



UNIwersYTET  
WARSAWski

# Dunkl operator and matrix models at finite $N$

2024-StringPoland

---

Pedram Karimi, IFT University of Warsaw

June 7, 2024

# Table of contents

1. Introduction
2. Ring of Symmetric polynomials
3. Dunkl Operator

# Introduction

---

## Correlation functions in Gaussian matrix model

The partition function for the  $\beta = \frac{1}{\alpha}$ -deformed Gaussian eigenvalue model can be written in eigenvalue form

$$Z \stackrel{\text{def}}{=} \int \left( \prod_{i=1}^N dx_i \right) w_{\beta}(x) \exp \left[ - \sum_{i=1}^N \frac{x_i^2}{2} \right]. \quad (1)$$

where  $w_{\beta}(x) = \prod_{1 \leq i < j \leq N} (x_i - x_j)^{2\beta}$  is the Vandermonde and  $x_i$  are diagonalized elements of Hermitian matrix  $H = h_{ij}$ . Our goal is to find a multipoint correlation of the gaussian model

$$\mathbb{E} [ \text{Tr}(H^{k_1}) \dots \text{Tr}(H^{k_l}) ] = \mathbb{E} \left[ \left( \sum_{i=1}^N x_i^{k_1} \right) \dots \left( \sum_{i=1}^N x_i^{k_l} \right) \right] \quad (2)$$

We add generating function with parameter  $q_k$  to the partition function:

$$Z(q_k) = \int \left( \prod_{i=1}^N dx_i \right) w_{\beta}(x) \exp \left[ - \sum_{i=1}^N \frac{x_i^2}{2} + x_i \right] \exp \left[ \beta \sum_{k=1}^{\infty} \frac{q_k}{k} \sum_{i=1}^N x_i^k \right]. \quad (3)$$

## Connection to supersymmetric theory

In 3d  $\mathcal{N} = 2$  supersymmetry on  $S_1 \times_q D_2$ , consider a vector multiplet, an adjoint chiral multiplet of mass  $t$ , and two antichiral multiplet of mass  $u_1, u_2$ . Then the partition function of the theory can be computed using supersymmetric localization. The integrand is

$$Z(\lambda) = \prod_{1 \leq k \neq l \leq N} \frac{(\lambda_k/\lambda_l; q)_\infty}{(t\lambda_k/\lambda_l; q)_\infty} \prod_{j=1}^N \prod_{k=1}^{N_f} (q\lambda_j u_k; q)_\infty, \quad (4)$$

After setting  $q = t^\beta$ ,  $\beta \in \mathbb{Z}$  and sending  $q \rightarrow 1$  we recover the  $\beta$ -model as a semiclassical limit.

# Ring of Symmetric polynomials

---

# Partition and Young diagram

**Symmetric polynomials** is a polynomial with  $n$  variable such that it remains invariant under permutation of variables.

These polynomials with multiplication form a ring of symmetric polynomials.

**Partition** of integer  $n$  is a way of writing  $n$  as a sum of integers. For instance partitions of 3 are  $\{\{3\}, \{2, 1\}, \{1, 1, 1\}\}$ .

**Young diagram** is a set of boxes with left justified rows. Young diagrams can be use for visual representation of partitions of number  $n$ .

$$\{3\} = \begin{array}{|c|c|c|} \hline \square & \square & \square \\ \hline \end{array}, \quad \{2, 1\} = \begin{array}{|c|c|} \hline \square & \square \\ \hline \square & \\ \hline \end{array}, \quad \{1, 1, 1\} = \begin{array}{|c|} \hline \square \\ \hline \square \\ \hline \square \\ \hline \end{array}$$

**(semi-)standard Young tableau** is labeling a boxes of a Young diagram (weakly) increasing to the right and increasing downward.

## Symmetric function I:

There are many symmetric polynomials.

Monomial symmetric functions:

$$\mathbf{m}_\lambda = \sum_{\mu \sim \lambda} x^\mu, \quad (5)$$

where  $\mu \sim \lambda$  means rearrangement of parts of  $\lambda$ .

Powersum symmetric functions:

$$p_\lambda = \prod_{i \in \ell(\lambda)} p_{\lambda_i} \quad \text{and,} \quad p_k = \sum_{i=1}^n x_i^k, \quad (6)$$

where  $\ell(\lambda)$  is the length of partition  $\lambda$  and  $\lambda_i$  its  $i$ -part. We equipped our ring with an inner product  $\langle \bullet, \bullet \rangle$

$$\langle p_\lambda, p_\mu \rangle = z_\lambda \delta_{\lambda, \mu}, \quad z_\lambda = \prod_{i \geq 1} i^{m_i} m_i!, \quad (7)$$

$m_i$  is the number the parts in  $\lambda$  equals to  $i$ .



## Symmetric functions II

**Schur polynomial** of  $n$  variable for a partition  $\lambda$  is

$$S_\lambda(x_1, x_2, \dots, x_n) = \sum_T \prod_{s \in \lambda} x^T = \sum_T x_1^{t_1} \dots x_n^{t_n}, \quad (8)$$

where summation is over all semistandard Young tableaux  $T$  of shape  $\lambda$ , and  $t_i$  is the weight of  $i$  in  $T$ . It counts the number of occurrence of  $i$  in

$T$ . As an example we calculate  $T(\{2\}) = \left\{ \begin{array}{|c|c|} \hline 1 & 1 \\ \hline \end{array}, \begin{array}{|c|c|} \hline 1 & 2 \\ \hline \end{array}, \begin{array}{|c|c|} \hline 2 & 2 \\ \hline \end{array} \right\}$ .

So we have

$$S_{\{2\}} = x_1^2 + x_1x_2 + x_2^2 = \frac{p_1^2}{2} + \frac{p_2}{2}. \quad (9)$$

In the last equality we use the fact that power sum and Schur polynomials form a complete orthonormal basis for the space of symmetric functions

$$\langle S_\lambda, S_\mu \rangle = \delta_{\lambda, \mu}. \quad (10)$$

# Deformation of Hall inner product and Jack polynomials:

We can deform the Hall inner product

$$\langle p_\lambda, p_\mu \rangle_\alpha = \delta_{\lambda, \mu} \alpha^{\ell(\lambda)} z_\lambda, \quad (11)$$

with  $\alpha \in \mathbb{R}$ . This deformation allow us to have a deformation of Schur functions with respect to this new inner product.

**Jack polynomials** defines uniquely by the following conditions:

$$\langle P_\lambda, P_\mu \rangle_\alpha = 0, \text{ if } \lambda \neq \mu,$$

$$P_\lambda = \sum_{\mu \leq_d \lambda} C_{\lambda\mu} \mathbf{m}_\mu$$

$$[\mathbf{m}_\lambda] P_\lambda = 1, \text{ P-normalization.}$$

$$[\mathbf{m}_\lambda] J_\lambda = |\lambda|!, \text{ J-normalization.}$$

## Some examples of Jack polynomials

Jack polynomials are one parameter  $\beta := \frac{1}{\alpha}$  deformation of Schur polynomials and form orthogonal bases for the space of symmetric polynomials. Here is Jack for integer partition of 3

$$P_{\{3\}} = \frac{\beta^2 p_1^3 + 3\beta p_2 p_1 + 2p_3}{\beta^2 + 3\beta + 2}, \quad (12)$$

$$P_{\{2,1\}} = -\frac{-\beta p_1^3 + (\beta - 1)p_2 p_1 + p_3}{2\beta + 1}, \quad (13)$$

$$P_{\{1,1,1\}} = \frac{1}{6} (p_1^3 - 3p_2 p_1 + 2p_3). \quad (14)$$

We write  $P_{\{2\}}$  to compare the result with example we directly calculate for Schur, when  $\beta = 1$

$$P_{\{2\}} = \frac{2\beta x_2 x_1}{\beta + 1} + x_1^2 + x_2^2 = \frac{\beta p_1^2 + p_2}{\beta + 1}. \quad (15)$$

# Connection of symmetric polynomials and correlation functions

As we have already seen correlation function can be seen as a symmetric polynomials labeled by a partition for instance

$$\mathbb{E} [Tr(H^2) Tr(H^2)] = \mathbb{E} \left[ \left( \sum_{i=1}^N x_i^2 \right)^2 \right] = \mathbb{E} [p_2^2] = \langle p_{\{2,2\}} \rangle = \mathbb{E} [p_{\{2^2\}}] \quad (16)$$

Using the fact that Jack polynomials are orthonormal bases we can expand powersum polynomials in these bases

$$p_\lambda = \sum_{\nu \vdash |\lambda|} R_\nu^\lambda P_\nu \quad (17)$$

where  $R_\nu^\lambda$  can be computed algebraically, as an example

$$p_{\{2^2\}} = P_{\{4\}} - \frac{4}{3\alpha + 1} P_{\{31\}} - \frac{4(\alpha^2 + \alpha + 1)}{(\alpha + 1)(2\alpha + 1)} P_{\{2^2\}} - \frac{4\alpha}{(\alpha + 1)^2} P_{\{211\}} + \frac{24\alpha}{\alpha^3 + 6\alpha^2 + 11\alpha + 6} P_{\{1111\}}. \quad (18)$$

**Question** Can we simplify the result of correlation function using other bases? Surprisingly, Yes!

# Dunkl Operator

---

# Dunkl Operator

Dunkl operator defines on using a certain data:

**Definition:**

- Let  $R$  be a root system.
- Let  $G$  be a reflection group on  $R^\vee$
- $k : R \rightarrow \mathbb{C}$  a  $G$ -invariant function.
- $\sigma_\alpha X := X - \frac{2\langle X, \alpha \rangle}{\langle \alpha, \alpha \rangle}$  is a reflection along the root  $\alpha \in R$ .

Then the Dunkl operator for  $\xi \in \mathbb{R}^N$  is

$$T_\xi f(x) := \partial_\xi + \sum_{\alpha \in R_+} k_\alpha \langle \alpha, \xi \rangle \frac{f(x) - f(\sigma_\alpha x)}{\langle \alpha, x \rangle}. \quad (19)$$

We set  $T_{\xi_i} = T_i$  for  $\xi_i \in \mathbb{R}^N$ .

# Why is Dunkl operator interesting?

**Crucial theorem:** Dunkl operators commutes with each others [C.Dunkl]

$$[T_i, T_j] = 0. \quad (20)$$

**Corollary(1):** Set  $\Pi = \mathbb{C}[\mathbb{R}^N]$  to be a  $\mathbb{C}$ -algebra of polynomials. Clearly this is a graded space  $\Pi = \bigoplus_{n \geq 0} p_n$ .

There is an algebra homomorphism  $\varphi : \Pi \rightarrow \text{End } \Pi$  such that:

$$\varphi_k : x_i \mapsto T_i, \quad 1 \mapsto \text{id}. \quad (21)$$

Therefore we can define  $p(T) := \varphi_k P$ .

**Corollary(2):** There exist a symmetric non-degenerate inner product on  $P_i$  called generalized Fischer product

$$[p, q]_k := (p(T)q)(0). \quad (22)$$

Now one can do harmonic analysis on Dunkl operators.

# Dunkl Kernel and Bessel function

**Definition/Theorem** Dunkl Kernel is  $f = E_k(\bullet, y)$  such that

$$T_\xi f = \langle \xi, y \rangle f, \quad f(0) = 1. \quad (23)$$

It is a unique and real analytic solution for  $\operatorname{Re}(k > 0)$  [Dunkl].

Some properties:

- $E_k(y, x) = E_k(x, y)$ .
- $E_k(gx, gy) = E_k(x, y)$ , for  $g \in G$ .

And the convolution/Fourier theorem [Dunkl]

$$\int_N E_k(x, y) E_k(x, z) e^{-|x|^2/2} w_k(x) dx = c_k e^{(y, y)/2 + (z, z)/2} E_k(z, y).$$



The following two theorems by [Opdam] and [Okounkov] connect the theory of Dunkl operators to matrix models.

**Definition:** **Generalized bessel function** is defined as

$$F_k(x, y) := \frac{1}{|G|} \sum_{f \in G} E_k(g^x, y). \quad (24)$$

**Theorem:** [Okounkov]

$$F_{\frac{1}{\alpha}}(x, y) = \sum_{\lambda} \frac{P_{\lambda}(x; \alpha) P_{\lambda}(y; \alpha)}{(n/\alpha)_{\lambda} \mathfrak{p}_{\lambda}}. \quad (25)$$

where  $(u)_{\lambda} = \prod_{(i,j) \in \lambda} (u + (j-1) - (i-1)/\alpha)$ , and  $\mathfrak{p}_{\lambda} = (P_{\lambda}, P_{\lambda})$  and  $\operatorname{Re}(\alpha) > 0$ .

## Result:

Using Okounkov Bessel function formula and Dunkl convolution theorem we can evaluate the average of two Jack polynomials

**Theorem:**

$$\mathbb{E}[P_\mu(x; \alpha)P_\beta(x; \alpha)] = \alpha^{-|\mu|-|\beta|} J_\mu(1; \alpha) J_\beta(1; \alpha) \times \left\langle P_\mu(y; \alpha), \left\langle P_\beta(z; \alpha), e^{|y|^2/2+|z|^2/2} \sum_\lambda \frac{P_\lambda(y; \alpha)P_\lambda(z; \alpha)}{(n/\alpha)_\lambda p_\lambda} \right\rangle \right\rangle \quad (26)$$

**Corollary:**

$$\mathbb{E}[P_\mu(x; \alpha)e^{-p_1(x)}] = \alpha^{-|\mu|} J_\mu(1; \alpha) e^{\frac{-N}{2}} [p_2^{|\mu|/2} - p_1^{|\mu|}] P_\mu \quad (27)$$

**Thanks.**

# Cauchy identity

Cauchy identity for Jack polynomials is

$$\exp \left[ \beta \sum_{k=1}^{\infty} \frac{p_k \bar{p}_k}{k} \right] = \sum_{\lambda} \frac{P_{\lambda} \{p_k\} P_{\lambda} \{\bar{p}_k\}}{\langle P_{\lambda}, P_{\lambda} \rangle}. \quad (28)$$

here  $p_k$  and  $\bar{p}_k$  are power sum polynomials of possibly different sets if variables. Putting  $\bar{p}_k = a_2 \beta^{-1} \delta_{k,2}$  we see that

$$\exp \left[ \frac{a_2 p_2}{2} \right] = \sum_{\lambda} \frac{P_{\lambda} \cdot P_{\lambda} \{a_2 \beta^{-1} \delta_{k,2}\}}{\langle P_{\lambda}, P_{\lambda} \rangle}. \quad (29)$$

But the left hand side can also obtain by multiplying  $p_2$  n-times on  $P_{\emptyset}$ . This gives us the expression for C's

$$\sum_{\lambda^{(2)} \dots \lambda^{(2n-2)}; \lambda^{(2n)} = \lambda} C_{\phi \lambda^{(2)}} C_{\lambda^{(2)} \lambda^{(4)}} \dots C_{\lambda^{(2n-2)} \lambda^{(2n)}} = \frac{2^{|\lambda|/2} (|\lambda|/2)! P_{\lambda} \{a_2 \beta^{-1} \delta_{k,2}\}}{a_2^{|\lambda|/2} \langle P_{\lambda}, P_{\lambda} \rangle}. \quad (30)$$

## A and Peri rule

First, we notice that

$$A_{\mu\lambda} = \frac{\langle P_\lambda, p_1^2 P_\mu \rangle}{\langle P_\lambda, p_2 P_\mu \rangle} = \frac{1}{1 - 2 \frac{\langle P_\lambda, P_{\{1,1\}} P_\mu \rangle}{\langle P_\lambda, P_1^2 P_\mu \rangle}}. \quad (31)$$

The last equality follows from  $P_{\{1\}} = p_1$  and  $P_{\{1,1\}} = \frac{1}{2} (p_1^2 - p_2)$ . Jack polynomials are satisfying a rule known as Pieri rule

$$P_{(1^r)} P_\mu = \sum_{\lambda} c_{\mu, (1^r)}^\lambda P_\lambda, \quad (32)$$

where  $\lambda - \mu$  is vertical  $r$ -strip and  $c_{\mu, (1^r)}^\lambda$  is known object. Using this rule we can calculate  $A_{\mu\nu}$

$$\beta(1 - \beta)A_{\mu\lambda} = (j_2 - j_1 + \beta(i_1 - i_2))^2 - (1 - \beta + \beta^2) \quad (33)$$

## Peri rule coefficients

The coefficients in Peri rule 32 is given by

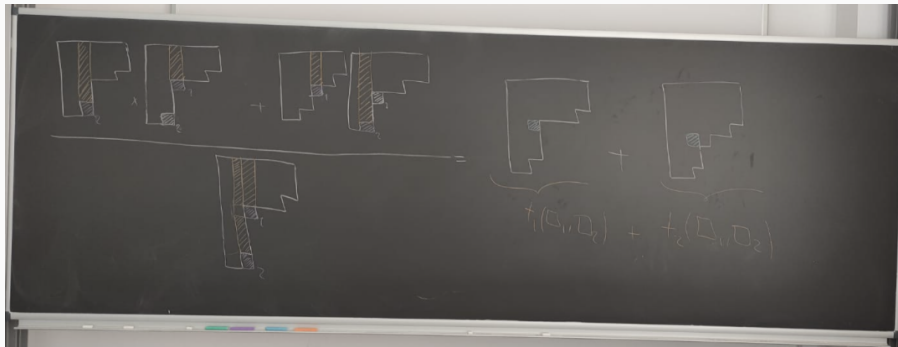
$$c_{\mu,(1^r)}^\lambda = \prod_{s \in X(\lambda/\mu)} \frac{h_*^\lambda(s) h_\mu^*(s)}{h_*^\mu(s) h_\lambda^*(s)} \quad (34)$$

$h_\lambda^*(s)$  and  $h_*^\lambda(s)$  are respectively the upper and lower hook lengths of the box  $s$ :

$$\begin{aligned} h_\lambda^*(s) &= \beta^{-1}(a_\lambda(s) + 1) + l_\lambda(s) \\ h_*^\lambda(s) &= \beta^{-1}a_\lambda(s) + l_\lambda(s) + 1 \end{aligned} \quad (35)$$

$$\frac{\langle P_\lambda, P_1^2 P_\mu \rangle}{\langle P_\lambda, P_{(1,1)} P_\mu \rangle} = \frac{\sum_{\sigma = \mu + \square} c_{\sigma,1}^\lambda c_{\mu,1}^\sigma}{c_{\mu,(1,1)}^\lambda} = \frac{c_{\mu+\square_2,1}^{\mu+\square_1+\square_2} c_{\mu,1}^{\mu+\square_2} + c_{\mu+\square_1,1}^{\mu+\square_1+\square_2} c_{\mu,1}^{\mu+\square_1}}{c_{\mu,(1,1)}^{\mu+\square_1+\square_2}} \quad (36)$$

# Depicted version of Peri rule



## References

---

- [1] L. Lando, A. Zvonkin , Graphs on Surfaces and Their Applications
- [2] L. Cassia, R. Lodin and M. Zabzine, JHEP **10** (2020), 126  
doi:10.1007/JHEP10(2020)126 [arXiv:2007.10354 [hep-th]].
- [3] C. Itzykson and J.-B. Zuber. Matrix integration and combinatorics of modular groups. Communications in Mathematical Physics, 134(1):197 – 207, 1990. doi: cmp/1104201618. URL <https://doi.org/>.



# Jack polynomials

**Jack Polynomials** is beta deform extension of Schur polynomials. We can define Jack polynomial  $P$  similar to Schur polynomial by adding an extra weight.

$$P_\lambda = \sum_T \psi_T(\beta) \prod_{s \in \lambda} z_{T(s)}, \quad (37)$$

where extra weight  $\psi_T$  is given with respect to sequence of partition in Young diagram,  $\emptyset = \nu_1 \rightarrow \nu_2 \rightarrow \dots \rightarrow \nu_n = \lambda$ .

$$\psi_T(\beta) = \prod_i \psi_{\nu_{i+1}/\nu_i} \quad \text{where,} \quad (38)$$

$$\psi_{\lambda/\mu} = \prod_{s \in R_{\lambda/\mu} - C_{\lambda/\mu}} \frac{\text{arm}_\mu(s) + \beta(\text{leg}_\mu(s) + 1)}{\text{arm}_\mu(s) + \beta \text{leg}_\mu + 1} \frac{\text{arm}_\lambda(s) + \beta(\text{leg}_\lambda(s) + 1)}{\text{arm}_\lambda(s) + \beta \text{leg}_\lambda + 1} \quad (39)$$

where  $\text{arm}(s)$  is number of boxes in the right of  $s$  and  $\text{leg}(s)$  is number of boxes below  $s$ . Jack polynomials form an orthogonal basis  $\langle P_\lambda, P_\mu \rangle = 0$  whenever  $\lambda \neq \mu$ .

If we put  $\beta = 1$ , we get  $\psi = 1$ , and we recover definition of Schur.