

Expansion coefficients and the Yang–Baxter equation

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Joint work with **Ajeeth Gunna**

Motivation

- Symmetric functions are central objects in algebraic combinatorics, representation theory and geometry.
- There are many well-studied bases of the ring of symmetric functions. We understand how to write most of them as **partition functions of vertex models**.
- The connection with integrability yields many nice features of the functions:

Symmetry	Commuting transfer matrices
Symmetrization identity	Coordinate Bethe Ansatz
Cauchy summation identity	Yang–Baxter algebra or RTT relation

- A much harder problem is to study the **expansion coefficients** from one basis to another:

$$F_\lambda(x) = \sum_{\mu} c_{\lambda,\mu} G_\mu(x).$$

Macdonald polynomials

- The Macdonald polynomials $P_\nu(x; q, t)$ depend on an alphabet $x = (x_1, x_2, \dots)$ and two further parameters q, t .
- They are uniquely characterized by the following two properties:

1

$$P_\nu(x; q, t) = m_\nu(x) + \sum_{\mu < \nu} C_{\mu, \nu}(q, t) m_\mu(x),$$

where $m_\mu(x)$ denotes a monomial symmetric function

$$m_\mu(x) = \text{Sym} \left(\prod_{i \geq 1} x_i^{\mu_i} \right)$$

2

$$\langle P_\lambda, P_\nu \rangle = 0, \quad \lambda \neq \nu$$

where the scalar product $\langle \cdot, \cdot \rangle$ is defined via its action on the power sum basis:

$$\langle p_\lambda, p_\nu \rangle = \delta_{\lambda, \nu} \cdot z_\lambda \cdot \prod_{i=1}^{\ell(\lambda)} \frac{1 - q^{\lambda_i}}{1 - t^{\lambda_i}} \quad p_\lambda(x) = \prod_{i \geq 1} \left(\sum_{k \geq 1} x_k^{\lambda_i} \right)$$

Macdonald polynomials

- Many reductions are of interest. When $q = t$, one has the following reduction to Schur polynomials:

$$P_\nu(x; q, q) = s_\nu(x) = \det_{1 \leq i, j \leq \ell(\nu)} \left(h_{\nu_i - i + j}(x) \right),$$

where $h_k(x)$ denotes a complete symmetric function

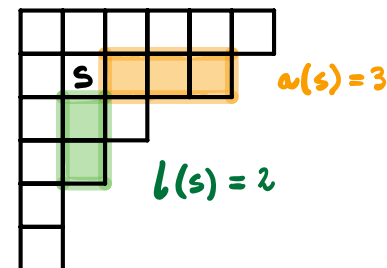
$$h_k(x) = \text{Coeff}_{z^k} \left\{ \prod_{i \geq 1} \frac{1}{1 - x_i z} \right\}$$

- In this talk we consider the integral form of the Macdonald polynomials:

$$J_\nu(x; q, t) = c_\nu(q, t) P_\nu(x; q, t)$$

where

$$c_\nu(q, t) = \prod_{s \in \nu} \left(1 - q^{a(s)} t^{l(s)+1} \right)$$



Macdonald polynomials

- The integral Macdonald polynomials take their name from the fact that

$$J_\nu(x; q, t) = \sum_{\mu \leq \nu} D_{\mu, \nu}(q, t) m_\mu(x),$$

where $D_{\mu, \nu}(q, t) \in \mathbb{Z}[q, t]$.

- This was originally conjectured by Macdonald, and later proved by [\[Sahi 96\]](#) and [\[Knop 97\]](#).
- In fact, one can get stronger results for these coefficients.

Macdonald polynomials

- For any $i, j \in \mathbb{Z}_{\geq 0}$ and $S \subset \mathbb{Z}_{\geq 0}^2$ define

$$f_S^{\pm, i, j}(q, t) = \pm q^i t^j \prod_{(\alpha, \beta) \in S} (1 - q^\alpha t^\beta).$$

Then for any $k \in \mathbb{N}$, define

$$\mathbb{F}_k^\pm(q, t) = \left\{ f_S^{\pm, i, j}(q, t) \mid i, j \in \mathbb{Z}_{\geq 0}, S \subset \mathbb{Z}_{\geq 0}^2 : |S| = k \right\}.$$

Theorem (Haglund–Haiman–Loehr 06)

There exists a family of tableaux T such that

$$J_\nu(x; q, t) = \sum_{T \text{ of shape } \nu} W_T(q, t) x^T$$

where $W_T(q, t) \in \mathbb{F}_{|\nu|}^+(q, t)$ is explicit.

Macdonald into Schur basis

- The topic of this talk is the expansion of $J_\nu(x; q, t)$ over the Schur basis:

$$J_\nu(x; q, t) = \sum_{\lambda \leq \nu} E_{\lambda, \nu}(q, t) s_\lambda(x)$$

where $E_{\lambda, \nu}(q, t) \in \mathbb{Z}[q, t]$.

Conjecture (Haglund 10)

For any partitions λ, ν such that $|\lambda| = |\nu|$ and $k \in \mathbb{N}$, one has

$$\frac{E_{\lambda, \nu}(q, q^k)}{(1 - q)^{|\nu|}} \in \mathbb{N}[q].$$

Partial progress by [\[Bhattacharya 22\]](#).

Main result

Theorem (Gunna–W 25+)

There exists an explicit partition function Z such that

$$E_{\lambda,\nu}(q,t) = \sum_{\mathcal{C} \text{ with frame } (\lambda,\nu)} W_{\mathcal{C}}(q,t),$$

where $W_{\mathcal{C}}(q,t) \in \mathbb{F}_{|v|}^{\pm}(q,t)$ is explicit.

- This just fails, however, to prove Haglund's conjecture:

Corollary

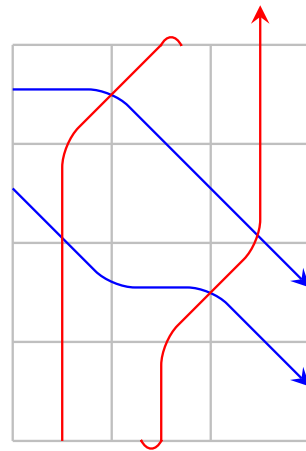
$$\frac{E_{\lambda,\nu}(q,q^k)}{(1-q)^{|v|}} = \sum_{\mathcal{C} \text{ with frame } (\lambda,\nu)} \text{sgn}(\mathcal{C}) \mathcal{P}_{\mathcal{C}}(q),$$

where $\text{sgn}(\mathcal{C}) \in \{\pm 1\}$ may be explicitly defined, and with each $\mathcal{P}_{\mathcal{C}}(q) \in \mathbb{N}[q]$.

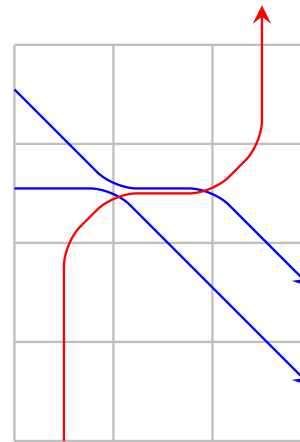
Examples

■ Here are some examples of the combinatorial formula.

■ $\nu = (2), \lambda = (1, 1)$



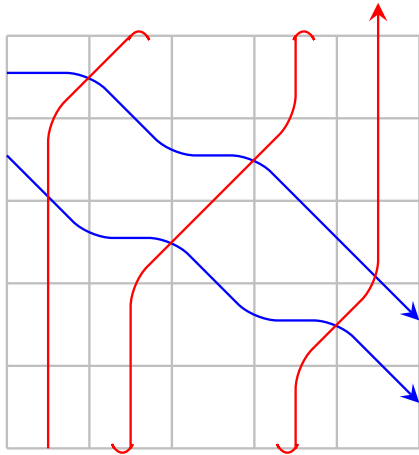
$$q(1-t)^2$$



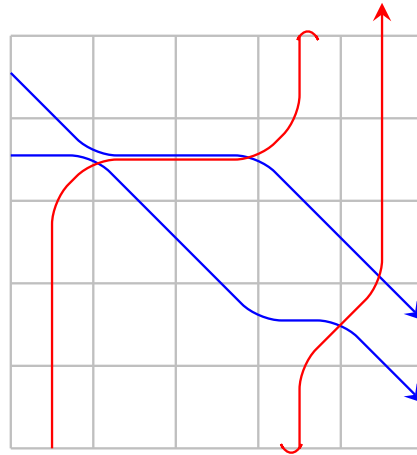
$$-(1-q)(1-t)t$$

Examples

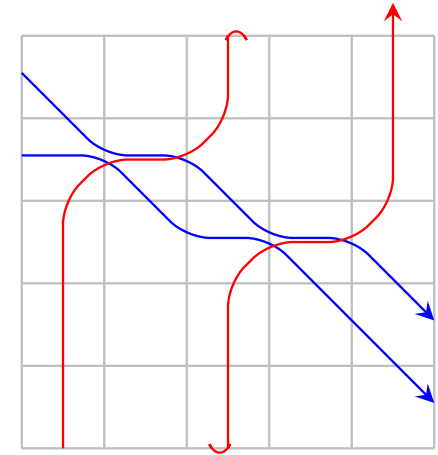
■ $\nu = (4), \lambda = (2, 2)$



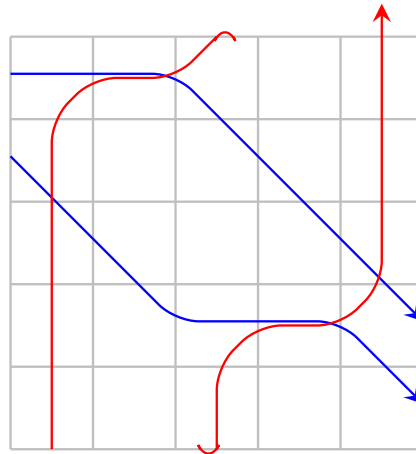
$$q^4(1-t)^4$$



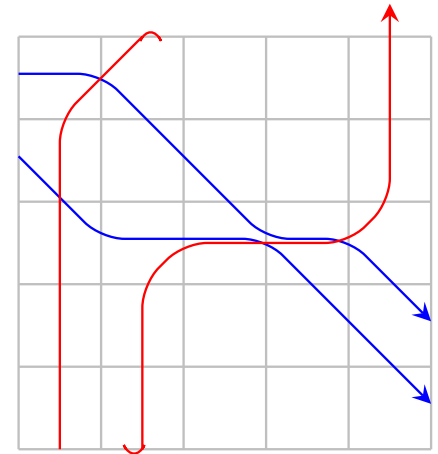
$$-q(1-q^3)(1-t)^2t(1-q^2t)$$



$$(1-q)q^2(1-q^3)(1-t)^2t^2$$



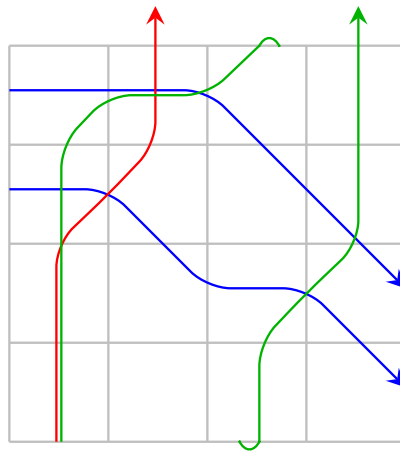
$$q^2(1-t)^2(1-qt)(1-q^3t)$$



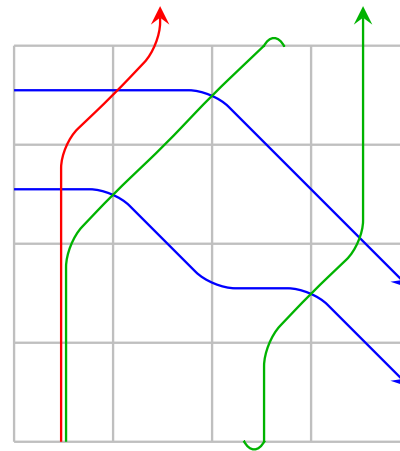
$$-(1-q)q^3(1-t)^2t(1-q^2t)$$

Examples

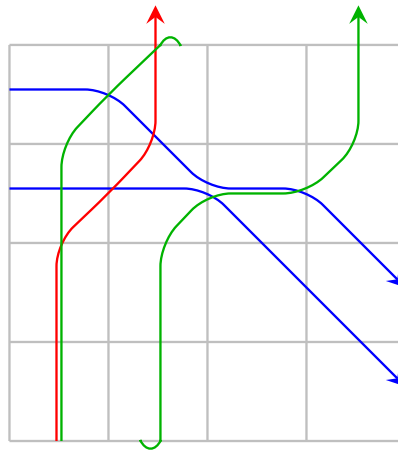
■ $v = (3, 1)$, $\lambda = (2, 2)$



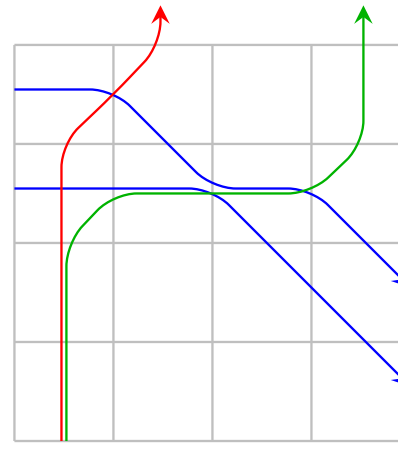
$$q(1-t)^3 t (1-q^2 t)$$



$$q(1-t)^4$$



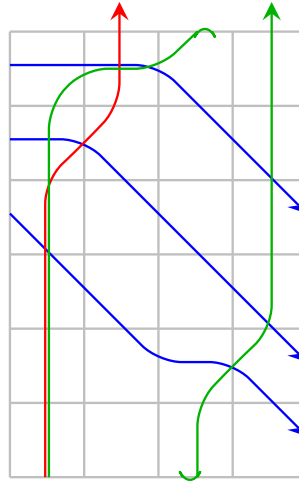
$$-(1-q)q^2(1-t)^3 t^2$$



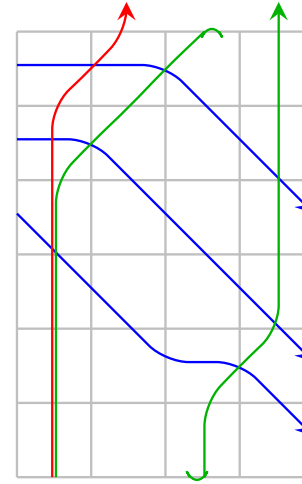
$$-(1-q)(1-t)^2 t (1-q^2 t)$$

Examples

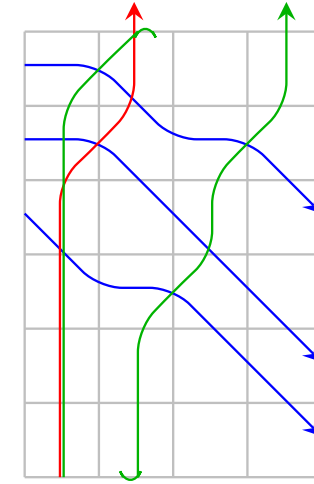
■ $\nu = (3, 1)$, $\lambda = (2, 1, 1)$



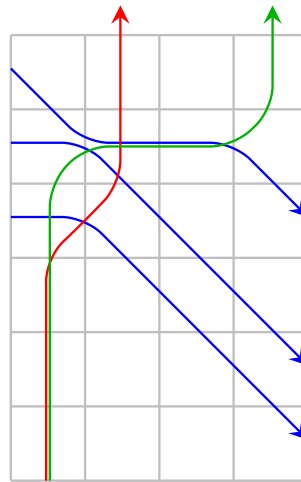
$$q(1-t)^3 t (1-q^2 t)$$



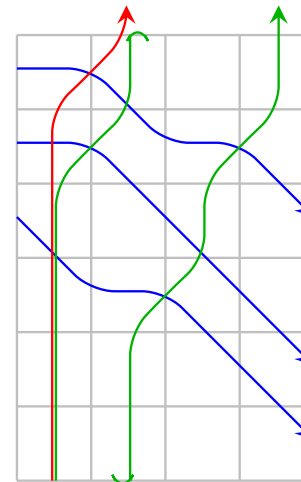
$$q(1-t)^4$$



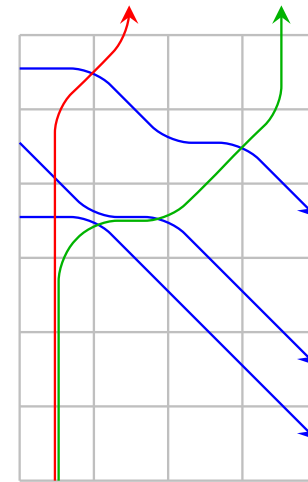
$$q^2(1-t)^4 t$$



$$-(1-q^2)(1-t)^2 t^2 (1-qt)$$



$$q^2(1-t)^4$$



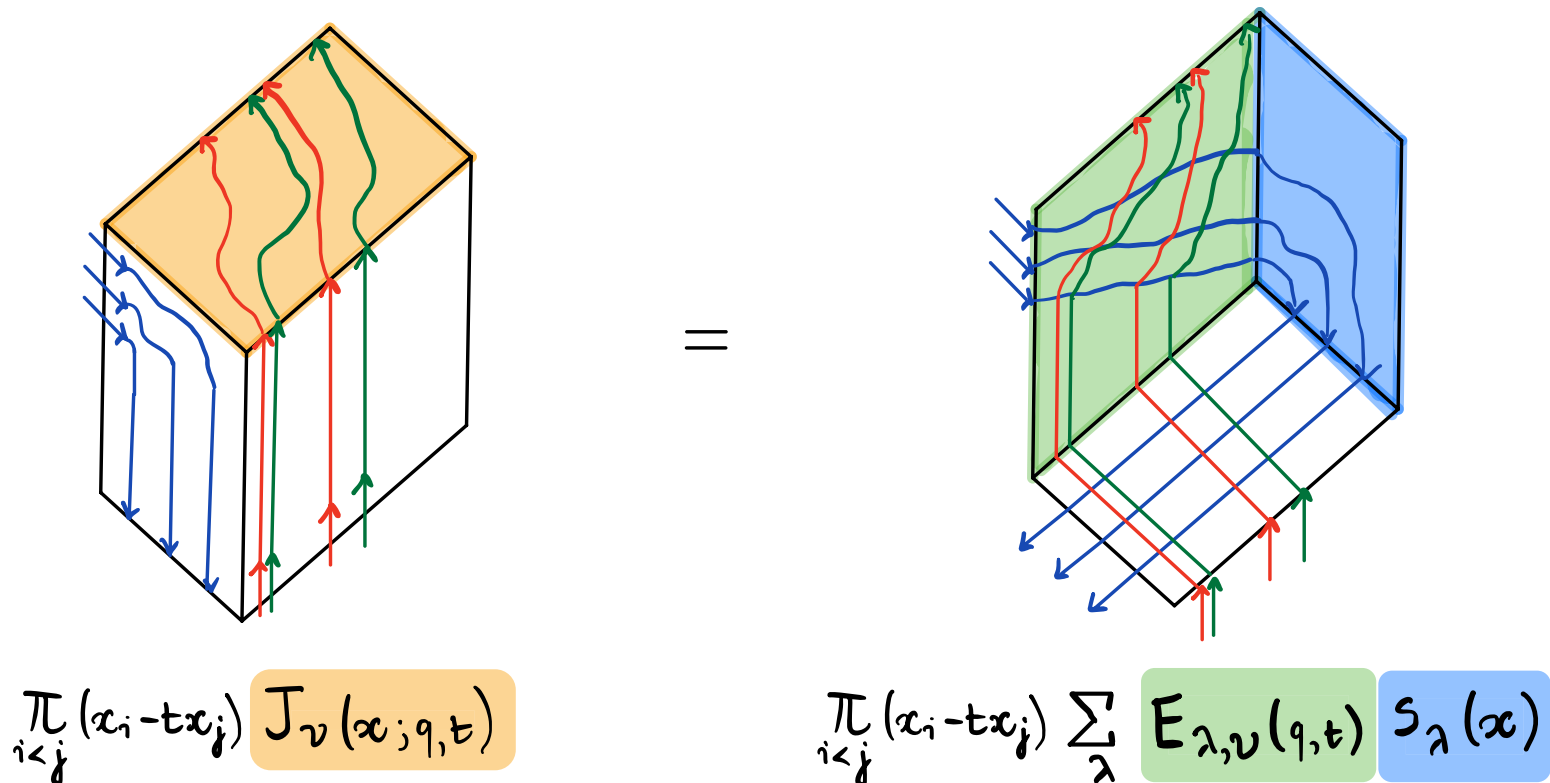
$$-(1-q^2)(1-t)^3 t$$

Proof via the Yang–Baxter equation

- The vertex weights used in the construction of Z satisfy a **Yang–Baxter equation**.
- The verification of the identity

$$J_\nu(x; q, t) = \sum_{\lambda \leq \nu} E_{\lambda, \nu}(q, t) s_\lambda(x)$$

proceeds by analysis of the left and right hand sides of our Yang–Baxter equation:



$$\prod_{i < j} (x_i - tx_j) J_\nu(x; q, t) = \prod_{i < j} (x_i - tx_j) \sum_{\lambda} E_{\lambda, \nu}(q, t) s_\lambda(x)$$

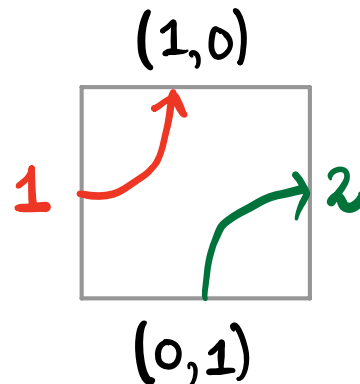
Macdonald polynomials from vertex models

- Consider the following vertex model:

$$w_x(A, b; C, d) = x \rightarrow b \begin{array}{c} \uparrow c \\ \text{---} \\ \downarrow A \end{array} \rightarrow d = x \rightarrow b \begin{array}{c} \text{---} c \\ \square \\ \text{---} A \end{array} \rightarrow d$$

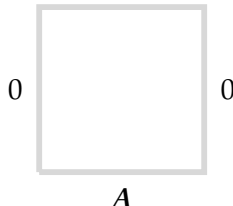
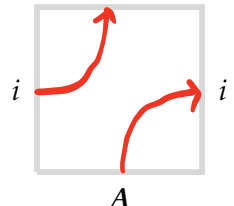
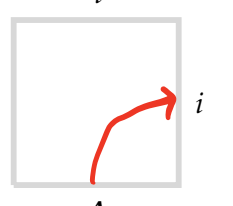
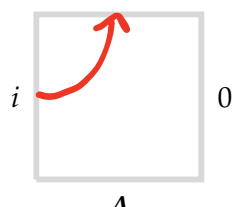
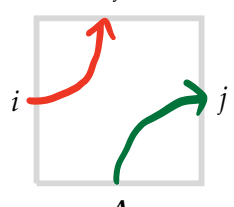
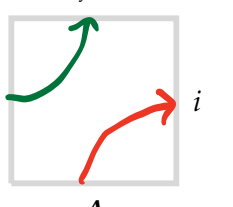
$$A, C \in \{0, 1\}^n, \quad b, d \in \{0, 1, \dots, n\}.$$

- We demand that $w_x(A, b; C, d) = 0$ unless $A + e_b = C + e_d$.



Macdonald polynomials from vertex models

- The nonzero vertices of the model are given as follows:

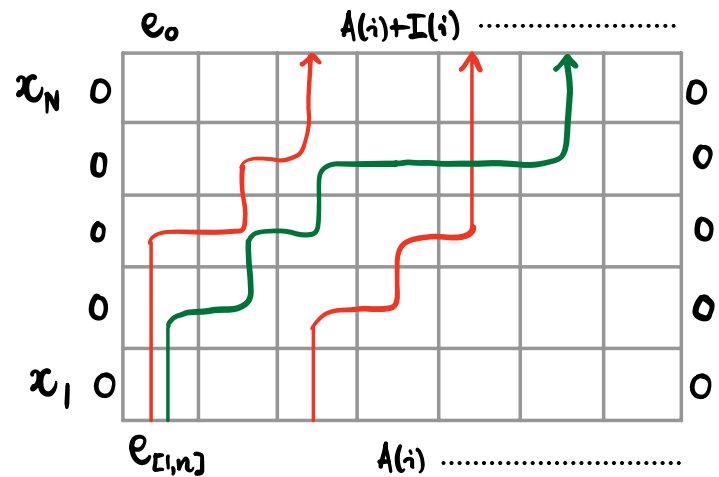
 1	 $(-1)^{A_i} t^{A_{[i,n]}} x$	 $(1 - t^{A_i}) t^{A_{(i,n]}} x$
 1	 $(1 - t^{A_j}) t^{A_{(j,n]}} x$	 0

where it is assumed that $0 \leq i < j \leq n$ in this tabulation.

Macdonald polynomials from vertex models

- Fix a composition $\nu = (1 \leq \nu_1 \leq \dots \leq \nu_n \leq M)$. Construct the partition function

$$\mathcal{P}_\nu(x; q, t) = \sum_{A(1), \dots, A(M)} \prod_{i=1}^n \prod_{j=1}^M w_{i,j}^{A(j)_i} x_N$$



where

$$I(j) = \sum_{i=1}^n \mathbf{1}_{\nu_i=j} \cdot e_i$$

and

$$w_{i,j} = \mathbf{1}_{\nu_i > j} \cdot q^{\nu_i - j}$$

for all $1 \leq i \leq n$ and $1 \leq j \leq M$.

Macdonald polynomials from vertex models

- Let ν^+ denote the partition obtained by sorting ν .

Theorem (Aggarwal–Borodin–W 21)

$$J_{\nu^+}(x; q, t) = \mathcal{P}_{\nu}(x; q, t).$$

- This result is motivated by, and proved in completely analogous fashion to an earlier formula for Macdonald polynomials in [\[Cantini–de Gier–W 15\]](#).
- By a relatively simple bijection, one may show that this expression is equivalent to the tableau formula of [\[Haglund–Haiman–Loehr 06\]](#).

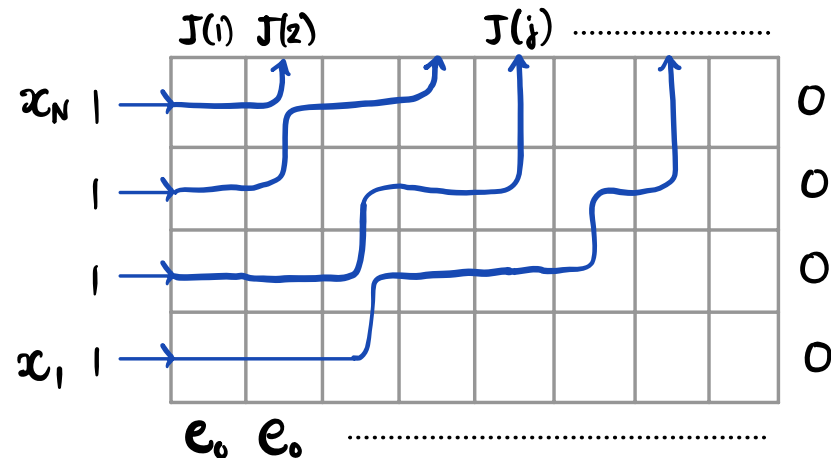
Schur polynomials from vertex models

- Fix a composition $\lambda = (0 \leq \lambda_1 \leq \dots \leq \lambda_N)$. Define a shifted version of λ as follows:

$$\tilde{\lambda} = \{\lambda_1 + 1, \lambda_2 + 2, \dots, \lambda_N + N\} \subset \mathbb{N}.$$

We define

$$\mathcal{Q}_\lambda(x; t) =$$



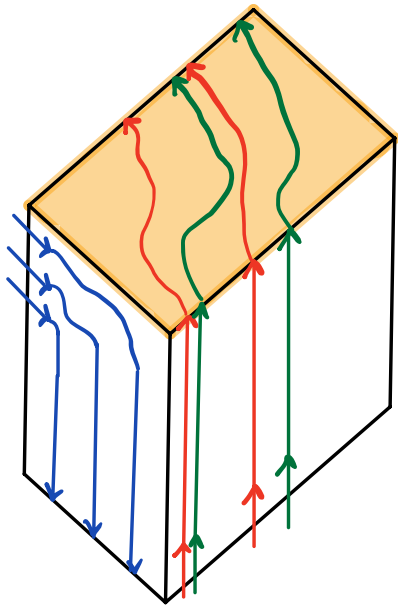
where $J(j) = \mathbf{1}_{j \in \tilde{\lambda}} \cdot e_1$ for all $j \geq 1$.

Theorem (Aggarwal–Borodin–W 21)

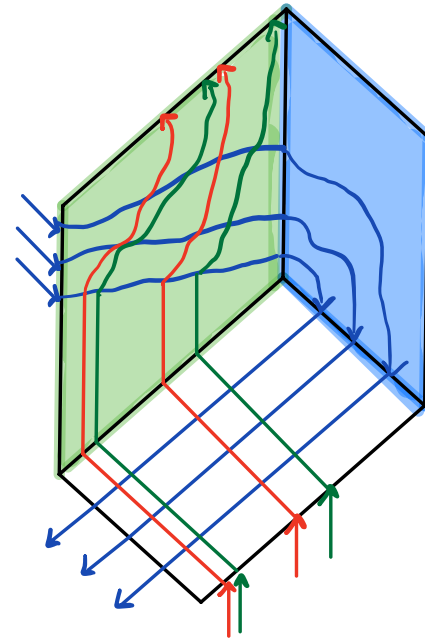
$$s_{\lambda^+}(x) \prod_{1 \leq i < j \leq N} (x_i - tx_j) = \mathcal{Q}_\lambda(x; t).$$

Thanks for listening!

$$J_v(x; q, t) = \sum_{\lambda \leq v} E_{\lambda, v}(q, t) s_{\lambda}(x)$$



=



$$\prod_{i < j} (x_i - tx_j) J_v(x; q, t)$$

$$\prod_{i < j} (x_i - tx_j) \sum_{\lambda} E_{\lambda, v}(q, t) s_{\lambda}(x)$$