p)^{-r(i)} \left[ \frac{d^r}{dy^r} y^s S_{(p-1)r}(y) e^{NV_1(\frac{d}{dy})} x^i e^{NV_2(x)} \right]_{y=0} \right\} . \]  

(68)

We now show that:

\[ \left[ \frac{d^r}{dy^r} y^s e^{NV_1(\frac{d}{dy})} x^i e^{NV_2(x)} \right]_{y=0} = e^{NV_1(\frac{d}{dy})} x^{r+i} \frac{d^s}{dx^s} \left[ x^i e^{NV_2(x)} \right]_{x=0} , \]  

(69)

that is, just as for Eq. (57) or Eq. (66), the result of the inner bracket \( \left[ \ldots \right]_{y=0} \) on the left hand side is obtained by “replacing” the \( y \)'s by \( \frac{d^i}{dx^i} \)'s, but also the \( \frac{d^r}{dy^r} \)'s by \( d^i \)'s, while reversing the left-to-right ordering.

Developing first the right hand side, we see that:

\[ e^{NV_1(\frac{d}{dx})} x^r \frac{d^s}{dx^s} \left[ x^i e^{NV_2(x)} \right]_{x=0} = \sum_{(n_k \geq 0)_k} \prod_{k=1}^{p} \frac{(N^{1-k} \alpha_k/k)^{n_k}}{n_k!} \left[ \frac{d^{\ell(n)}}{dx^{\ell(n)}} x^r \frac{d^s}{dx^s} \left[ x^i e^{NV_2(x)} \right]_{x=0} \right] , \]  

(70)

where \( \ell(n) = \sum_{k=1}^{p} k n_k \). The bracket on the right hand side of this equation is non-vanishing only if \( \ell(n) = \sum_{k=1}^{p} k n_k \geq r \) and \( r \) of the derivatives act on \( x^r \), in which case:

\[ \left[ \frac{d^{\ell(n)}}{dx^{\ell(n)}} x^r \frac{d^s}{dx^s} \left[ x^i e^{NV_2(x)} \right]_{x=0} \right]_{x=0} = r! \left( \frac{\ell(n)}{r} \right) \left[ \frac{d^{\ell(n)-r+s}}{dx^{\ell(n)-r+s}} x^i e^{NV_2(x)} \right]_{x=0} . \]  

(71)

\(^7\) The factors \( \sqrt[N]{y}^{-1-i} \) from the change of variable compensate when factorized out of the determinant.
The right hand side of Eq. (69) is therefore equal to:

\[
\sum_{\{n_k \geq 0 \mid \ell(n) \geq r\}} \prod_{k=1}^{p} \frac{(N^{1-k} \alpha_k/k)^{n_k}}{n_k!} \frac{\ell(n)!}{(\ell(n) - r)!} \left[ \frac{d^{\ell(n)-r+s}}{dx^{\ell(n)-r+s}} x^j e^{NV_2(x)} \right]_{x=0}.
\]  

(72)

We now focus on the left hand side of Eq. (69):

\[
\left[ \frac{d}{dx} \frac{d}{dy} y^s \frac{d^r}{dy^r} e^{NV_1(\frac{1}{N} y)} \right]_{y=0} = \sum_{q \geq 0} \frac{1}{q!} \left[ \frac{d^q}{dy^q} y^s \frac{d^r}{dy^r} e^{NV_1(\frac{1}{N} y)} \right]_{y=0} d^q y^s = \sum_{q \geq s} \frac{s!}{q!} \left( \frac{d^{q+r-s}}{dy^{q+r-s}} e^{NV_1(\frac{1}{N} y)} \right) d^q y^s.
\]  

(73)

in which the three brackets are evaluated in \( y = 0 \). We develop:

\[
\left[ \frac{d^{q+r-s}}{dy^{q+r-s}} e^{NV_1(\frac{1}{N} y)} \right]_{y=0} = \sum_{\{n_k \geq 0\}} \prod_{k=1}^{p} \frac{(N^{1-k} \alpha_k/k)^{n_k}}{n_k!} \left[ \frac{d^{q+r-s}}{dy^{q+r-s}} y^{\ell(n)} \right]_{y=0} = \sum_{\{n_k \geq 0\} \text{ s.t.} \ell(n) = q-r-s} \prod_{k=1}^{p} \frac{(N^{1-k} \alpha_k/k)^{n_k}}{n_k!} \ell(n)!
\]  

(74)

(75)

Replacing this in Eq. (73) and then Eq. (73) in the left hand side of Eq. (69) leads to the same expansion Eq. (72). We have shown so far that:

\[
Z_{V_1,V_2} = f_N \det \left\{ M_{i,j} \right\}_{0 \leq i,j \leq N-1},
\]  

(76)

where:

\[
M_{(p-1)r+s,j} = (N^{p-1}/\alpha_p)^r \left[ e^{NV_1(\frac{1}{N} y)} \right]_{x=0} \frac{d^s}{dx^s} \left[ x^j e^{NV_2(x)} \right]_{x=0} = (N^{p-1}/\alpha_p)^r \int_{\mathbb{R}^2} \frac{dx \, dy}{2\pi i} e^{-xy} e^{NV_1(\frac{1}{N} y)} x^j e^{NV_2(x)}.
\]  

(77)

(78)

The equivalence between Eq. (77) and Eq. (78) is the usual one-matrix differential formulation of two-matrix models of Sec. 11.2 but for \( A, B \) two real variables, and seeing \( x^r \frac{d^s}{dx^s} [x^j e^{NV_2(x)}] \) as a function of \( x \).

(b) Computation in the integral formulation. Starting from Eq. (19), we change variable as \( y' = Ny \) and include the normalization by \( 2\pi i \):

\[
Z_{V_1,V_2} = f_N \det \left\{ \int_{\mathbb{R}^2} \frac{dx \, dy}{2\pi i} e^{-xy} x^i e^{NV_1(\frac{1}{N} x)} y^j e^{NV_2(y)} \right\}_{0 \leq i,j \leq N-1},
\]  

(79)

where we have used the fact that \( f_N = (2\pi i)^N d_N^{-1} N^{-N(N+1)/2} \). The steps up to equation Eq. (68) are precisely the same in integral form, so we do not detail them again. While a perturbative proof of Eq. (69) is certainly possible, it does not seem to us that the proof above can directly be translated in integral form (in particular the steps Eq. (73) to Eq. (75)), so we propose a different proof. In integral form, Eq. (68) reads

\[
Z_{V_1,V_2} = f_N \det \left\{ (N^{1-p} \alpha_p)^{-r(i)} \int_{\mathbb{R}^2} \frac{dx \, dy}{2\pi i} e^{-xy} y^i \left[ \frac{d^{r(i)}}{dy^{r(i)}} e^{NV_1(y)} \right] x^j e^{NV_2(x)} \right\}_{0 \leq i,j \leq N-1}.
\]  

(80)
By performing $r$ integrations by part:
\[ \int dx \, dy \, e^{-xy} y^s \left[ \frac{d^r}{dy^r} e^{NV_1(y)} \right] x^r e^{NV_2(x)} = (-1)^r \int dx \, dy \, \left[ \frac{d^r}{dy^r} e^{-xy} y^s \right] e^{NV_1(y)} x^j e^{NV_2(x)}, \] (81)
where the boundary terms are assumed to vanish every time due to the exponential terms. We use the following identity:
\[ \frac{d^r}{dy^r} [(ay)^{s} e^{axy}] = \frac{d^s}{dx^s} [(ax)^r e^{axy}]. \] (82)
Indeed, using Leibniz formula:
\[ \frac{d^r}{dy^r} [(ay)^{s} e^{axy}] = \sum_{k=0}^{r} \binom{r}{k} a^s \frac{d^k}{dy^k} [y^s] \frac{d^{r-k}}{dy^{r-k}} [e^{axy}] = e^{axy} \sum_{k=0}^{\min(r,s)} \frac{1}{k!} \frac{s!}{(s-k)!} \frac{r!}{(r-k)!} (axy)^k. \] (83)
The same expression is obtained developing $\frac{d^s}{dx^s} [(ax)^r e^{axy}]$. Using this identity, Eq. (81) is equal to:
\[ (-1)^s \int dx \, dy \, e^{NV_1(y)} \left[ \frac{d^s}{dx^s} e^{-xy} x^r \right] x^j e^{NV_2(x)} = \int dx \, dy \, e^{-xy} y^s e^{NV_1(y)} \left[ \frac{d^s}{dx^s} x^r e^{NV_2(x)} \right], \] (84)
where the equality is obtained by doing again $s$ integrations by part. We thus recover Eq. (78).

**IV. 3. Potential application to new orthogonal polynomial method**

(a) **Equivalence between the diagonalized differential formulations for one-matrix models.** Performing the steps of Sec. IV.1.(a), but starting from the differential formulation Eq. (49) of one-matrix models leads to:
\[ Z_V = f_N \det \left\{ \left[ \frac{d^i}{dx^i} \frac{d^j}{dx^j} x^i e^{NV(x)} \right] \right\}_{x=0}^{0 \leq i,j \leq N-1}. \] (85)
This is also the differential formulation of the Slater determinant expression Eq. (56) for a two-matrix model with $V_1(x) = \frac{x^2}{2}$.

On the other hand, performing steps similar to Sec. II.2.(d) to express Eq. (50) in determinant form, we obtain:
\[ Z_V = N! e_N \det \left\{ \left[ \frac{1}{e^{2N \frac{d^2}{dx^2}}} \frac{d^i}{dx^i} x^{i+j} e^{NV(x)} \right] \right\}_{x=0}^{0 \leq i,j \leq N-1}. \] (86)
From the simpler formulation Eq. (86), one can use orthogonal polynomials satisfying
\[ e^{2N \frac{d^2}{dx^2}} P_i(x) P_j(x) e^{NV(x)} \bigg|_{x=0} = \delta_{ij} p_i \] (87)
instead of biorthogonal polynomials Eq. (20) satisfying
\[ P_i \left( \frac{d}{dx} \right) e^{2N \frac{d^2}{dx^2}} Q_j(x) e^{NV(x)} \bigg|_{x=0} = \delta_{ij} p_i, \] (88)
The computations of Sec. [V.2] in the differential formulation, in the case where \( V_1(x) = \frac{x^2}{2} \) (\( p = 2, \alpha = 1 \)), prove the equivalence between Eq. (53) and Eq. (86). Indeed for \( p = 2 \), \( r(i) = i \) so that the coefficients \( (N^{p-1}/\alpha^p)^{r(i)} = N^i \) in the determinant just provide an overall factor \( N^{N(N-1)/2} \), and \( f_N N^{N(N-1)/2} = N! e_N \).

In integral form, the proof is simpler, as the Gaussian integration for \( V_1(x) = \frac{x^2}{2} \) can be carried out explicitly after changing variables for \( x' = N x / \alpha \). Equivalently, one may recover directly the Slater determinant form of the partition function of one-matrix models in the integral formulation, that is

\[
Z_{V_2} = \left( \frac{N!}{b_N} \right) \det \left\{ \int dy e^{-N^2 \frac{x^2}{2}} e^{N V_2(y)} y^i j \right\} _{0 \leq i, j \leq N-1} , \tag{89}
\]

from the differential formulation of the Slater determinant Eq. (56) (after the change of variables Eq. (84)),

\[
Z_{x^2/2, V_2} = f_N \det \left\{ \left[ \frac{d^i}{dx^i} e^{\frac{x^2}{2N} \sum_j x^j e^{N V_2(x)}} \right] _{x=0} \right\} _{0 \leq i, j \leq N-1}, \tag{90}
\]

by using the heat kernel formulation:

\[
e^{\frac{d^2}{dx^2}} F(x) = \int dy K_t(x - y) F(y), \quad K_t(x - y) = \frac{1}{\sqrt{4\pi t}} e^{-\frac{(x-y)^2}{4t}}, \tag{91}
\]

for \( t = \frac{1}{2N} \) and \( F(x) = x^i e^{N V_2(x)} \). We get:

\[
Z_{x^2/2, V_2} = f_N \det \left\{ \sqrt{\frac{N}{2\pi}} \int dy \left[ \frac{\partial^i}{\partial x^i} e^{-\frac{N(x-y)^2}{2}} \right] _{x=0} y^j e^{N V_2(y)} \right\} _{0 \leq i, j \leq N-1}, \tag{92}
\]

where the bracket evaluating \( x \) at 0 has been moved inside the integral, as only the derivative of the kernel depends on \( x \). One can verify that

\[
\left[ \frac{\partial^i}{\partial x^i} e^{-\frac{N(x-y)^2}{2}} \right] _{x=0} = N^i \hat{Q}_i(y) e^{-\frac{N}{2} y^2}, \tag{93}
\]

where \( \hat{Q}_i(y) \) is a monic polynomial of degree \( i \), so we do recover Eq. (89) as \( f_N (\frac{N}{2\pi})^\frac{N}{2} N^{N(N-1)/2} = N! / b_N \).

One can also use the heat kernel to find the integral representation of Eq. (49) and the one-matrix version of Eq. (54), again obtained by setting \( V_1(x) = \frac{x^2}{2} \). Applying the \( N \)-dimensional version of Eq. (91) to Eq. (49) for each \( x_i \) we find

\[
Z_V = \left( \frac{N}{2\pi} \right) \frac{N}{2} \left[ \frac{1}{\Delta(x)} \int dy e^{-\frac{N(x-y)^2}{2}} \Delta(y) e^{N V(y)} \right] _{x=0} , \tag{94}
\]

Similarly, in the one-matrix version of Eq. (54) we can use the heat kernel to find

\[
Z_{x^2/2, V_2} = \left[ \frac{1}{\Delta(x)} \det \left\{ \frac{1}{\sqrt{2\pi}} \int dy e^{-\frac{(x-y)^2}{2}} y^j e^{N V_2 \left( \frac{1}{\sqrt{N}} \right) y} \right\} _{1 \leq i, j \leq N} \right] _{x=0}. \tag{95}
\]

---

\*There is an important subtlety in changing variables from a real variable to a pure complex variable: in addition to the Jacobian of the change of variables, an additional factor \(-1\) has to be taken into account. This is detailed in Appendix [B] see Eq. (B11). The problem does not appear for the one-matrix model in the computations in differential formulation of Sec. [IV.2] or using the heat kernel Eq. (91).
(b) Application to new orthogonal polynomial methods. As we have seen at the beginning of the present section, the computations of Sec. IV.2 in the case where \( V_1(x) = \frac{x^2}{2} \), prove the equivalence in the differential formulation between Eq. (55) and Eq. (56), and the latter expression allows using orthogonal polynomials Eq. (77) instead of biorthogonal polynomials Eq. (87). One may naturally wonder if for potentials \( V_1 \) of degree higher than two, our new expressions Eq. (77), Eq. (78) could also allow the use of orthogonal polynomials instead of biorthogonal polynomials for two-matrix models.

To this aim, we reorganize the lines of the matrix \( M \) in Eq. (77) according to their remainder modulo \( p - 1 \), that is:

\[
Z_{V_1,V_2} = h_N \det \{ \tilde{M}_{i,j} \}_{0 \leq i,j \leq N-1}, \quad \text{where} \quad \tilde{M} = \begin{pmatrix} T_0 \\ \vdots \\ T_{p-1} \end{pmatrix} \quad \text{and} \quad (T_s)_{r,j} = M_{(p-1)r+s,j}. \tag{96}
\]

where \( h_N = f_N \text{sgn}(N,p), \text{sgn}(N,p) \) being the parity of the permutation of rows, and we recall (Eq. (77)) that in differential formulation,

\[
M_{(p-1)r+s,j} = (N^{p-1}/\alpha_p)^r \left[ e^{NV_1(x)} x^r \frac{d^s}{dx^s} [x^j e^{NV_2(x)}] \right]_{x=0}. \tag{97}
\]

If \( s_0 \) and \( r_0 \) are respectively the remainder and the quotient of the Euclidean division of \( N - 1 \) by \( p - 1 \), then for \( 0 \leq s \leq s_0 \), the matrix \( T_s \) has \( r_0 + 1 \) lines, while for \( s_0 < s < p - 1 \), \( T_s \) has \( r_0 \) lines.

By reorganizing the columns of \( \tilde{M} \), and the lines of \( T_s \) independently for each \( s \), we may replace \( (N^{p-1}/\alpha_p)^r (T_s)_{r,j} = (N^{p-1}/\alpha_p)^{-r} M_{(p-1)r+s,j} \) in the determinant by

\[
\left[ e^{NV_1(x)} (x) \frac{d^s}{dx^s} [Q_j(x) e^{NV_2(x)}] \right]_{x=0} = \int \frac{dx dy}{2\pi i} e^{-xy} e^{NV_1(x)} P_r(x) e^{NV_2(x)} \int dy e^{-xy} e^{NV_1(y)} Q_j(x) e^{NV_2(x)}, \tag{98}
\]

where for \( 0 \leq s < p - 1 \), the \( P_r(s)(x) \) are monic polynomials of degree \( r \), and \( Q_j \) is a monic polynomial of degree \( j \).

One may for instance choose \( P_r(s)(x) = Q_r \) for all \( 0 \leq s < p - 1 \). We use the following notation for the determinant obtained this way:

\[
Z_{V_1,V_2} = h_N \det \{ W_{i,j} \}_{0 \leq i,j \leq N-1}, \quad \text{where} \quad W = \begin{pmatrix} J_0 \\ \vdots \\ J_{p-1} \end{pmatrix}. \tag{99}
\]

For \( s = 0 \), the elements \( (J_0)_{r,j} \) of the resulting matrix \( J_0 \) define the symmetric bilinear form

\[
\langle Q_r \parallel Q_j \rangle = \int dx Q_r(x) Q_j(x) = \frac{1}{2\pi} \int dx Q_r(x) Q_j(x) e^{NV_2(x)} \int dy e^{-xy} e^{NV_1(y)}, \tag{100}
\]

and requiring \( \langle Q_r \parallel Q_j \rangle = h_r \delta_{r,j} \) defines a family of orthogonal polynomials for the following weight:

\[
e^{NV_1(x)} e^{NV_2(x)} \leftrightarrow e^{NV_2(x)} \int dy e^{-xy} e^{NV_1(y)}. \tag{101}
\]
In addition to the usual three-terms recurrence $Q_{n+1}(x) = (x - \beta_n)Q_n - \frac{h_n}{h_{n-1}}Q_{n-1}$ satisfied by any family of orthogonal polynomials, to obtain families of recurrence relations on the coefficients $\beta_n$ and $h_n$, one may for instance use the fact that

$$\left[e^{N V_1(\frac{1}{N} \frac{d}{dx})} V_1^{(r)} \left( \frac{1}{N} \frac{d}{dx} \right) F(x) \right]_{x=0} = \left[e^{N V_1(\frac{1}{N} \frac{d}{dy})} V_1^{(r)} \left( \frac{1}{N} \frac{d}{dy} \right) e^{N V_1(\frac{1}{N} y)} F(x) \right]_{y=0} \quad (102)$$

so that

$$\left[e^{N V_1(\frac{1}{N} \frac{d}{dx})} V_1^{(r)} \left( \frac{1}{N} \frac{d}{dx} \right) F(x) \right]_{x=0} = \left[e^{N V_1(\frac{1}{N} \frac{d}{dx}) x F(x)} \right]_{x=0}, \quad (103)$$

applied to $F(x) = Q_r(x)Q_j(x)e^{NV_2(x)}$ for any $r, j$. For instance for the case $V_1(x) = x^3/3$, for which the weight for the orthogonal polynomials includes the Airy function as a factor, this equation leads to the family of equations

$$\frac{2}{N} \langle V_2' Q_r' \parallel Q_j \rangle + \frac{2}{N} \langle V_2 Q_r || Q_j' \rangle + \frac{1}{N^2} \left( \langle Q_r'' || Q_j \rangle + \langle Q_r || Q_j'' \rangle + 2 \langle Q_r' || Q_j' \rangle + \langle R Q_r || Q_j \rangle \right) = 0, \quad (105)$$

with $R(x) = \frac{1}{N} V_2''(x) + (V_2'(x))^2 - x$.

The rest of the determinant Eq. (103) for $s > 0$, obtained from Eq. (102), also involves sums of terms of the form $\langle g Q_r || Q_j^{(k)} \rangle$ for some polynomials $g(x)$. Computing the determinant therefore requires knowledge about the derivatives of the families of orthogonal polynomials with weight Eq. (104). It is not clear at this point whether the family of orthogonal polynomials can indeed be constructed and whether the rest of the determinant can be computed this way.

On the other hand, everything seems to simplify drastically at large $N$: for the choice $P_r^{(s)} = Q_r$ for all $0 \leq s < p - 1$, the leading contributions in $N$ to the matrix elements Eq. (105) for the part of the determinant corresponding to $s > 0$ are obtained when all $s$ derivatives act on $e^{NV_2(x)}$, raising a factor $N^s(V_2'(x))^s$, so that the elements of the determinant simplify to

$$(J_{s})_{r,j} \sim_{N \to \infty} \frac{N^{(p-1)r+s}}{\alpha_r^{p}} \langle (V_2')^s Q_r \parallel Q_j \rangle. \quad (106)$$

In the same way, the leading contribution to the family of relations Eq. (106) seems to be given by

$$\left[e^{N V_1(\frac{1}{N} \frac{d}{dx}) e^{NV_2(x)} V_1^{(r)} (V_2'(x)) Q_r(x)Q_j(x)} \right]_{x=0} = \left[e^{N V_1(\frac{1}{N} \frac{d}{dx}) x F(x)} \right]_{x=0}, \quad \Leftrightarrow \quad \langle U Q_r \parallel Q_j \rangle = 0, \quad (107)$$

with $U(x) = V_1'(V_2'(x)) - x$. However, this already goes beyond the original scope of the paper, and we leave this for future work.

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A. Normalizations of the differential formulations

Normalization of the differential formulation of diagonalized one-matrix models.
The normalization $e_N$ in Eq. (50) is computed developing the Vandermonde determinants as in Sec. II.2. (d)

$$\frac{1}{e_N} = \left[ e^{N(\frac{\partial}{\partial x})^2} \Delta^2(x) \right]_{x=0} = N! \det_{i,j} \left\{ \left[ e^{N(\frac{\partial}{\partial x})^2} x_i^j \right]_{x=0} \right\} = N! \det_{i,j} \left\{ \sum_{n \geq 0} \frac{(2N)^{-n}}{n!} \left[ \frac{d^{2n}}{d^2x} x_i^j \right]_{x=0} \right\},$$

where the indices in the determinants range from 0 to $N-1$. The bracket in the rightmost expression is non-vanishing only if $2n = i + j$, in which case $\left[ \frac{d^{2n}}{d^2x} x_i^j \right]_{x=0} = (i + j)!$. Therefore, $1/e_N = N! \det(R)(2N)^{-N(N-1)/2}$, where the $N \times N$ matrix $R$ is such that $R_{ij} = 0$ for $i + j$ odd, and $R_{ij} = (i + j)!/[(i + j)/2]!$ for $i + j$ even. This determinant can be computed to be $\det(R) = 2^{N(N-1)/2} \prod_{i=1}^{N-1} j!$, so that $e_N = N^{N(N-1)/2} / \prod_{i=1}^{N-1} j!$.

Normalization of the differential formulation of diagonalized two-matrix models. The normalization is the same as for the one-matrix model, since:

$$\left[ e^{N \sum_{i=1}^{N} \frac{\partial}{\partial a_i} \frac{\partial}{\partial b_i}} \Delta(a) \Delta(b) \right]_{a=b=0} = N! \det_{i,j} \left\{ e^{N \sum_{i=1}^{N} \frac{\partial}{\partial a_i} \frac{\partial}{\partial b_i} a_i b_i} \right\}_{a=b=0} = N! \det_{i,j} \left\{ \sum_{n \geq 0} \frac{(N)^{-n}}{n!} \left[ \frac{d^n}{da^n} a_i b_i \right]_{a=0} \left[ \frac{d^n}{db^n} b_i b_i \right]_{b=0} \right\} = N! \det_{i,j} \left\{ \delta_{i,j} i!N^{-i} \right\} = \frac{1}{e_N}. $$

B. Alternative representations using generalized heat kernels

In Sec. IV.3. (a) we have detailed how, for one-matrix models, the Slater determinant form Eq. (89) in the integral formulation can be recovered directly from that in differential formulation Eq. (90) by using the heat kernel formula Eq. (91).

In this appendix, we detail for completeness the analogous computation for two-matrix models, that is, the integral analogues of expressions Eq. (51), Eq. (54), and Eq. (56) using generalized heat kernels. Together these form the back lower edge of the diagram shown in Fig. 2.
Our starting point is the differential representation Eq. [51]:
\[ Z_{V_1, V_2} = \left[ \frac{1}{\Delta(x)} e^{NV_1(x)} \Delta(x) e^{NV_2(x)} \right]_{x=0}. \] (B1)

In Eq. [B1], we used the heat kernel \( K_t(x) \) to transform a differential representation of the one-model to an integral form. For two-matrix models, we can use the following analogous identity (see e.g. [29]):
\[ e^{W(d\alpha) F(x)} = \int_{\mathbb{R}^N} dy K_W(x - y) F(y), \] (B2)

where \( W \) is a polynomial, and \( K_W \) is a generalized heat kernel with the following Fourier representation:
\[ K_W(x - y) = \frac{1}{(2\pi)^N} \int_{\mathbb{R}^N} dp e^{ip(x - y) + W(ip)}. \] (B3)

If we set \( W(x) = NV_1 \left( \frac{1}{\sqrt{N}} x \right) \) and \( F(x) = \Delta(x) e^{NV_2 \left( \frac{1}{\sqrt{N}} x \right)} \), we can substitute Eq. [B2] in Eq. [B1] to find
\[ Z_{V_1, V_2} = \left[ \frac{1}{\Delta(x)} \right] \left[ \int_{\mathbb{R}^N} dy \int_{\mathbb{R}^N} dp e^{ip(x - y) + NV_1 \left( \frac{1}{\sqrt{N}} p \right)} \Delta(y) e^{NV_2 \left( \frac{1}{\sqrt{N}} y \right)} \right]_{x=0}. \] (B4)

This is the integral analogue of Eq. [B1]. Subsequently, we can rewrite this expression as a Slater determinant by expanding the Vandermonde determinant, just like we did when rewriting Eq. [51] to Eq. [54], thus obtaining:
\[ Z_{V_1, V_2} = \frac{1}{(2\pi)^N} \left[ \int_{\mathbb{R}^N} dy \int_{\mathbb{R}^N} dp \left\{ e^{ip(x - y)} + NV_1 \left( \frac{1}{\sqrt{N}} p \right) \Delta(y) e^{NV_2 \left( \frac{1}{\sqrt{N}} y \right)} \right\} \right]_{x=0}. \] (B5)

Finally, we can use identity Eq. [55] to obtain
\[ Z_{V_1, V_2} = \frac{f_N}{(2\pi)^N} \det \left\{ \int_{\mathbb{R}^N} dy \int_{\mathbb{R}^N} dp \left[ \frac{\partial^i}{\partial x^j} e^{ip(x - y)} \right]_{x=0} e^{NV_1 \left( \frac{1}{\sqrt{N}} p \right) \Delta(y) e^{NV_2 \left( \frac{1}{\sqrt{N}} y \right)}} \right\}_{0 \leq i, j \leq N-1}, \] (B6)

where the bracket evaluating \( x \) at 0 could now be moved inside the determinant and the integral. This equation mirrors Eq. [51] using the generalized heat kernel. Carrying out the differentiation and evaluating \( x \) at 0:
\[ Z_{V_1, V_2} = \frac{f_N}{(2\pi)^N} \det \left\{ \int_{\mathbb{R}^N} dy \int_{\mathbb{R}^N} dp \left( ip \right)^i e^{ip \Delta(y) e^{NV_1 \left( \frac{1}{\sqrt{N}} p \right)}} \right\}_{0 \leq i, j \leq N-1}. \] (B7)

To verify that the equation is well-normalized we may use Eq. [11] to carry out the integration of the matrix elements. There is a subtlety however: Eq. [9] and Eq. [11] have to be modified for \( \alpha \in i\mathbb{R} \). For \( n = m = 0 \) and \( \alpha \in i\mathbb{R} \), Eq. [9] should be replaced with:
\[ \int_{\mathbb{R}^2} dx \, dy \, e^{-i\alpha N xy} = \frac{2\pi}{\alpha N}, \] (B8)

and for positive \( n \) or \( m \):
\[ \int_{\mathbb{R}^2} dx \, dy \, e^{-i\alpha N xy} x^n y^m = \delta_{n,m} \frac{\partial^n}{(-iN)^n} \int_{\mathbb{R}^2} dx \, dy \, e^{-i\alpha N xy} = \delta_{n,m} \frac{\partial^n}{(-iN)^n} \frac{2\pi}{\alpha N}, \] (B9)
so Eq. (11) has to be modified for:

\[ \int \mathbb{R}^2 \, dx \, dy \, e^{-iaNxy} \, x^n \, y^m = \delta_{n,m} \frac{2 \pi i}{(iaN)^{n+1}} \, n!. \] (B10)

These expressions allow showing that

\[ \int \mathbb{R}^2 \, dy \, dp \, e^{-i(py)} \, (ip)^i \, e^{NV_1(\sqrt{N}x')} \, x'^i \, e^{NV_2(\sqrt{N}y')} \, y'^j = i \int \mathbb{R}^2 \, dx \, dy \, e^{-xy} \, x^i \, e^{NV_1(\sqrt{N}x)} \, x^i \, e^{NV_2(\sqrt{N}y)}, \] (B11)

in the sense that their perturbative expansions match, where Eq. (B10) has to be used to expand the left hand side and Eq. (11) for the right hand side, which explains the factor \( i \), while \( i^{-1} \) would naively be expected as a Jacobian by directly changing variables for \( x = ip \) in the integral. Finally, changing variables for \( \sqrt{N}x' = x \) and \( \sqrt{N}y' = y \), we recover the integral Slater determinant form Eq. (19), since \( \frac{i_N}{(2\pi)^N} N^{N(N+1)/2} N! = \frac{1}{d_N} \).

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