

Coloured vertex models and non-symmetric Macdonald polynomials

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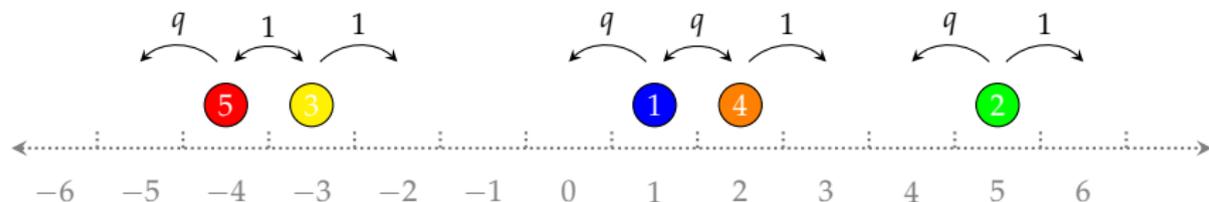
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Outline

- ▶ Motivational examples (**multi-species ASEP** and **coloured bosons**).
- ▶ 1: Non-symmetric rational functions and their properties.
 - ▶ Fundamental $U_q(\widehat{\mathfrak{sl}}_{n+1})$ vertex model.
 - ▶ Non-symmetric rational functions f_μ and g_μ .
 - ▶ Transfer matrix $G_{\mu/\nu}$.
 - ▶ Summation identities involving f_μ , g_μ and $G_{\mu/\nu}$.
 - ▶ Orthogonality relations.
 - ▶ An integral formula for $G_{\mu/\nu}$ and the **ASEP limit**.
- ▶ 2: Non-symmetric Hall–Littlewood polynomials and distribution matches.
 - ▶ Fusion of the $U_q(\widehat{\mathfrak{sl}}_{n+1})$ vertex model.
 - ▶ Non-symmetric Hall–Littlewood polynomials, E_μ .
 - ▶ Hall–Littlewood processes.
 - ▶ A remarkable distribution match.
- ▶ 3: Non-symmetric Macdonald polynomials; **no time today!**
 - ▶ Cherednik–Dunkl operators and their eigenaction.
 - ▶ A deformation of the partition function for non-symmetric Hall–Littlewood polynomials.
 - ▶ Lattice model construction of non-symmetric Macdonald polynomials.

Motivation: the multi-species ASEP



- ▶ A configuration of the multi-species ASEP is given by

$$\eta(t) = \{\dots, \eta_{-1}(t), \eta_0(t), \eta_1(t), \dots\}, \quad \eta_i(t) \in \mathbb{N}.$$

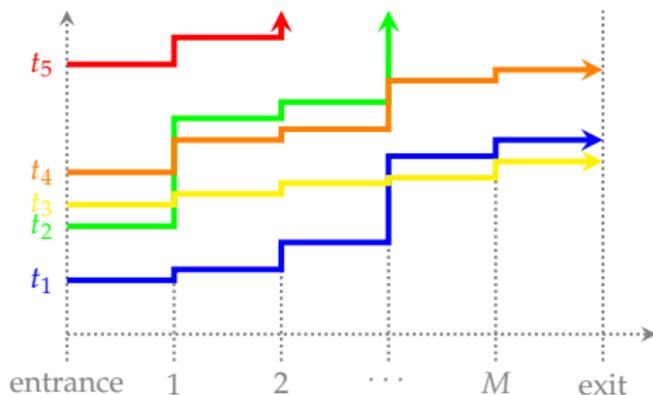
- ▶ 0 indicates an unoccupied site, and an integer $j \geq 1$ indicates a particle of colour j .
- ▶ Let $\{\eta_{i-1}, \eta_i, \eta_{i+1}\}$ be the occupation data for the site i and its two neighbours, at some point in time. This site is assigned two exponential clocks: a left clock of rate $q \cdot \mathbf{1}_{\eta_i > \eta_{i-1}}$ and a right clock of rate $\mathbf{1}_{\eta_i > \eta_{i+1}}$.
- ▶ When the left clock rings, the occupation data is updated as

$$\{\eta_{i-1}, \eta_i, \eta_{i+1}\} \mapsto \{\eta_i, \eta_{i-1}, \eta_{i+1}\}.$$

- ▶ When the right clock rings, the occupation data is updated as

$$\{\eta_{i-1}, \eta_i, \eta_{i+1}\} \mapsto \{\eta_{i-1}, \eta_{i+1}, \eta_i\}.$$

Motivation: coloured bosons



- ▶ Particles arrive at site 1 of the lattice according to a Poisson process of rate r_0 . The i -th particle to enter the system is assigned colour $t_i \in \mathbb{R}_{>0}$, where t_i is its time of entrance.
- ▶ There is an exponential clock of rate $r_i \in \mathbb{R}_{>0}$ attached to the i -th site of the lattice, for each $1 \leq i \leq M$.
- ▶ When the i -th clock rings, $1 \leq i \leq M - 1$, the particle with maximal colour at position i hops to position $i + 1$.
- ▶ When the M -th clock rings, the particle with maximal colour at position M exits the system.

1: Fundamental $U_q(\widehat{\mathfrak{sl}}_{n+1})$ vertex model

- ▶ The model has a **sum-to-unity** property:

$$\sum_{0 \leq k, \ell \leq n} j \begin{array}{c} \uparrow k \\ \leftarrow \ell \\ \rightarrow \ell \\ \downarrow i \end{array} = 1, \quad \forall i, j \in \{0, 1, \dots, n\}.$$

With mild assumptions on the parameters z and q (that ensure positivity), this allows us to construct discrete-time Markov processes from the model.

- ▶ The model satisfies the **Yang–Baxter equation**:

$$\sum_{0 \leq k_1, k_2, k_3 \leq n} \begin{array}{c} \textcircled{x} \\ \textcircled{y} \end{array} \begin{array}{c} i_1 \quad k_2 \\ \diagdown \quad \diagup \\ i_2 \quad k_1 \\ \diagup \quad \diagdown \\ i_3 \end{array} \begin{array}{c} j_3 \\ \uparrow \\ j_2 \\ \rightarrow \\ j_1 \\ \rightarrow \\ \downarrow \\ \textcircled{z} \end{array} = \sum_{0 \leq k_1, k_2, k_3 \leq n} \begin{array}{c} \textcircled{x} \\ \textcircled{y} \end{array} \begin{array}{c} i_1 \quad k_1 \\ \rightarrow \quad \diagdown \\ i_2 \quad k_2 \\ \rightarrow \quad \diagup \\ i_3 \end{array} \begin{array}{c} j_3 \\ \uparrow \\ j_2 \\ \rightarrow \\ j_1 \\ \rightarrow \\ \downarrow \\ \textcircled{z} \end{array}$$

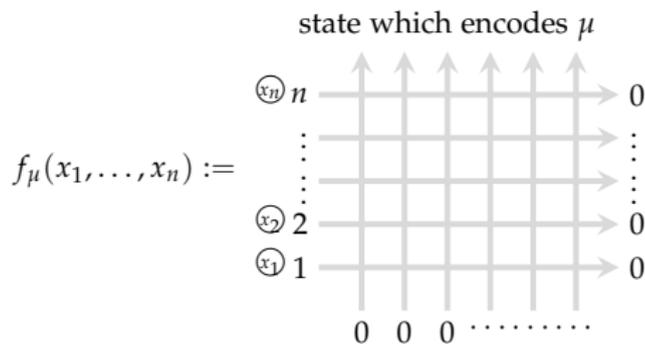
for all fixed $i_1, \dots, i_3, j_1, \dots, j_3 \in \{0, 1, \dots, n\}$.

1: Non-symmetric rational functions f_μ and g_μ

- ▶ In what follows, all vertical lines carry a trivial rapidity variable (= 1):

$ \begin{array}{c} i \\ \uparrow \\ i \rightarrow i \\ \downarrow \\ i \\ 1 \end{array} $	$ \begin{array}{c} j \\ \uparrow \\ i \rightarrow i \\ \downarrow \\ j \\ \frac{q(x-1)}{x-q} \end{array} $	$ \begin{array}{c} i \\ \uparrow \\ i \rightarrow j \\ \downarrow \\ j \\ \frac{(1-q)x}{x-q} \end{array} $
$0 \leq i < j \leq n$	$ \begin{array}{c} i \\ \uparrow \\ j \rightarrow j \\ \downarrow \\ i \\ \frac{x-1}{x-q} \end{array} $	$ \begin{array}{c} j \\ \uparrow \\ j \rightarrow i \\ \downarrow \\ i \\ \frac{(1-q)}{x-q} \end{array} $

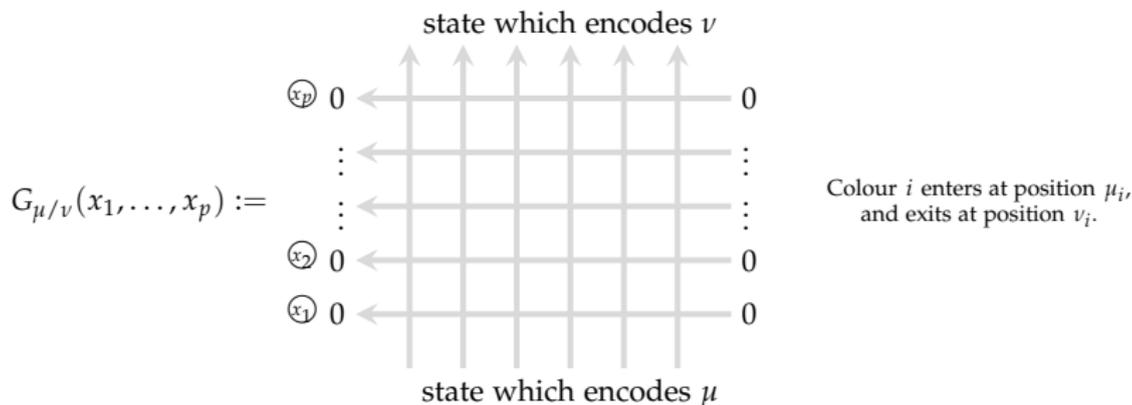
- ▶ Let $\mu = (\mu_1, \dots, \mu_n)$ be a vector of pairwise-distinct natural numbers (a **strict composition**). We define f_μ as a partition function in our vertex model:



Colour i exits at position μ_i .

1: Transfer matrix $G_{\mu/\nu}$

- ▶ We will define one more type of partition function, which plays the role of a **multivariate transfer matrix**.
- ▶ Let $\mu = (\mu_1, \dots, \mu_n)$ and $\nu = (\nu_1, \dots, \nu_n)$ be two strict, natural-number compositions, with $\mu_i \geq \nu_i$ for all $i \in \{1, \dots, n\}$. We define $G_{\mu/\nu}$ as follows:



- ▶ The value of p should be sufficiently large, but is otherwise unrelated to n .

1: Summation identities

- ▶ Let (x_1, \dots, x_n) and (y_1, \dots, y_p) be two alphabets which satisfy the technical constraint

$$\left| \frac{x_i - 1}{x_i - q} \cdot q \cdot \frac{1 - y_j}{1 - qy_j} \right| < 1, \quad \forall i \in \{1, \dots, n\}, \quad j \in \{1, \dots, p\}.$$

- ▶ We have the following **skew Cauchy summation identity**:

$$\sum_{\mu} f_{\mu}(x_1, \dots, x_n) G_{\mu/\nu}(y_1, \dots, y_p) = \prod_{i=1}^n \prod_{j=1}^p \frac{x_i - y_j}{x_i - qy_j} \cdot f_{\nu}(x_1, \dots, x_n),$$

where the summation is taken over all length- n , strict compositions $\mu = (\mu_1, \dots, \mu_n)$.

- ▶ The proof makes use of the Yang–Baxter equation of the model.

1: Summation identities

- ▶ Introduce a function g_μ^* which differs from g_μ by an overall multiplicative factor:

$$g_\mu^*(x_1, \dots, x_n) := \frac{q^{n(n+1)/2}}{(q-1)^n} \cdot g_\mu(x_1, \dots, x_n).$$

- ▶ Let (x_1, \dots, x_n) and (y_1, \dots, y_n) be two alphabets which satisfy the technical constraint

$$\left| \frac{x_i - 1}{x_i - q} \cdot q \cdot \frac{1 - y_j}{1 - qy_j} \right| < 1, \quad \forall i, j \in \{1, \dots, n\}.$$

- ▶ We have the following summation identity of **Mimachi–Noumi type**:

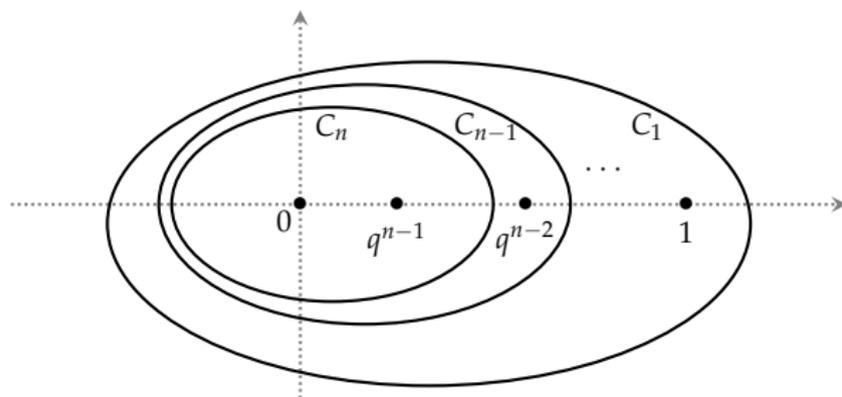
$$\sum_{\mu} f_{\mu}(x_1, \dots, x_n) g_{\mu}^*(y_1, \dots, y_n) = \prod_{i=1}^n \frac{qy_i}{qy_i - x_i} \prod_{n \geq i > j \geq 1} \frac{x_i - y_j}{x_i - qy_j},$$

where the summation is over all length- n , strict compositions $\mu = (\mu_1, \dots, \mu_n)$.

- ▶ The proof makes use of the Yang–Baxter equation of the model.

1: Orthogonality

- ▶ Consider n positively-oriented, q -nested integration contours as shown below:



- ▶ The rational functions f_μ and g_ν^* satisfy the following **orthogonality relation**:

$$\left(\frac{1}{2\pi\sqrt{-1}}\right)^n \oint_{C_1} \frac{dx_1}{x_1} \cdots \oint_{C_n} \frac{dx_n}{x_n} \prod_{1 \leq i < j \leq n} \left(\frac{x_i - x_j}{x_i - qx_j}\right) f_\mu(qx_1, \dots, qx_n) g_\nu^*(x_1, \dots, x_n) = \mathbf{1}_{\mu=\nu}.$$

- ▶ The proof is long and complicated!

1: Spectral decomposition of $G_{\mu/\nu}$

- ▶ It is now very easy to combine the **skew Cauchy identity** and the **orthogonality relation** to obtain the following formula for our **transfer matrix**:

$$G_{\mu/\nu}(y_1, \dots, y_p) = \frac{q^{-np}}{(2\pi\sqrt{-1})^n} \oint_{C_1} \frac{dx_1}{x_1} \dots \oint_{C_n} \frac{dx_n}{x_n} \prod_{1 \leq i < j \leq n} \left(\frac{x_i - x_j}{x_i - qx_j} \right) \prod_{i=1}^n \prod_{j=1}^p \frac{qx_i - y_j}{x_i - y_j} f_\nu(qx_1, \dots, qx_n) g_\mu^*(x_1, \dots, x_n),$$

with the same integration contours as previously.

- ▶ So far all rational functions were indexed by natural-number compositions. It is easy to extend to **integer compositions** by noting a simple shift property of the functions:

For all $\mu \in \mathbb{N}^n$ and $k > 0$ there holds

$$f_{\mu+k^n}(x_1, \dots, x_n) = \prod_{i=1}^n \left(\frac{x_i - 1}{x_i - q} \right)^k f_\mu(x_1, \dots, x_n),$$
$$g_{\mu+k^n}^*(x_1, \dots, x_n) = \prod_{i=1}^n \left(q \cdot \frac{1 - x_i}{1 - qx_i} \right)^k g_\mu^*(x_1, \dots, x_n).$$

1: Spectral decomposition of $G_{\mu/\nu}$

- ▶ For strict compositions $\mu \in \mathbb{Z}^n$, we then define

$$f_{\mu}(x_1, \dots, x_n) := \prod_{i=1}^n \left(\frac{x_i - q}{x_i - 1} \right)^k f_{\mu+k^n}(x_1, \dots, x_n),$$

$$g_{\mu}^*(x_1, \dots, x_n) := \prod_{i=1}^n \left(q^{-1} \cdot \frac{1 - qx_i}{1 - x_i} \right)^k g_{\mu+k^n}^*(x_1, \dots, x_n),$$

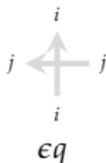
where k is sufficiently large such that $\mu + k^n \in \mathbb{N}^n$.

- ▶ The stated formula for $G_{\mu/\nu}$ continues to hold for arbitrary strict compositions $\mu, \nu \in \mathbb{Z}^n$:

$$G_{\mu/\nu}(y_1, \dots, y_p) = \frac{q^{-np}}{(2\pi\sqrt{-1})^n} \oint_{C_1} \frac{dx_1}{x_1} \dots \oint_{C_n} \frac{dx_n}{x_n} \\ \prod_{1 \leq i < j \leq n} \left(\frac{x_i - x_j}{x_i - qx_j} \right) \prod_{i=1}^n \prod_{j=1}^p \frac{qx_i - y_j}{x_i - y_j} f_{\nu}(qx_1, \dots, qx_n) g_{\mu}^*(x_1, \dots, x_n).$$

1: Reduction to ASEP

- Let us study the 90°-rotated vertex weights in the case when all rapidity variables take the form $x = 1 - \epsilon(1 - q)$, where $\epsilon > 0$ is small. The weights become

 <p style="text-align: center;">1</p>	 <p style="text-align: center;">ϵq</p>	 <p style="text-align: center;">$1 - \epsilon q$</p>
$0 \leq i < j \leq n$	 <p style="text-align: center;">ϵ</p>	 <p style="text-align: center;">$1 - \epsilon$</p>

where we only show $O(\epsilon)$ dependence.

- Fix a parameter $t \in \mathbb{R}_{>0}$ and two strict compositions $\mu, \nu \in \mathbb{Z}^n$. We find that

$$\lim_{\epsilon \rightarrow 0} G_{\mu/(v-p^n)}(y_1, \dots, y_p) \Big|_{p \rightarrow t/\epsilon} \Big|_{y_i \rightarrow 1 - \epsilon(1-q)} = \mathbb{P}_t^{\text{ASEP}}(\mu \rightarrow \nu),$$

where the right hand side is the probability of being in state ν at time t , given that the system was initialized in state μ .

1: Reduction to ASEP

- ▶ Performing this limit in the explicit formula for $G_{\mu/(v-p^n)}$, we obtain all **ASEP transition probabilities**:

$$\mathbb{P}_i^{\text{ASEP}}(\mu \rightarrow \nu) = \frac{1}{(2\pi\sqrt{-1})^n} \oint_{C_1} \frac{dx_1}{x_1} \cdots \oint_{C_n} \frac{dx_n}{x_n} \prod_{1 \leq i < j \leq n} \left(\frac{x_i - x_j}{x_i - qx_j} \right) \prod_{i=1}^n \exp \left[\frac{(1-q)^2 x_i t}{(1-x_i)(1-qx_i)} \right] f_\nu(qx_1, \dots, qx_n) g_\mu^*(x_1, \dots, x_n).$$

- ▶ Explicit formulae for f_ν and g_μ^* (as sums over the symmetric group) exist, but are rather complicated.

If we assume that $\mu_1 > \cdots > \mu_n$ and $\nu_1 < \cdots < \nu_n$ (so that the particles have totally reversed their order after time t), then both f_ν and g_μ^* **factorize**.

2: Fusion and higher-spin vertex models

- ▶ The vertex weights that one obtains in this way depend on M only via q^M . One may **analytically continue** q^M to \mathbb{C} , writing $q^M = s^{-2}$.
- ▶ We then consider the following sequence of limits:

$$\begin{array}{c} \mathbf{K} \\ \square \\ \mathbf{I} \end{array} \begin{array}{c} j \\ x \\ \ell \end{array} := (-s)^{-1_{\ell > 0}} \mathcal{L}_{y/x}^{(M)}(\mathbf{I}, j; \mathbf{K}, \ell) \Big|_{y \rightarrow s} \Big|_{q^M \rightarrow s^{-2}} \Big|_{s \rightarrow 0}.$$

- ▶ The weights of this model can be computed explicitly:

$ \begin{array}{c} \mathbf{I} \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} 0 \\ \square \\ 0 \end{array} $ $ \begin{array}{c} 1 \end{array} $	$ \begin{array}{c} \mathbf{I} \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} i \\ \square \\ i \end{array} $ $ \begin{array}{c} xq^{I_{[i+1,n]}} \end{array} $	$ \begin{array}{c} \mathbf{I}_i^- \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} 0 \\ \square \\ i \end{array} $ $ \begin{array}{c} x(1 - q^{I_i})q^{I_{[i+1,n]}} \end{array} $
$ \begin{array}{c} \mathbf{I}_i^+ \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} i \\ \square \\ 0 \end{array} $ $ \begin{array}{c} 1 \end{array} $	$ \begin{array}{c} \mathbf{I}_{ij}^{+-} \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} i \\ \square \\ j \end{array} $ $ \begin{array}{c} x(1 - q^{I_j})q^{I_{[j+1,n]}} \end{array} $	$ \begin{array}{c} \mathbf{I}_{ji}^{+-} \\ \square \\ \mathbf{I} \end{array} $ $ \begin{array}{c} j \\ \square \\ i \end{array} $ $ \begin{array}{c} 0 \end{array} $

where it is assumed that $1 \leq i < j \leq n$.

2: Fusion and higher-spin vertex models

- ▶ The model satisfies the following Yang–Baxter (RLL) equation:

$$\sum_{0 \leq k_1, k_2 \leq n} \sum_{K \in \mathbb{N}^n} \begin{array}{c} \textcircled{x} \quad i_1 \\ \textcircled{y} \quad i_2 \\ \begin{array}{|c|} \hline J \\ \hline y \\ \hline K \\ \hline x \\ \hline I \\ \hline \end{array} \\ \begin{array}{l} k_2 \nearrow \\ \searrow k_1 \end{array} \\ j_2 \\ j_1 \end{array} = \sum_{0 \leq k_1, k_2 \leq n} \sum_{K \in \mathbb{N}^n} \begin{array}{c} \textcircled{x} \quad i_1 \\ \textcircled{y} \quad i_2 \\ \begin{array}{|c|} \hline J \\ \hline x \\ \hline K \\ \hline y \\ \hline I \\ \hline \end{array} \\ \begin{array}{l} k_1 \nearrow \\ \searrow k_2 \end{array} \\ j_2 \\ j_1 \end{array}$$

2: Non-symmetric Hall–Littlewood polynomials, E_μ

- ▶ In a similar vein to before, we define

$$E_\mu(x_1, \dots, x_n) :=$$

	$\mu^{(0)}$		\dots	\dots		$\mu^{(N)}$	
n	x_n		\dots	\dots		x_n	0
\vdots	\vdots					\vdots	\vdots
\vdots	\vdots					\vdots	\vdots
2	x_2		\dots	\dots		x_2	0
1	x_1		\dots	\dots		x_1	0
	0		\dots	\dots		0	

- ▶ Here we have defined

$$N = \max_{1 \leq i \leq n} (\mu_i), \quad \mu^{(j)} = \sum_{i=1}^n \mathbf{1}(\mu_i = j) e_i.$$

- ▶ These functions obey two (uniquely-determining) properties:

$$E_\delta(x_1, \dots, x_n) = \prod_{i=1}^n x_i^{\delta_i}, \quad \delta = (\delta_1 \leq \dots \leq \delta_n).$$

$$T_i \cdot E_\mu(x_1, \dots, x_n) = E_{\mathfrak{s}_i \mu}(x_1, \dots, x_n), \quad \mu_i < \mu_{i+1}, \quad T_i := q - \frac{x_i - qx_{i+1}}{x_i - x_{i+1}} (1 - \mathfrak{s}_i).$$

2: Non-symmetric Hall–Littlewood processes

- ▶ One can extend this probability measure to the non-symmetric case, as follows:

$$\mathbb{P}_{m,n}^{\text{ns}}(\mu, \lambda) := Q_{\mu^+/\lambda(1)}(x_1) \cdot \prod_{j=2}^n Q_{\lambda(j-1)/\lambda(j)}(x_j) \cdot E_{\mu}(y_1, \dots, y_m) \prod_{i=1}^n \prod_{j=1}^m \frac{1 - x_i y_j}{1 - q x_i y_j}.$$

- ▶ The measure satisfies the sum-to-unity property:

$$\sum_{\mu} \sum_{\lambda} \mathbb{P}_{m,n}^{\text{ns}}(\mu, \lambda) = 1.$$

- ▶ This can be proved by making use of the **branching rule** for **symmetric Hall–Littlewood polynomials**:

$$Q_{\lambda}(x_1, \dots, x_n) = \sum_{\nu \prec \lambda} Q_{\lambda/\nu}(x_1) Q_{\nu}(x_2, \dots, x_n),$$

and the **symmetrization property** of non-symmetric Hall–Littlewood polynomials:

$$\sum_{\mu: \mu^+ = \lambda} E_{\mu}(y_1, \dots, y_m) = P_{\lambda}(y_1, \dots, y_m).$$

- ▶ One can proceed to compute the averages of observables by using the action of **Cherednik–Dunkl operators**.

2: A remarkable distribution match

- ▶ The observables that will interest us will be the **zero set** $z(\mu)$ of the composition (μ_1, \dots, μ_m) , defined as

$$z(\mu) = \{1 \leq i \leq m : \mu_i = 0\},$$

and a further set

$$\zeta(\mu, \lambda) = \{1 \leq j \leq n : \ell(\lambda^{(j-1)}) - \ell(\lambda^{(j)}) = 0\}, \quad \lambda^{(0)} \equiv \mu^+, \quad \lambda^{(n)} \equiv \emptyset,$$

which records the instances where neighbouring partitions in the extended Gelfand–Tsetlin pattern $\mu^+ \succ \lambda^{(1)} \succ \dots \succ \lambda^{(n-1)} \succ \emptyset$ have the same length.

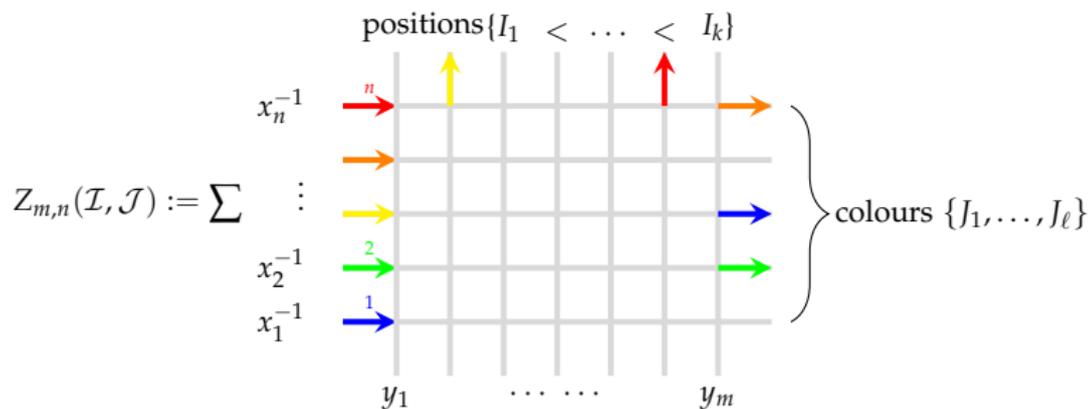
- ▶ In particular, we define

$$\mathbb{P}_{m,n}(\mathcal{I}, \mathcal{J}) := \sum_{\mu} \sum_{\lambda} \mathbb{P}_{m,n}^{\text{ns}}(\mu, \lambda) \cdot \mathbf{1}_{z(\mu)=\mathcal{I}} \cdot \mathbf{1}_{\zeta(\mu,\lambda)=\mathcal{J}},$$

which is the joint distribution of the random variables $z(\mu)$, $\zeta(\mu, \lambda)$ in the pair (μ, λ) sampled with respect to the non-symmetric Hall–Littlewood process.

2: A remarkable distribution match

- Compare this against the following probability distribution, of our starting vertex model in a quadrant:



Theorem

Fix two integers $m, n \geq 1$ and two sets $\mathcal{I} = \{1 \leq I_1 < \dots < I_k \leq m\}$ and $\mathcal{J} = \{1 \leq J_1 < \dots < J_\ell \leq n\}$ whose cardinalities satisfy $k + \ell = n$. The following equality of distributions holds:

$$\mathbb{P}_{m,n}(\mathcal{I}, \mathcal{J}) = Z_{m,n}(\mathcal{I}, \mathcal{J}).$$