

Convolutions and fluctuations: free, finite, quantized.

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Hermitian matrix operations

$$A = \begin{pmatrix} a_1 & 0 & & \\ 0 & a_2 & 0 & \\ & 0 & \ddots & 0 \\ & & 0 & a_N \end{pmatrix} \quad B = \begin{pmatrix} b_1 & 0 & & \\ 0 & b_2 & 0 & \\ & 0 & \ddots & 0 \\ & & 0 & b_N \end{pmatrix}$$

U, V – Haar-random in $\text{Unitary}(N; \mathbb{R} / \mathbb{C} / \mathbb{H})$

$$C = UAU^* + VBV^*$$

↖ ↗
uniformly random eigenvectors

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or $C = (UAU^*) \cdot (VBV^*)$

or $C = P_k(UAU^*)P_k$

Question: What can you say about **eigenvalues** of C ?

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Question: What can you say about **eigenvalues** of C ?

- I) As $\beta (= 1, 2, 4) \rightarrow \infty$
- II) As $N \rightarrow \infty$
- III) In discretization

As $\beta (= 1, 2, 4) \rightarrow \infty$

Theorem. (Gorin–Marcus–17) Eigenvalues of C **crystallize**
(=they become deterministic) as $\beta \rightarrow \infty$:

$$\lim_{\beta \rightarrow \infty} C = UAU^* + VBV^*$$

$$\prod_{i=1}^N (z - c_i) = \frac{1}{N!} \sum_{\sigma \in S(N)} \prod_{i=1}^N (z - a_i - b_{\sigma(i)})$$

$$\lim_{\beta \rightarrow \infty} C = (UAU^*) \cdot (VBV^*)$$

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$$\prod_{i=1}^k (z - c_i) \sim \frac{\partial^{N-k}}{\partial z^{N-k}} \prod_{i=1}^N (z - a_i)$$

Finite free convolutions and projection

As $N \rightarrow \infty$

Theorem. (Voiculescu, 80s) At $\beta = 1, 2$ empirical measure of eigenvalues of C becomes deterministic as $N \rightarrow \infty$.

$$\mu_A = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \delta_{a_i} \quad \mu_B = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \delta_{b_i} \quad \mu_C = \lim_{N \rightarrow \infty} \frac{1}{N} \sum_{i=1}^N \delta_{c_i}$$

$$G_\mu(z) = \int \frac{\mu(dx)}{z - x}$$

$$R_\mu(z) = (G_\mu(z))^{-1} - \frac{1}{z}, \quad S_\mu(z) = \frac{z}{1+z} (1 - zG_\mu(z))^{-1}$$

$$\lim_{N \rightarrow \infty} C = UAU^* + VBV^*$$

$$R_{\mu_C}(z) = R_{\mu_A}(z) + R_{\mu_B}(z)$$

$$\lim_{N \rightarrow \infty} C = (UAU^*) \cdot (VBV^*)$$

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$$R_{\mu_C}(z) = \frac{N}{k} R_{\mu_A}(z)$$

Free convolutions and projection

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Free convolutions and projection

Conjecture. Same is true for any fixed $\beta > 0$.

Discretization

T_λ irreducible (linear) representations of $U(N; \mathbb{C})$

$$\lambda_1 > \lambda_2 > \cdots > \lambda_N, \quad \lambda_i \in \mathbb{Z}.$$

$$T_\lambda \otimes T_\nu = \bigoplus_{\kappa} c_{\lambda, \nu}^\kappa T_\kappa$$

Littlewood–Richardson coefficients $c_{\lambda, \nu}^\kappa$ intractable as $N \rightarrow \infty$.

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Random κ through $P(\kappa) = \frac{\dim(T_\kappa) c_{\lambda,\nu}^\kappa}{\dim T_\lambda \cdot \dim T_\nu}$.

Semi-classical limit degenerates representations of a Lie group into orbital measures on its Lie algebra

$$T_\lambda \otimes T_\nu \longrightarrow UAU^* + VBV^*$$

Discretization

Theorem. (Gorin–Bufetov–13; following Biane in 90s)

Empirical measure of ν becomes deterministic as $N \rightarrow \infty$.

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Scaling is important!

$$G_\mu(z) = \int \frac{\mu(dx)}{z - x}, \quad R_\mu^{\text{quant}}(z) = \left(G_\mu(z)\right)^{-1} - \frac{1}{1 - \exp(-z)}$$

$$R_{\mu_\kappa}^{\text{quant}}(z) = R_{\mu_\lambda}^{\text{quant}}(z) + R_{\mu_\nu}^{\text{quant}}(z)$$

Quantized free convolution

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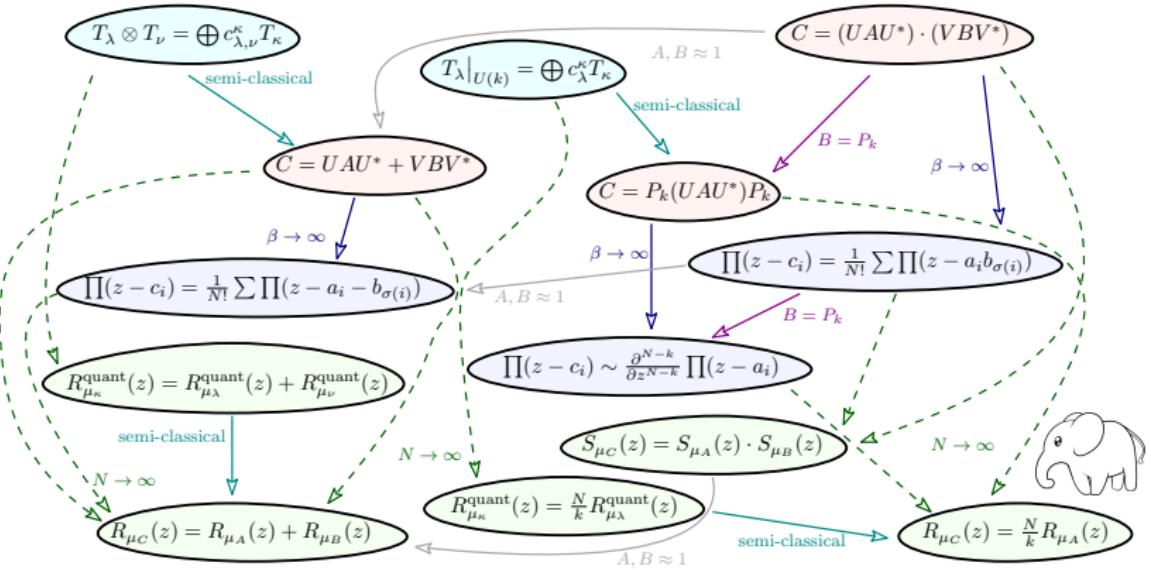
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Quantized free convolution

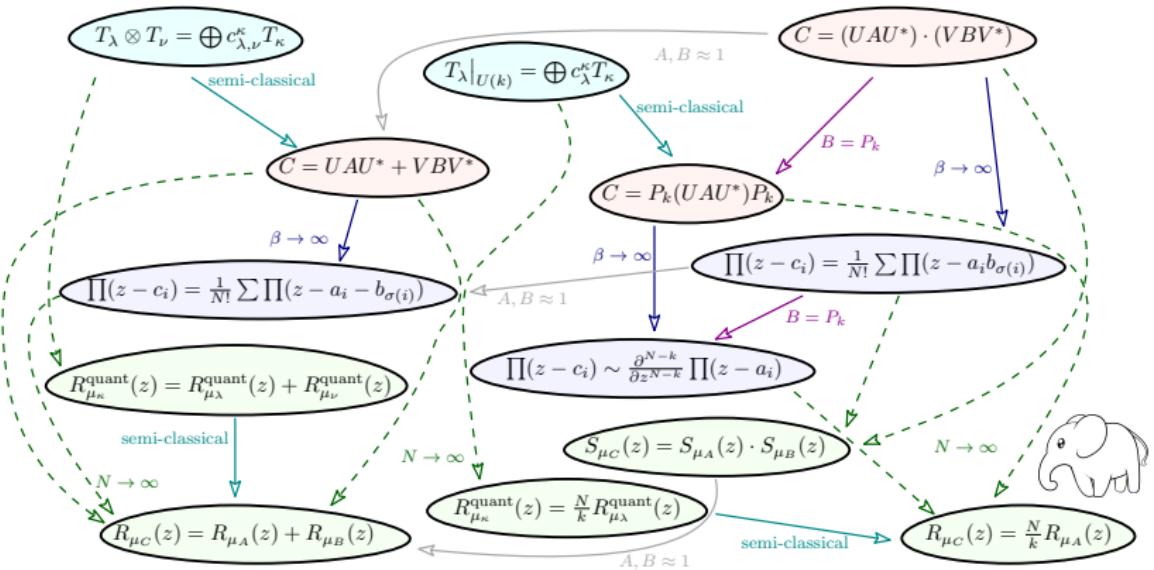
Restrictions to smaller subgroups lead to projection.

Zoo of operations



Operations on matrices and representations lead to a variety of **Laws of Large Numbers**, resulting in **convolutions**. They are all cross-related by limit transitions.

Zoo of operations



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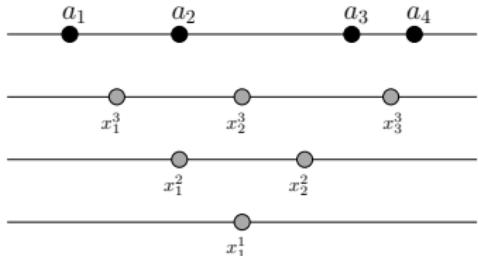
We present a **unifying framework** for the operations.

Matrix corners

The most explicit case.

$N \times N$ matrix UAU^*

$$\left(\begin{array}{c|cc|cc} M_{11} & M_{12} & M_{13} & M_{14} \\ \hline M_{21} & M_{22} & M_{23} & M_{24} \\ \hline M_{31} & M_{32} & M_{33} & M_{34} \\ \hline M_{41} & M_{42} & M_{43} & M_{44} \end{array} \right)$$



Theorem. (Gelfand–Naimark–50s; Baryshnikov, Neretin – 00s)
With $(x_1^N, \dots, x_N^N) = (a_1, \dots, a_N)$, the joint law of particles is

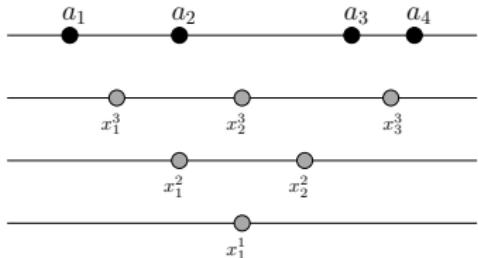
$$\prod_{k=1}^{N-1} \prod_{1 \leq i < j \leq k} (x_i^k - x_j^k)^{2-\beta} \prod_{a=1}^k \prod_{b=1}^{k+1} |x_a^k - x_b^{k+1}|^{\beta/2-1}$$

- A basis of extension from $\beta = 1, 2, 4$ to **general** $\beta > 0$.
- Consistent with (Hermite/Laguerre/Jacobi) β log-gases

Matrix corners

$N \times N$ matrix UAU^*

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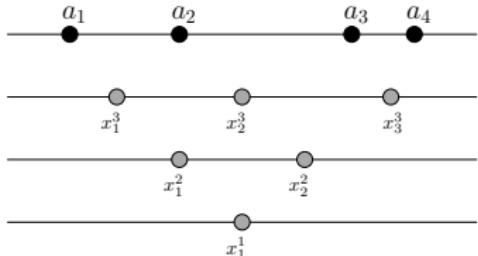
Multivariate Bessel Function

$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) = \mathbb{E} \exp \left(\sum_{k=1}^N z_k \cdot \left(\sum_{i=1}^k x_i^k - \sum_{j=1}^{k-1} x_j^{k-1} \right) \right)$$

Matrix corners

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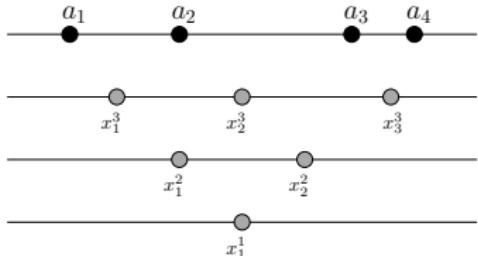
Diagonal matrix elements at $\beta = 1, 2, 4$

Proposition. $\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N)$ is **symmetric** in z_1, \dots, z_N .

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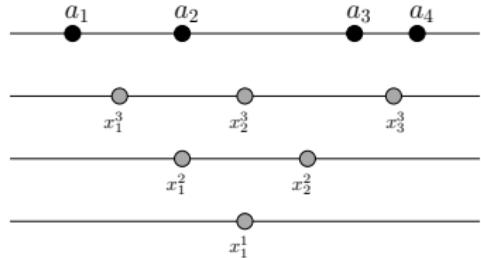
Laplace transform for full matrix at $\beta = 1, 2, 4$.

$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) = \mathbb{E} \exp [\text{Trace}(UAU^* Z)], \quad Z \sim \{z_i\} \text{ eigenvalues}$$

Matrix corners

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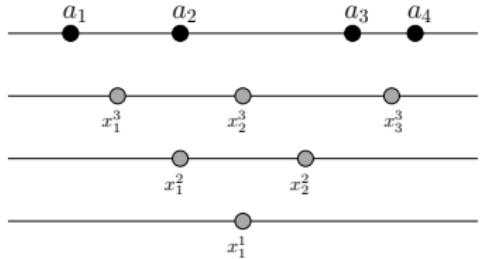
(Harish Chandra; Itzykson–Zuber) At $\beta = 2$:

$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) \sim \frac{\det [\exp(a_i z_j)]_{i,j=1}^N}{\prod_{i < j} (a_i - a_j)(z_i - z_j)}$$

Matrix corners

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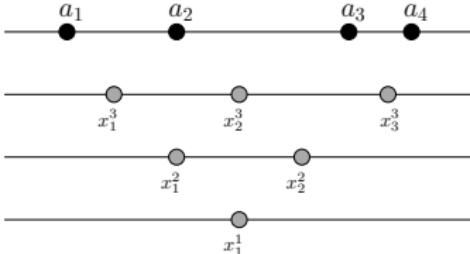
$\left[\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) \cdot \prod_{i < j} (z_i - z_j)^{\beta/2} \right]$ is an eigenfunction of
rational **Calogero–Sutherland Hamiltonian** for any $\beta > 0$

$$\sum_{i=1}^N \frac{\partial^2}{\partial z_i^2} + \frac{\beta(2-\beta)}{2} \sum_{i < j} \frac{1}{(z_i - z_j)^2}$$

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The probability law of interest is readily **reconstructed**:

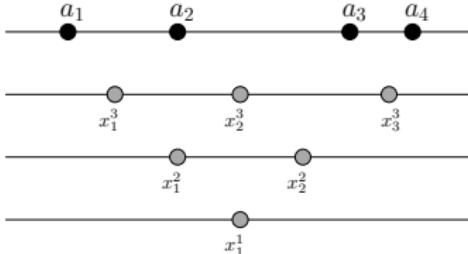
$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_k, 0, \dots, 0)$$

$$= \iiint \mathcal{B}_{x_1^k, x_2^k, \dots, x_k^k}(z_1, \dots, z_k) \mathbb{P}(dx_1^k, dx_2^k, \dots, dx_k^k)$$

Matrix corners

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The probability law of interest is readily **reconstructed**:

$$\begin{aligned} \mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_k, 0, \dots, 0) \\ = \iiint \mathcal{B}_{x_1^k, x_2^k, \dots, x_k^k}(z_1, \dots, z_k) \mathbb{P}(dx_1^k, dx_2^k, \dots, dx_k^k) \end{aligned}$$

Equivalently, we have a family of **observables**

$$\mathbb{E} \mathcal{B}_{x_1^k, x_2^k, \dots, x_k^k}(z_1, \dots, z_k) = \mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_k, 0, \dots, 0)$$

Matrix addition

$N \times N$ matrix UAU^*

$N \times N$ matrix VBV^*

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$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) = \mathbb{E} \exp \left(\sum_{k=1}^N z_k \cdot \left(\sum_{i=1}^k x_i^k - \sum_{j=1}^{k-1} x_j^{k-1} \right) \right)$$

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$$C = UAU^* + VBV^*$$

Question: $\mathbb{E} \mathcal{B}_{c_1, \dots, c_N}(z_1, \dots, z_N) = ?$

Matrix addition

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$$\mathcal{B}_{b_1, \dots, b_N}(z_1, \dots, z_N) = \mathbb{E} \exp \left(\sum_{k=1}^N z_k \cdot \left(\sum_{i=1}^k y_i^k - \sum_{j=1}^{k-1} y_j^{k-1} \right) \right)$$

$C = UAU^* + VBV^*$

$\mathbb{E} \mathcal{B}_{c_1, \dots, c_N}(z_1, \dots, z_N) = \mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) \cdot \mathcal{B}_{b_1, \dots, b_N}(z_1, \dots, z_N)$

Matrix addition

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$$\mathcal{B}_{a_1, \dots, a_N}(z_1, \dots, z_N) = \mathbb{E} \exp \left(\sum_{k=1}^N z_k \cdot \left(\sum_{i=1}^k x_i^k - \sum_{j=1}^{k-1} x_j^{k-1} \right) \right)$$

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Any $\beta > 0!$ **Caveat:** positivity in \mathbb{E} is **open** outside $\beta = 1, 2, 4$.

Operations through symmetric functions

$C = UAU^* + VBV^*$	Multivariate Bessel $\mathbb{E}\mathcal{B}_C(\vec{z}) = \mathcal{B}_A(\vec{z}) \cdot \mathcal{B}_B(\vec{z})$
$C = P_k(UAU^*)P_k$	Multivariate Bessel $\mathbb{E}\mathcal{B}_C(\vec{z}) = \mathcal{B}_A(\vec{z}, 0^{N-k})$
$C = (UAU^*) \cdot (VBV^*)$	Heckman–Opdam hypergeometric $\mathbb{E}\mathcal{HO}_C(\vec{z}) = \mathcal{HO}_A(\vec{z}) \cdot \mathcal{HO}_B(\vec{z})$
$T_\lambda \otimes T_\nu = \bigoplus_{\kappa} c_{\lambda,\nu}^\kappa T_\kappa$	Schur polynomials $\mathbb{E} \left[\frac{s_\kappa(\vec{z})}{s_\kappa(1^N)} \right] = \frac{s_\lambda(\vec{z})}{s_\lambda(1^N)} \cdot \frac{s_\nu(\vec{z})}{s_\nu(1^N)}$
$T_\lambda _{U(k)} = \bigoplus_{\kappa} c_\lambda^\kappa T_\kappa$	Schur polynomials $\mathbb{E} \left[\frac{s_\kappa(\vec{z})}{s_\kappa(1^k)} \right] = \frac{s_\lambda(\vec{z}, 1^{N-k})}{s_\lambda(1^N)}$

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These all are degenerations of **Macdonald polynomials**.

Macdonald polynomials

$P_\lambda(x_1, \dots, x_N; q, t)$: homogenous, symmetric, leading term $\prod_{i=1}^N x_i^{\lambda_i}$

$$[T_{i;q} f](x_1, \dots, x_N) = f(x_1, \dots, x_{i-1}, \textcolor{blue}{q}x_i, x_{i+1}, \dots, x_N)$$

$$\mathcal{D} = \sum_{i=1}^N \prod_{j \neq i} \frac{tx_i - x_j}{x_i - x_j} T_{i;q}$$

$$\mathcal{D}P_\lambda(x_1, \dots, x_N; q, t) = \left[\sum_{i=1}^N q^{\lambda_i} t^{N-i} \right] P_\lambda(x_1, \dots, x_N; q, t)$$

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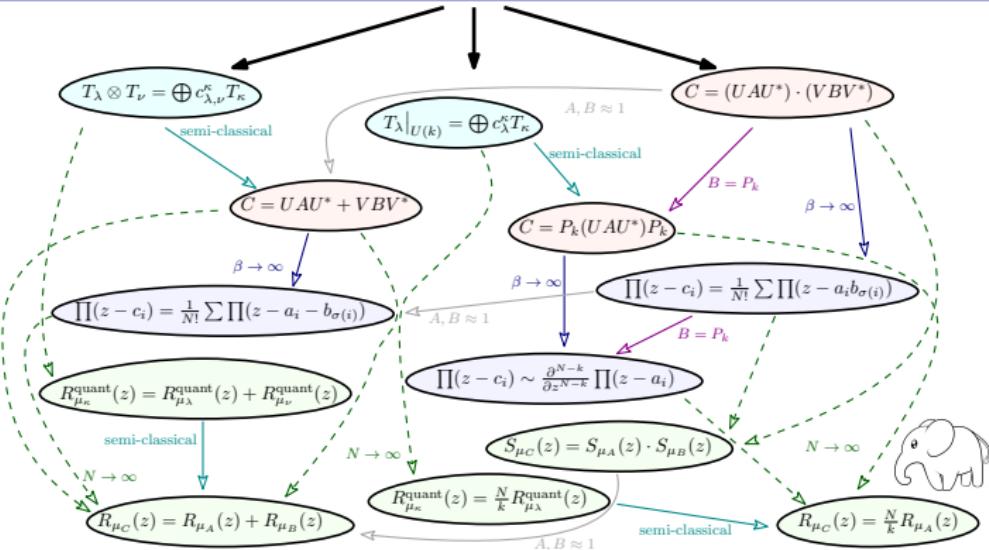
$$\frac{P_\lambda(x_1, \dots, x_N)}{P_\lambda(1, t, \dots, t^{N-1})} \frac{P_\nu(x_1, \dots, x_N)}{P_\nu(1, t, \dots, t^{N-1})} = \sum_{\kappa} c_{\lambda, \nu}^\kappa \frac{P_\kappa(x_1, \dots, x_N)}{P_\kappa(1, t, \dots, t^{N-1})}$$

Conjecture. $c_{\lambda, \nu}^\kappa \geq 0$ whenever $0 < q, t < 1$.

Known cases: $q = t$; $q = 0$; $t = 0$; $q \rightarrow 1$, $t = q^{\beta/2}$, $\beta = 1, 2, 4$.

Macdonald polynomials

$$\frac{P_\lambda(x_1, \dots, x_N)}{P_\lambda(1, t, \dots, t^{N-1})} \frac{P_\nu(x_1, \dots, x_N)}{P_\nu(1, t, \dots, t^{N-1})} = \sum_{\kappa} c_{\lambda, \nu}^{\kappa} \frac{P_{\kappa}(x_1, \dots, x_N)}{P_{\kappa}(1, t, \dots, t^{N-1})}$$



$\sum_{\kappa} c_{\lambda, \nu}^{\kappa} = 1$ and they define a distribution on κ 's.
Everything we saw so far is a limit of this distribution.

Macdonald polynomials

$$\frac{P_\lambda(x_1, \dots, x_N)}{P_\lambda(1, t, \dots, t^{N-1})} \frac{P_\nu(x_1, \dots, x_N)}{P_\nu(1, t, \dots, t^{N-1})} = \sum_{\kappa} c_{\lambda, \nu}^{\kappa} \frac{P_\kappa(x_1, \dots, x_N)}{P_\kappa(1, t, \dots, t^{N-1})}$$

Our analysis of $c_{\lambda, \nu}^{\kappa}$ -random κ :

1. Direct combinatorial limits of Macdonald polynomials.
2. Applying difference/differential operators in x_1, \dots, x_N to the defining identity.
3. Contour integral expressions for $\frac{P_\lambda(x_1, 1, t, \dots, t^{N-2})}{P_\lambda(1, t, \dots, t^{N-1})}$.

Then $N \rightarrow \infty$ analysis of the most general Macdonald case is **open**.

Beyond the Law of Large Numbers

$$\frac{P_\lambda(x_1, \dots, x_N)}{P_\lambda(1, t, \dots, t^{N-1})} \frac{P_\nu(x_1, \dots, x_N)}{P_\nu(1, t, \dots, t^{N-1})} = \sum_{\kappa} c_{\lambda, \nu}^{\kappa} \frac{P_\kappa(x_1, \dots, x_N)}{P_\kappa(1, t, \dots, t^{N-1})}$$

Information is lost by considering deterministic limits of **random** κ .

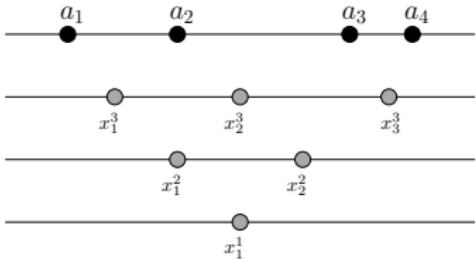
What can be recovered?

Matrix corners again

Back to the most explicit case.

$N \times N$ matrix UAU^*

$$\left(\begin{array}{c|cc|cc} M_{11} & M_{12} & M_{13} & M_{14} \\ \hline M_{21} & M_{22} & M_{23} & M_{24} \\ \hline M_{31} & M_{32} & M_{33} & M_{34} \\ \hline M_{41} & M_{42} & M_{43} & M_{44} \end{array} \right)$$



With $(x_1^N, \dots, x_N^N) = (a_1, \dots, a_N)$, the joint law of particles is

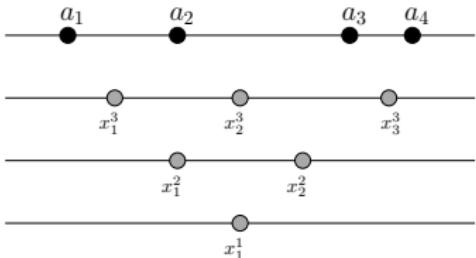
$$\prod_{k=1}^{N-1} \prod_{1 \leq i < j \leq k} (x_i^k - x_j^k)^{2-\beta} \prod_{a=1}^k \prod_{b=1}^{k+1} |x_a^k - x_b^{k+1}|^{\beta/2-1}$$

What is happening as $\beta \rightarrow \infty$?

Matrix corners again

$N \times N$ matrix UAU^*

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As $\beta \rightarrow \infty$, particles maximize

$$\prod_{k=1}^{N-1} \prod_{1 \leq i < j \leq k} (x_i^k - x_j^k)^{-2} \prod_{a=1}^k \prod_{b=1}^{k+1} |x_a^k - x_b^{k+1}| \rightarrow \max$$

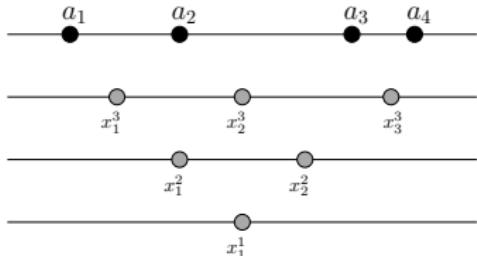
Matrix corners again

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Proposition. The optimal configuration is given by

$$\prod_{i=1}^k (z - \bar{x}_i^k) \sim \frac{\partial^{N-k}}{\partial z^{N-k}} \prod_{j=1}^N (z - a_j), \quad k = 1, 2, \dots, N.$$

Matrix corners again

With $(x_1^N, \dots, x_N^N) = (a_1, \dots, a_N)$, the joint law of particles is

$$\prod_{k=1}^{N-1} \prod_{1 \leq i < j \leq k} (x_i^k - x_j^k)^{2-\beta} \prod_{a=1}^k \prod_{b=1}^{k+1} |x_a^k - x_b^{k+1}|^{\beta/2-1}$$

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Proposition. $\xi_j^i := \lim_{\beta \rightarrow \infty} \sqrt{\beta}(x_j^i - \bar{x}_j^i)$ has Gaussian density

$$\exp \left(\sum_{k=1}^{N-1} \left[\sum_{1 \leq i < j \leq k} \frac{(\xi_i^k - \xi_j^k)^2}{2(x_i^k - x_j^k)^2} - \sum_{a=1}^k \sum_{b=1}^{k+1} \frac{(\xi_a^k - \xi_b^{k+1})^2}{4(x_a^k - x_b^{k+1})^2} \right] \right).$$

A version of **discrete Gaussian Free Field**.

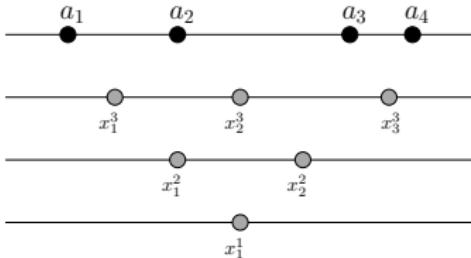
Central Limit Theorems for fluctuations

1. $\lim_{\beta \rightarrow \infty} [C = P_k(UAU^*)P_k]$. (Gorin–Marcus–17): **discrete GFF**
2. $\lim_{\beta \rightarrow \infty} [C = (UAU^*)(+ \text{ or } \cdot)(VBV^*)]$. **Open problem**
3. $\lim_{N \rightarrow \infty} \left[T_\lambda \Big|_{U(k)} = \bigoplus_{\kappa} c_{\lambda}^{\kappa} T_{\kappa} \right]$. (Petrov–12, Bufetov–Gorin–15):
Fluctuations are given by **Gaussian Free Field** (= covariance given by the Green function of the Laplace operator in suitable complex structure); bijection with random lozenge tilings.
4. $\lim_{N \rightarrow \infty} [C = P_k(UAU^*)P_k]$. Same methods and results apply.
5. $\lim_{N \rightarrow \infty} [C = UAU^* + VBV^*]$. (Collins–Mingo–Sniady–Speicher–04)
Second order freeness. Covariance similar to GFF [\(??\)](#).
6. $\lim_{N \rightarrow \infty} [C = (UAU^*) \cdot (VBV^*)]$. (Vasilchuk–16) Similar [\(??\)](#).
7. **Conjecture:** The last 3 answers extend to general $\beta > 0$.
8. $\lim_{N \rightarrow \infty} \left[T_\lambda \otimes T_\nu = \bigoplus_{\kappa} c_{\lambda,\nu}^{\kappa} T_{\kappa} \right]$. (Bufetov–Gorin–15) CLT with covariance similar to GFF. **Conceptual explanation?**

The simplest open problem

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Fix **arbitrary** $\beta > 0$ and send $N \rightarrow \infty$

Conjecture. The Law of Large Numbers is β -independent.

Conjecture. The Central Limit Theorem as $N \rightarrow \infty$ is β -independent after rescaling by $\sqrt{\beta}$.

Difficulty: Negative powers in the interaction.

Summary

- Operations on matrices and representations possess Laws of Large Numbers as $N \rightarrow \infty$ or $\beta \rightarrow \infty$ leading to various **convolutions**.
- Fluctuations are described by Gaussian fields. Whenever formula-less identification is made, it is **GFF**.
- Macdonald polynomials multiplication** is the mother of all.
- Many **open** questions remain!

