

# Algebraic Fourier bases and probability

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## Rational Schur symmetric functions

$$S_{\lambda}(z_1, \dots, z_N) = \frac{\det [z_i^{\lambda_j + N - j}]_{i,j=1}^N}{\det [z_i^{N-j}]_{i,j=1}^N} \in \mathbb{C}[z_1^{\pm 1}, \dots, z_N^{\pm 1}]^{\text{Symm}}, \quad \lambda = (\lambda_1 \geq \dots \geq \lambda_N) \in \mathbb{Z}^N.$$

Two orthogonality relations:

$$\frac{1}{N!} \frac{1}{(2\pi i)^N} \oint_{|z_j|=1} \dots \oint S_{\lambda}(z) S_{\mu}(z^{-1}) \prod_{i < j} |z_i - z_j|^2 \frac{dz_1 \dots dz_N}{z_1 \dots z_N} = \mathbb{1}_{\lambda=\mu}$$

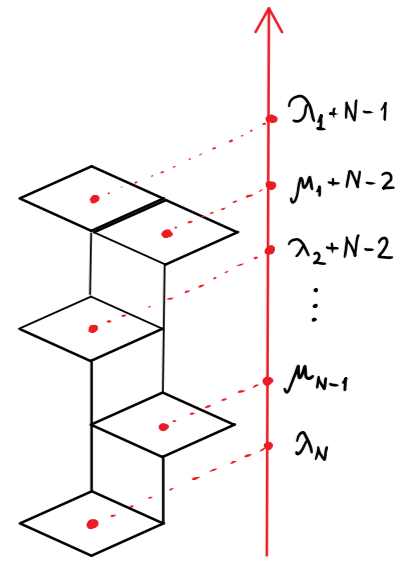
$$\sum_{-\infty < \lambda_N \leq \dots \leq \lambda_1 < +\infty} S_{\lambda}(z) S_{\lambda}(w^{-1}) \cdot \prod_{i < j} (z_i - z_j)(w_i^{-1} - w_j^{-1}) \cdot \frac{1}{w_1 \dots w_N} = \det [\delta(w_i - z_j)]_{i,j=1}^N$$

The Schur functions are characters of the (complex) irreducible representations of  $GL(N, \mathbb{C})$  (or  $U(N)$ ).

# Rational Schur symmetric functions

Branching rule (restriction from  $U(N)$  to  $U(N-1)$ )

$$S_\lambda(z_1, \dots, z_{N-1}, c) = \sum_{\lambda_1 \geq \mu_1 \geq \lambda_2 \geq \dots \geq \mu_{N-1} \geq \mu_N} c^{\sum \lambda_i - \sum \mu_i} \cdot S_\mu(z_1, \dots, z_{N-1})$$



Cauchy identity (reproducing kernel)

$$\sum_{\lambda_1 \geq \dots \geq \lambda_N \geq 0} S_\lambda(z_1, \dots, z_N) S_\lambda(w_1, \dots, w_N) = \prod_{i,j=1}^N \frac{1}{1 - z_i w_j}$$

$$\text{Pol} \left( \begin{array}{c} \square^N \\ \text{GL}(N) \\ \text{GL}(N) \end{array} \right) = \bigoplus_{\lambda} T_{\lambda}^1 \otimes T_{\lambda}^2$$

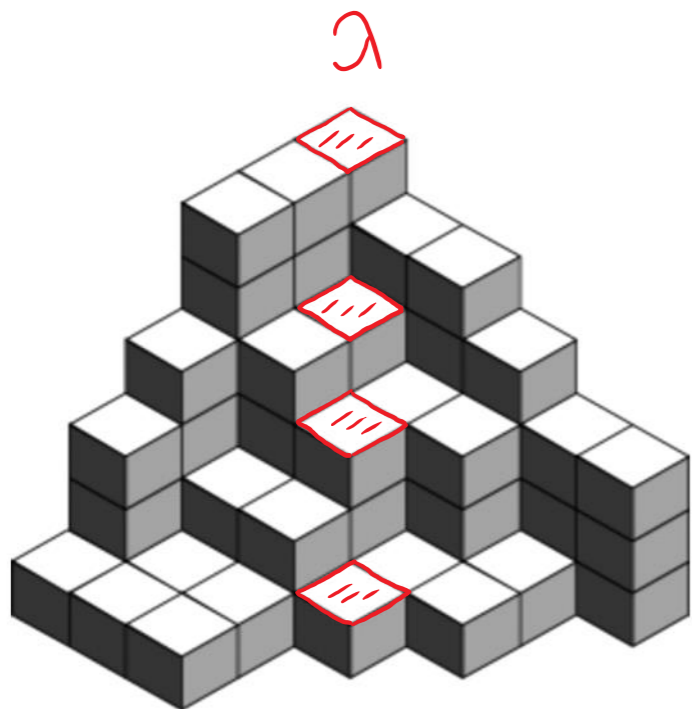
Difference operators

$$(z_1 + \dots + z_N) S_\lambda(z) = \sum_{\substack{\mu = \lambda + \vec{e}_r \\ (0, \dots, 0, 1, 0, \dots, 0)}} S_\mu(z)$$

$$\sum_{i=1}^N \prod_{j \neq i} \frac{z_j - qz_i}{z_j - z_i} S_\lambda(z_1, \dots, qz_i, \dots, z_N) = \left( \sum_{i=1}^N q^{\lambda_i + N - i} \right) \cdot S_\lambda(z)$$

Eigenvalues

# Random plane partitions



$$\begin{aligned} \text{weight (plane partition)} &= Q^{\text{volume}} \\ &= \left( S_\lambda(Q^{1/2}, Q^{3/2}, Q^{5/2}, \dots) \right)^2 \end{aligned}$$

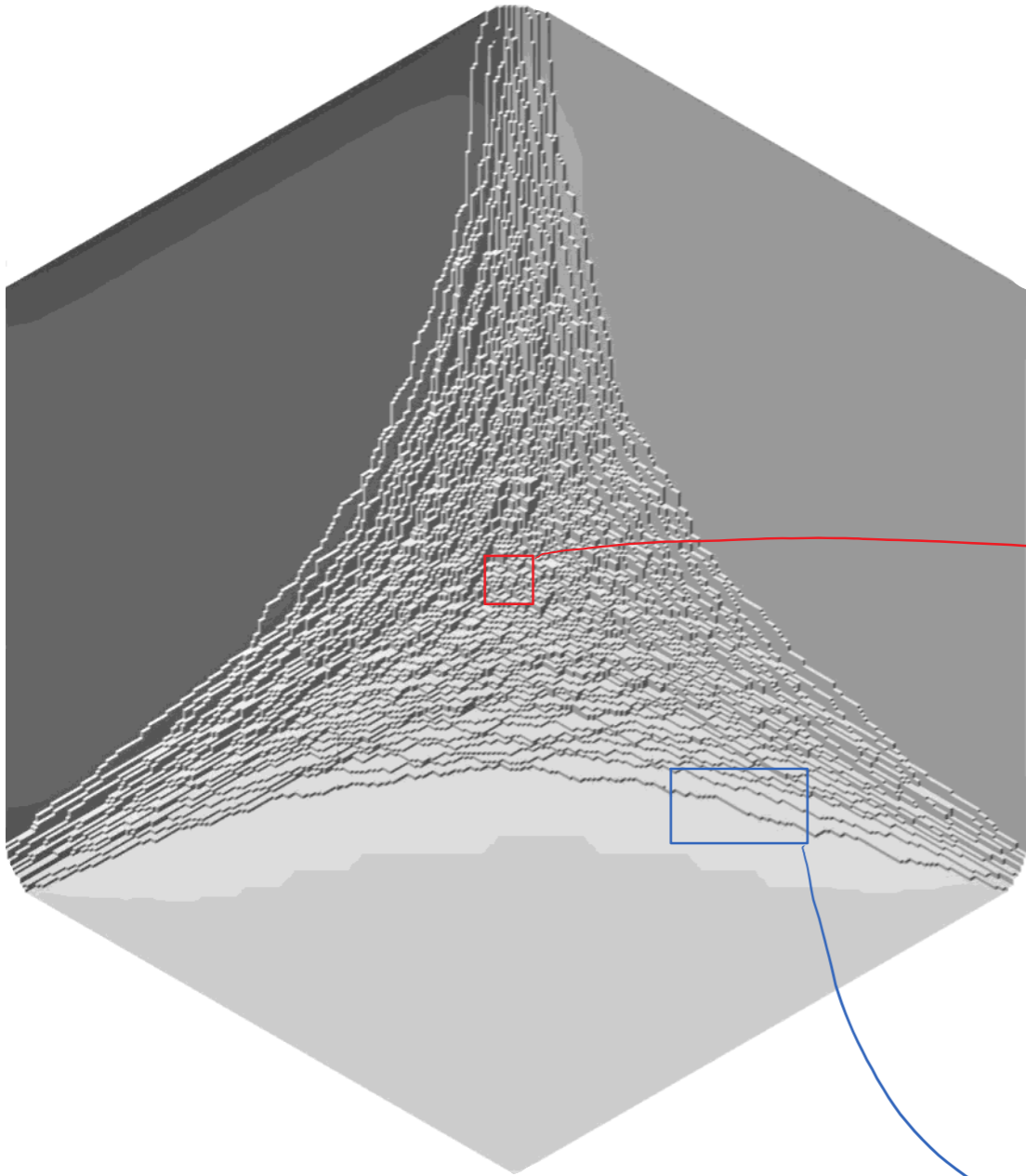
$$\sum_{\text{plane partitions}} Q^{\text{volume}} = \prod_{n \geq 1} \frac{1}{(1 - Q^n)^n} \quad \text{Cauchy/MacMahon identity}$$

$$E \left( \sum_i q_1^{\lambda_i + N - i} \dots \sum_i q_m^{\lambda_i + N - i} \right) = \frac{\mathcal{D}_1^{(q_1)} \dots \mathcal{D}_1^{(q_m)} \sum_{\lambda} S_\lambda(z) S_\lambda(w)}{\sum_{\lambda} S_\lambda(z) S_\lambda(w)} \Bigg|_{z=w=(Q^{1/2}, Q^{3/2}, \dots)}$$

$$\mathcal{D}_1^{(q)} S_\lambda(z) := \sum_{i=1}^N \prod_{j \neq i} \frac{z_j - qz_i}{z_j - z_i} S_\lambda(z_1, \dots, qz_i, \dots, z_N) = \left( \sum_{i=1}^N q^{\lambda_i + N - i} \right) \cdot S_\lambda(z)$$

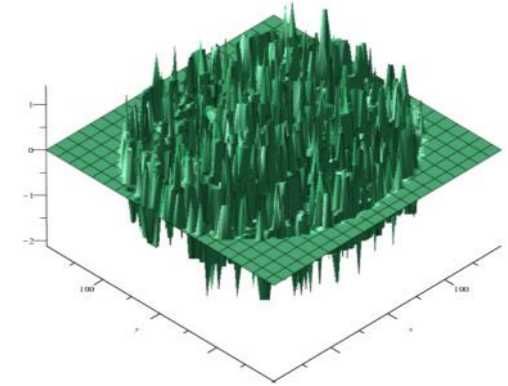
$$\sum_{\lambda_1 \geq \dots \geq \lambda_N \geq 0} S_\lambda(z_1, \dots, z_N) S_\lambda(w_1, \dots, w_N) = \prod_{i,j=1}^N \frac{1}{1 - z_i w_j}$$

# Random plane partitions

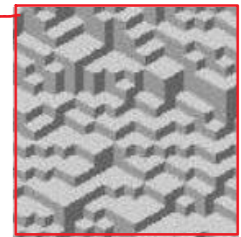


*Global limit shape (Wulff droplet or 'crystal',  
Ronkin function of a complex line)*

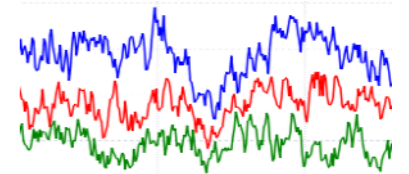
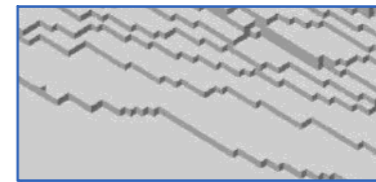
*Global fluctuations  
(Gaussian Free Field)*



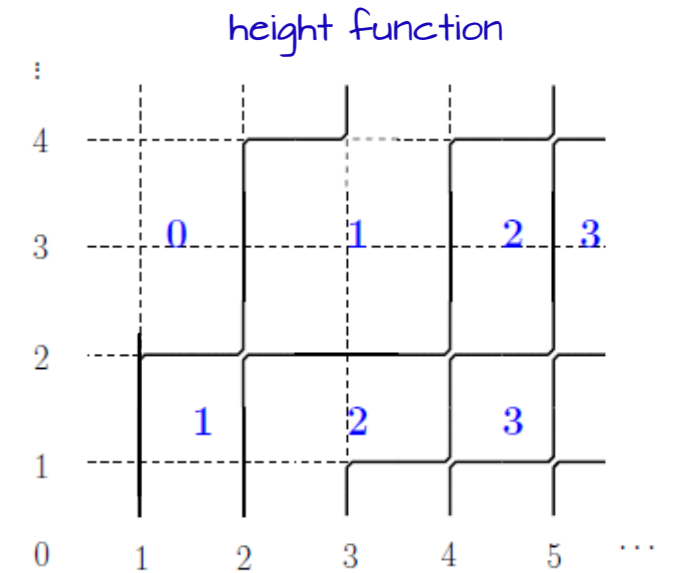
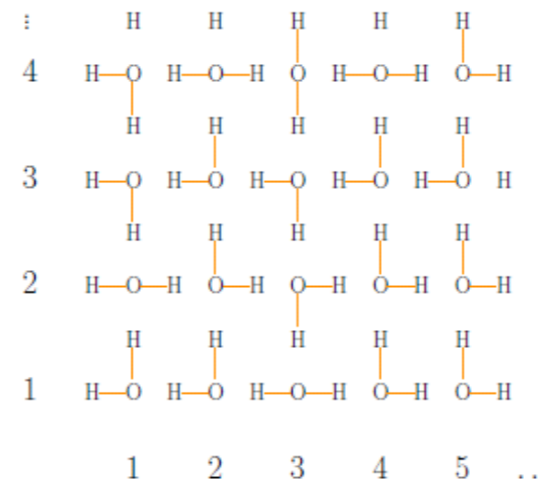
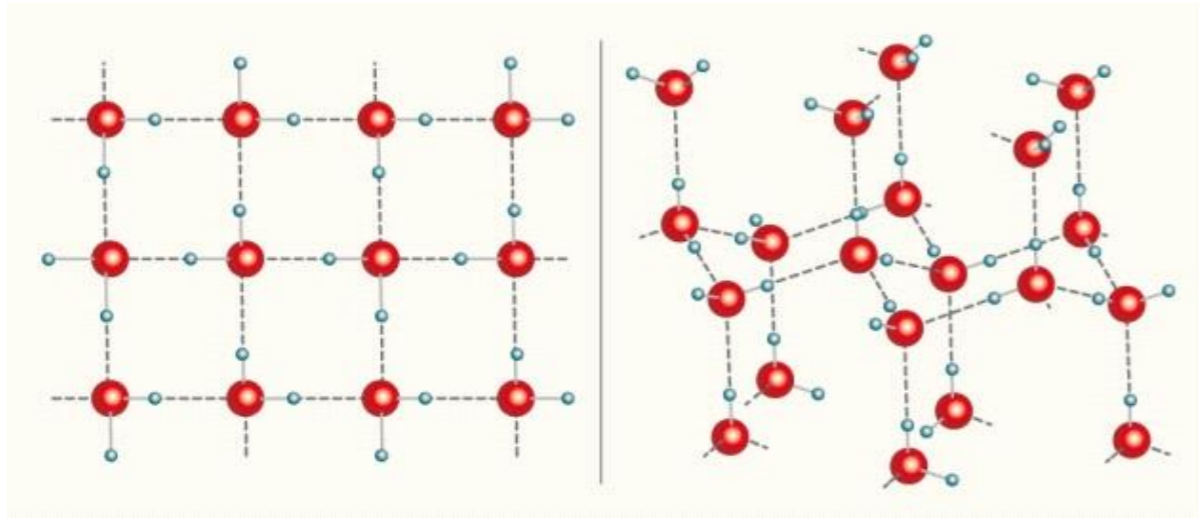
*Local correlations (translation  
invariant Gibbs measures)*



*Edge fluctuations  
(Airy processes)*

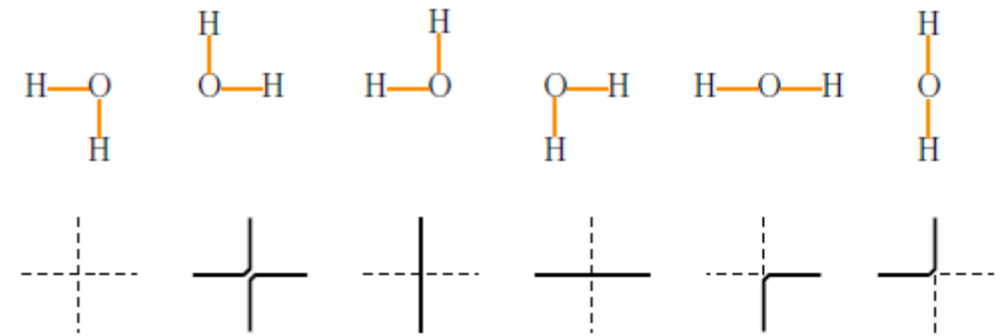


# The six vertex model (Pauling, 1935)



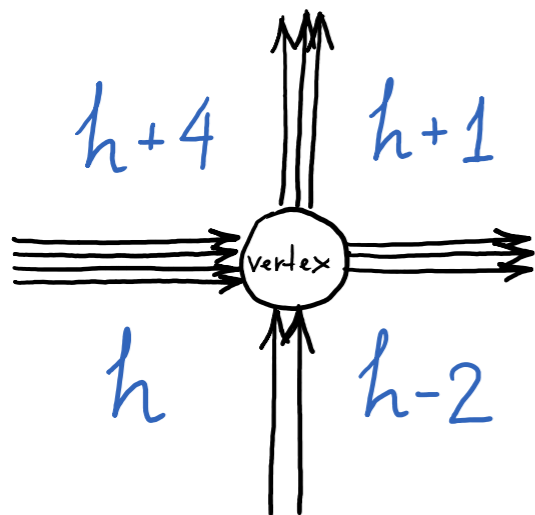
In 'square ice', which has been seen between graphene sheets, water molecules lock flat in a right-angled formation. The structure is strikingly different from familiar hexagonal ice (right).

From <http://www.nature.com/news/graphene-sandwich-makes-new-form-of-ice-1.17175>

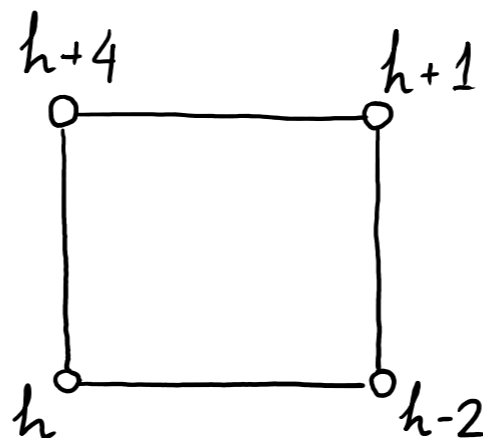


Lieb in 1967 computed the partition function of the square ice on a large torus – an estimate for the residual entropy of real ice.

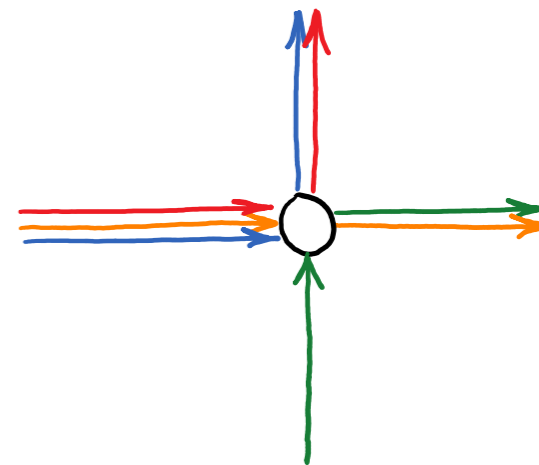
## More general models



Higher spin vertex models  
(only gradient of the height function matters)



SOS (Solid-On-Solid)  
or  
IRF (Interaction-Round-a-Face)  
models



Colored (higher rank)  
models

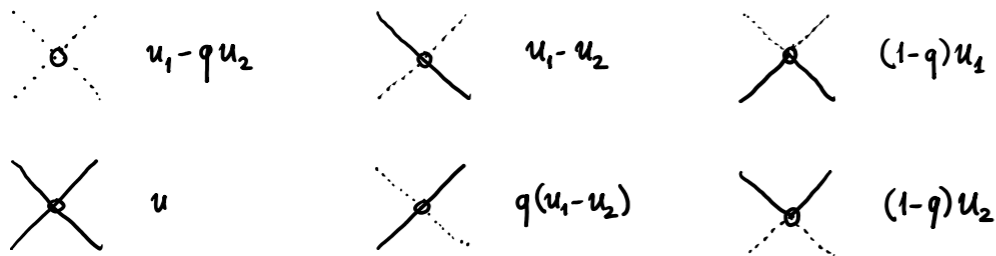
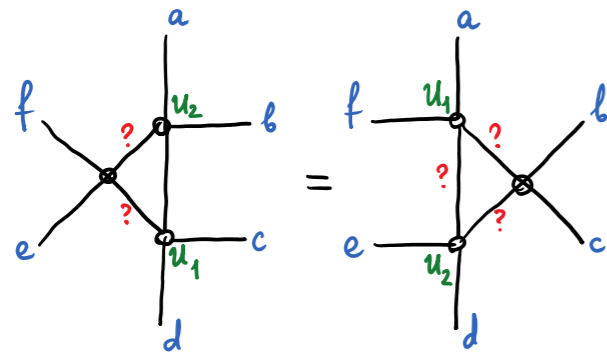
# Key property: commutation of transfer-matrices

$w_{u,s}$	$\frac{1 - sq^g u}{1 - su}$	$\frac{(1 - s^2 q^{g-1})u}{1 - su}$	$\frac{u - sq^g}{1 - su}$	$\frac{1 - q^{g+1}}{1 - su}$

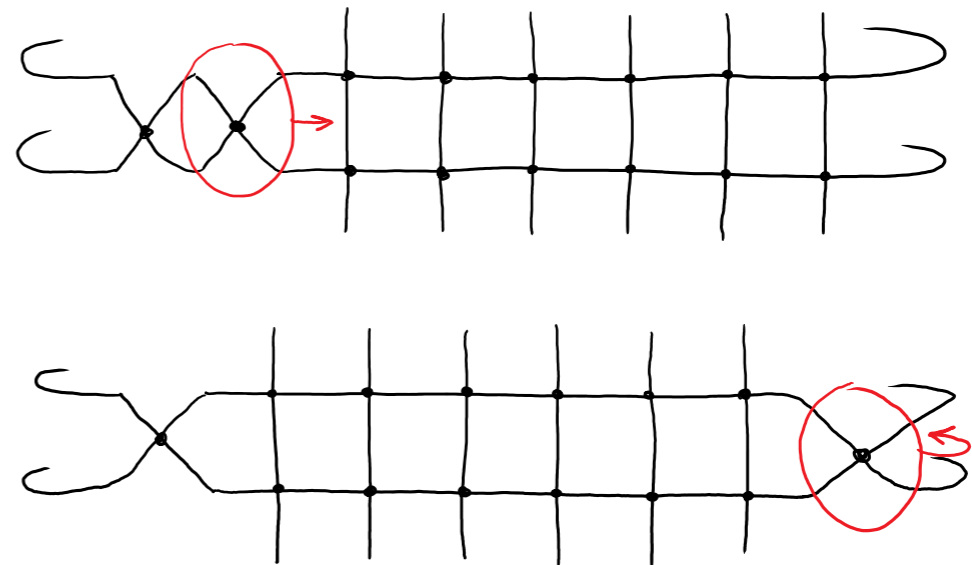
$$X_{\lambda/\mu}(u) = \left( \text{Diagram of a lattice path with nodes } \lambda_N, \lambda_3, \lambda_1 = \lambda_2 \text{ and weights } M_N, M_2 = M_3, M_1 \text{ enclosed in a red dashed oval} \right)$$

$$X_{\lambda/\mu}(u_1) X_{\lambda/\mu}(u_2) = X_{\lambda/\mu}(u_2) X_{\lambda/\mu}(u_1)$$

The **Yang-Baxter** (star-triangle) equation:

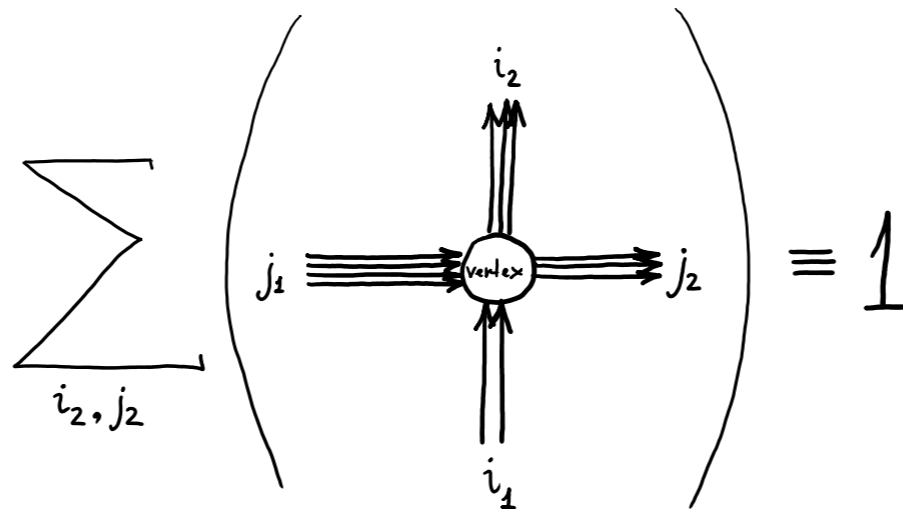


Zipper argument:

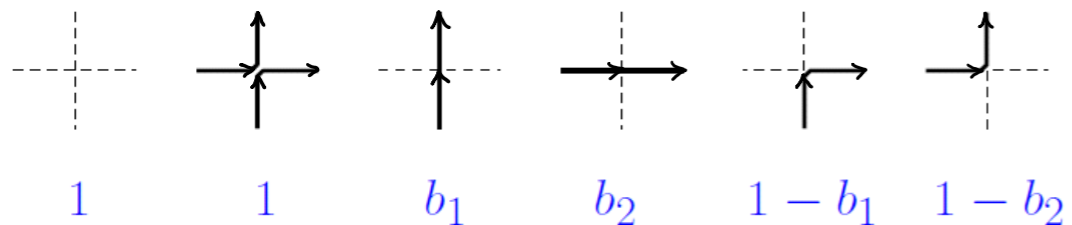




# New ingredient: stochasticity

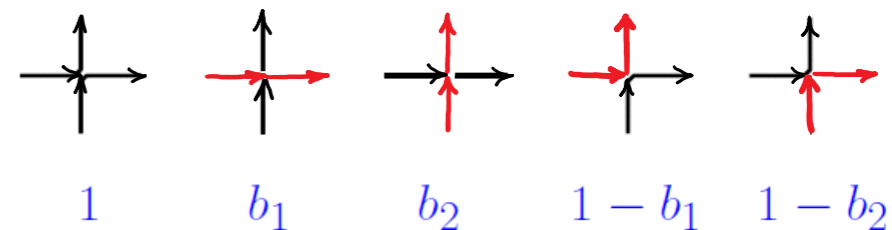


Example 1: stochastic six vertex model



[Gwa-Spohn 1992]

Example 2: *colored* stochastic six vertex model



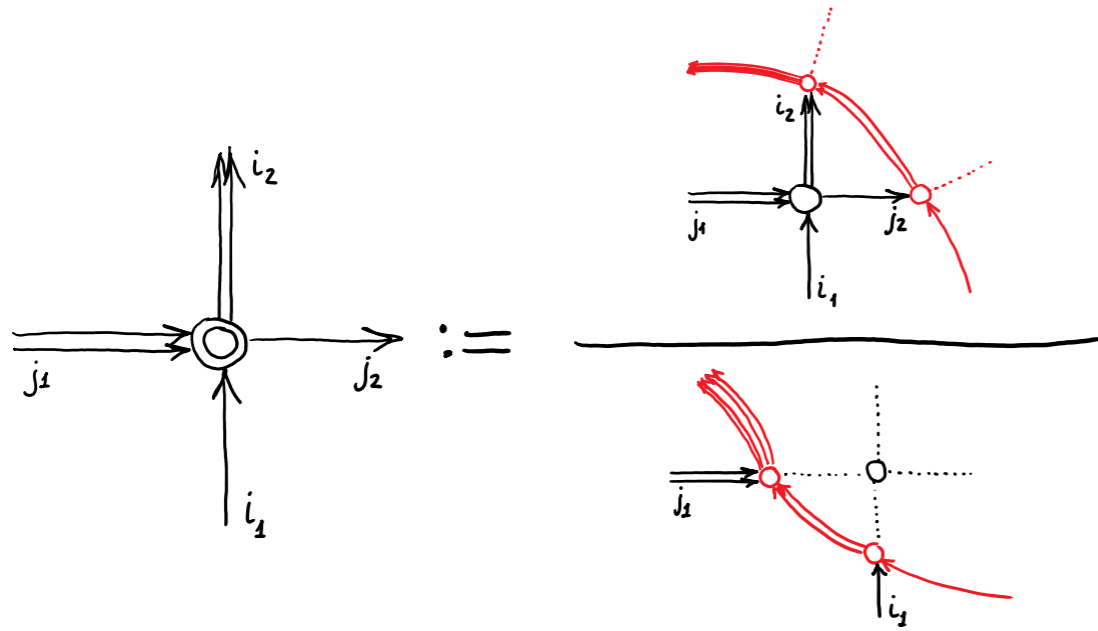
for colors  < 

[Kuniba-Mangazeev-Maruyama-Okado 2016]

[Kuan 2017]

[B-Wheeler 2018]

# Stochastization

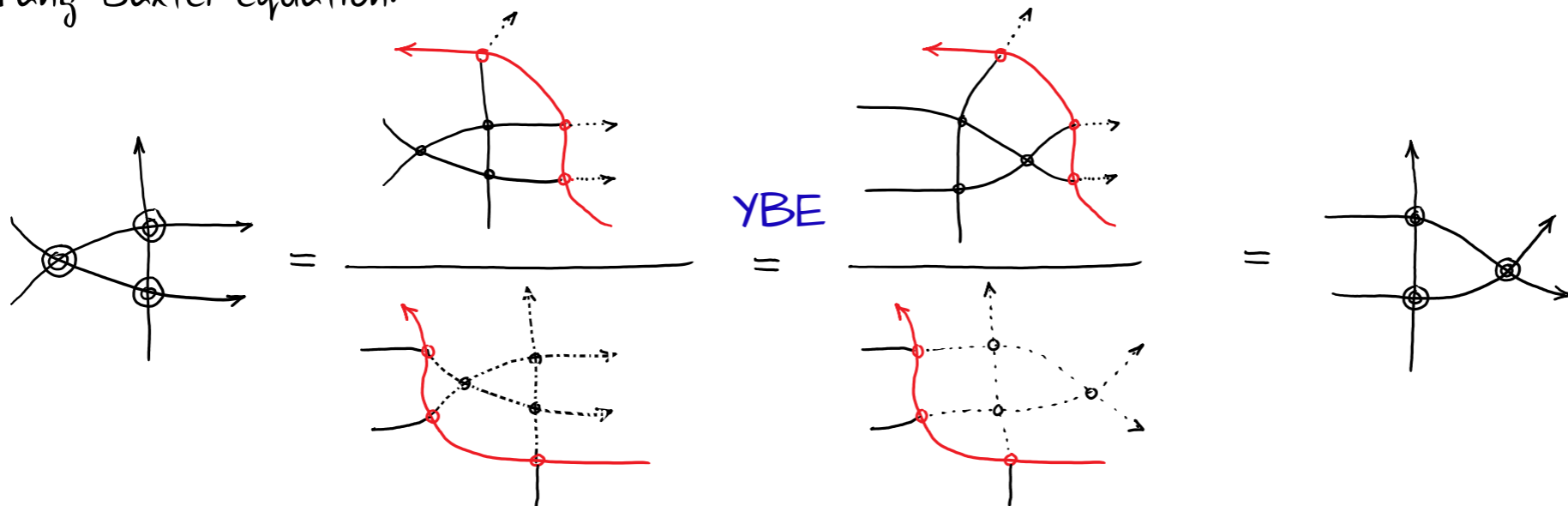


The Yang-Baxter equation implies

$$\sum_{i_2, j_2} \left[ \text{Diagram with crossing} \right] \left[ \text{Diagram with central node} \right] = 1$$

The diagram shows a summation over  $i_2, j_2$  of a crossing diagram multiplied by a central node diagram, which is equal to 1.

Stochastic Yang-Baxter equation:



# Higher spin stochastic six vertex model on $\mathbb{Z}$

$$P\left\{ \dots \uparrow_m \rightsquigarrow \dots \uparrow_m^m \right\} = \frac{1 - sq^m u}{1 - su} \quad P\left\{ \dots \rightarrow \uparrow_m \rightsquigarrow \dots \rightarrow \uparrow_m^m \right\} = \frac{s^2 q^m - su}{1 - su}$$

$$P\left\{ \dots \uparrow_m \rightsquigarrow \dots \rightarrow \uparrow_m^{m-1} \right\} = \frac{(q^m - 1) su}{1 - su} \quad P\left\{ \dots \rightarrow \uparrow_m \rightsquigarrow \dots \rightarrow \uparrow_m^{m+1} \right\} = \frac{-s^2 q^m}{1 - su}$$

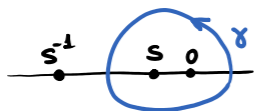
Complete basis of eigenfunctions of the transfer matrix  $X_{\lambda/\mu}(u) = \left( \dots \uparrow_{\lambda_N} \rightsquigarrow \uparrow_{\lambda_3} \rightsquigarrow \uparrow_{\lambda_1 = \lambda_2} \rightsquigarrow \dots \right)$

$$F_{\lambda}(u_1, \dots, u_N) = \frac{(1-q)^N}{\prod_{i=1}^N (1 - su_i)} \cdot \sum_{\sigma \in S_N} \sigma \left( \prod_{i < j} \frac{u_i - qu_j}{u_i - u_j} \cdot \prod_{i=1}^N \left( \frac{u_i - s}{1 - su_i} \right)^{\lambda_i} \right)$$

$$\lambda = (\lambda_1 \geq \dots \geq \lambda_N) \in \mathbb{Z}^N$$

Orthogonality

$$\frac{c(\lambda)}{(2\pi i)^N (1-q)^N N!} \oint_{\gamma} \dots \oint_{\gamma} \prod_{1 \leq A \neq B \leq N} \frac{u_A - u_B}{u_A - qu_B} F_{\lambda}(u_1, \dots, u_N) F_{\mu}(u_1^{-1}, \dots, u_N^{-1}) \prod_{i=1}^N \frac{du_i}{u_i} = \mathbb{1}_{\lambda = \mu}$$



[Tarasov-Varchenko 1997]

[Povolotsky 2013]

[B-Corwin-Petrov Sasamoto 2014-15]

[Corwin-Petrov 2014]

[B-Petrov 2016]

## Spin Hall-Littlewood symmetric rational functions

$$F_{\lambda}(u_1, \dots, u_N) = \frac{(1-q)^N}{\prod_{i=1}^N (1-su_i)} \cdot \sum_{\sigma \in S_N} \sigma \left( \prod_{i < j} \frac{u_i - qu_j}{u_i - u_j} \cdot \prod_{i=1}^N \left( \frac{u_i - s}{1 - su_i} \right)^{\lambda_i} \right)$$

$\lambda = (\lambda_1 \geq \dots \geq \lambda_N) \in \mathbb{Z}^N$

Specializing  $s=q=0$  brings us back to the Schur, while setting  $s=0$  yields the **Hall-Littlewood polynomials** that arise in connection with finite  $p$ -groups and representation theory of groups of  $p$ -adic type.

In  $\text{Span}\{e_{\lambda} : \lambda = (\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N \geq 0)\}$  define

$$A(u)e_{\lambda} = \sum_{\mu} \text{weight}_u \left( \begin{array}{c} \dots \quad \uparrow \quad \uparrow \quad \uparrow \quad \dots \\ \lambda_N \quad \lambda_3 \quad \lambda_1 = \lambda_2 \quad \lambda_1 \\ M_N \quad M_2 = M_3 \quad M_1 \end{array} \right) e_{\mu}$$

$$B(u)e_{\lambda} = \sum_{\mu} \text{weight}_u \left( \begin{array}{c} \dots \quad \uparrow \quad \uparrow \quad \uparrow \quad \dots \\ \lambda_N \quad \lambda_3 \quad \lambda_1 = \lambda_2 \quad \lambda_1 \\ M_{N+1} \quad M_N \quad M_2 = M_3 \quad M_1 \end{array} \right) e_{\mu}$$

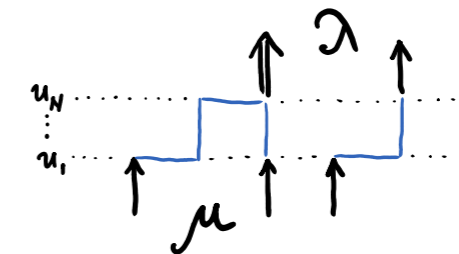
# Spin Hall-Littlewood symmetric rational functions

$$F_{\lambda}(u_1, \dots, u_N) = \langle B(u_1) \cdots B(u_N) e_{\emptyset}, e_{\lambda} \rangle = \text{weight} \left( \begin{array}{c} \lambda_N \dots \lambda_2 \lambda_1 \\ \begin{array}{c} u_N \rightarrow \\ \vdots \\ u_2 \rightarrow \\ u_1 \rightarrow \end{array} \\ 0 \end{array} \right)$$

$$= \frac{(1-q)^N}{\prod_{i=1}^N (1-su_i)} \cdot \sum_{\sigma \in S_N} \left( \prod \frac{u_i - qu_j}{u_i - u_j} \cdot \prod_{i=1}^N \left( \frac{u_i - s}{1 - su_i} \right)^{\lambda_i} \right)$$

$$G_{\lambda}(u_1, \dots, u_N) = \langle A(u_1) \cdots A(u_N) e_{(0^M)}, e_{\lambda} \rangle = \text{weight} \left( \begin{array}{c} \lambda_M \dots \lambda_2 \lambda_1 \\ \begin{array}{c} u_N \dots \\ \vdots \\ u_2 \dots \\ u_1 \uparrow \uparrow \uparrow \end{array} \\ 0 \end{array} \right)$$

More generally,

$$G_{\lambda/\mu}(u_1, \dots, u_N) = \langle A(u_1) \cdots A(u_N) e_{\mu}, e_{\lambda} \rangle$$


# Spin Hall-Littlewood symmetric rational functions

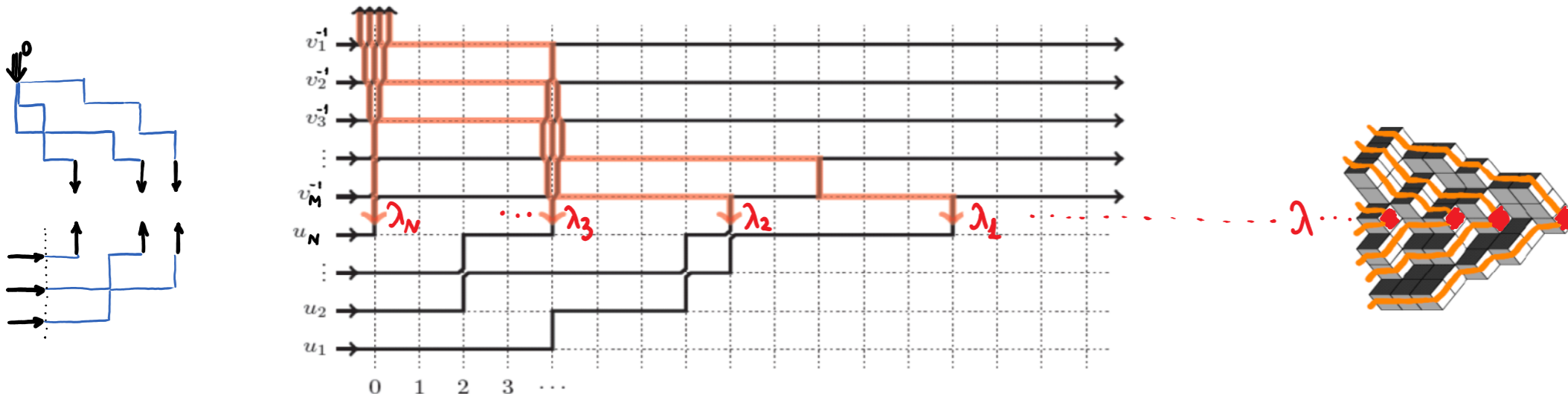
Difference operator (transfer-matrix)

$$\sum_{\mu} \frac{c(\mu)}{c(\lambda)} G_{\mu/\lambda}(v) \cdot F_{\mu}(u_1, \dots, u_N) = \prod_{i=1}^N \frac{1 - qu_i v}{1 - u_i v} \cdot F_{\lambda}(u_1, \dots, u_N)$$

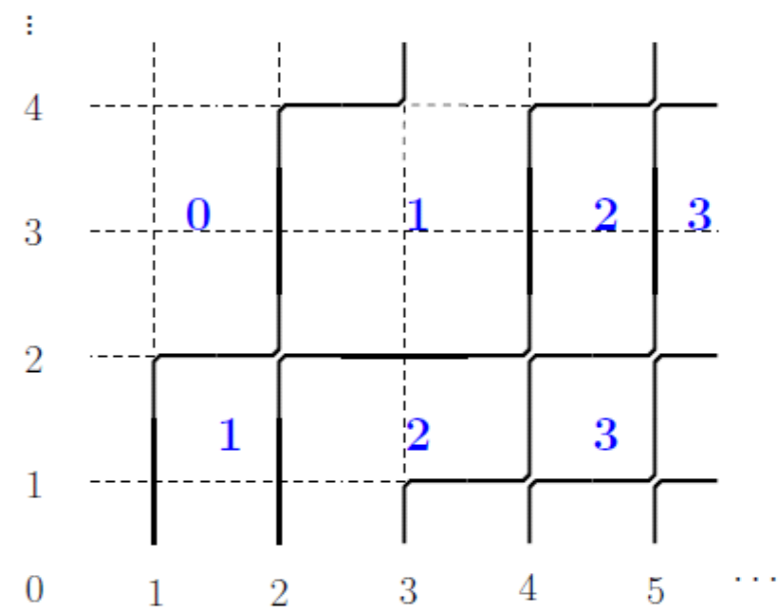
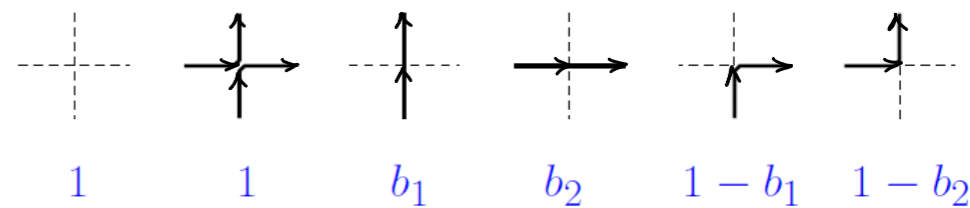
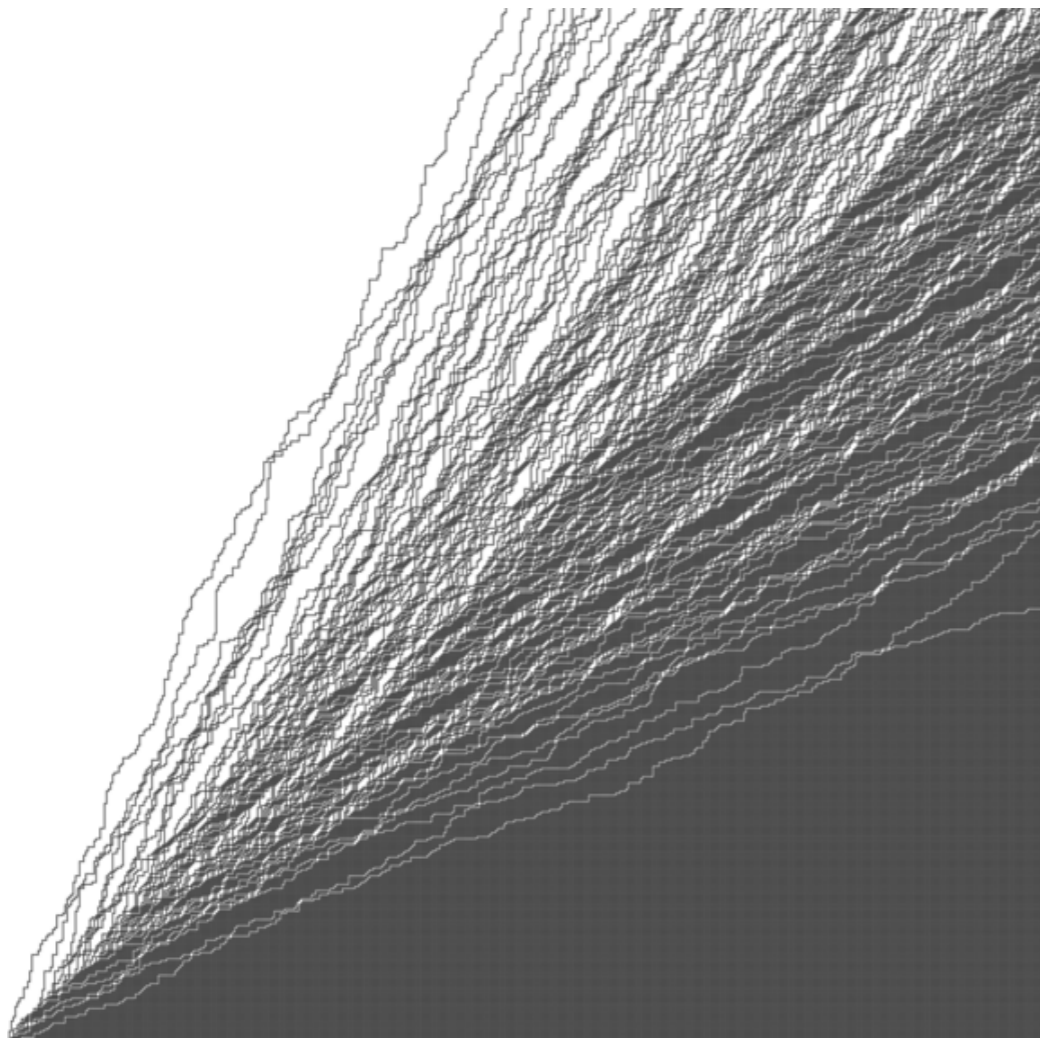
Cauchy identity [B.'14, B.-Petrov '16]

$$c(0^{m_0} 1^{m_1} 2^{m_2} \dots) = \prod_{k \geq 0} \frac{(s^2; q)_{m_k}}{(q; q)_{m_k}}$$

$$\sum_{\lambda = (\lambda_1 \geq \dots \geq \lambda_N \geq 0)} F_{\lambda}(u_1, \dots, u_N) \cdot \frac{c(\lambda)}{c(0^N)} G_{\lambda}(v_1, \dots, v_M) = \prod_{i=1}^N \frac{1 - q^i}{1 - su_i} \cdot \prod_{i,j} \frac{1 - qu_i v_j}{1 - u_i v_j}$$



# Stochastic six vertex model on $\mathbb{Z}$



Courtesy of [Leo Petrov](#)

# Stochastic six vertex model on $\mathbb{Z}$

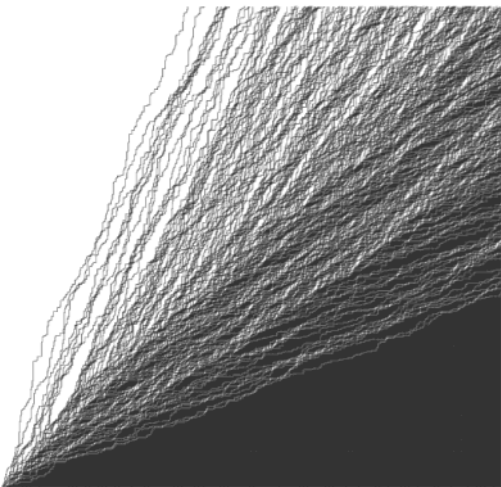
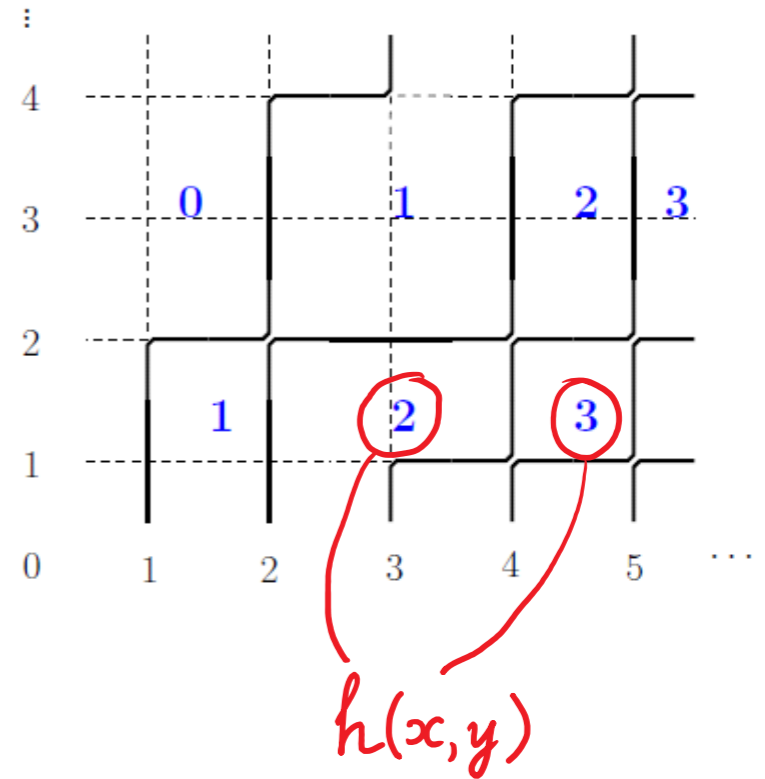
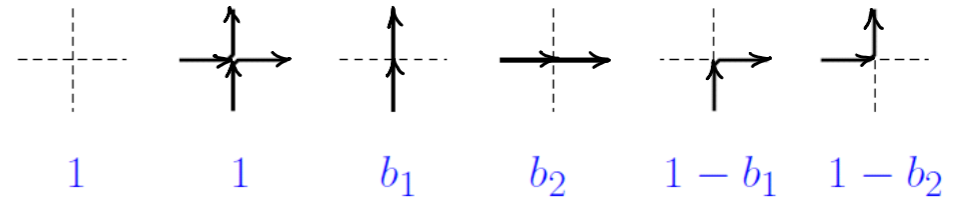
Theorem [B-Corwin-Gorin 2014]

Assume  $b_1 > b_2$ . Then for  $\frac{1-b_1}{1-b_2} < \frac{x}{y} < \frac{1-b_2}{1-b_1}$

$$\lim_{L \rightarrow \infty} \mathbb{P} \left\{ \frac{h(Lx, Ly) - L \cdot H(x, y)}{L^{1/3} \cdot \sigma(x, y)} \leq -s \right\} = F_{GUE}(s)$$

where  $H(x, y) = \frac{(\sqrt{x(1-b_2)} - \sqrt{y(1-b_1)})^2}{x-y}$ ,  $\sigma(x, y)$  is explicit,

$F_{GUE}(s)$  is the GUE Tracy-Widom distribution.



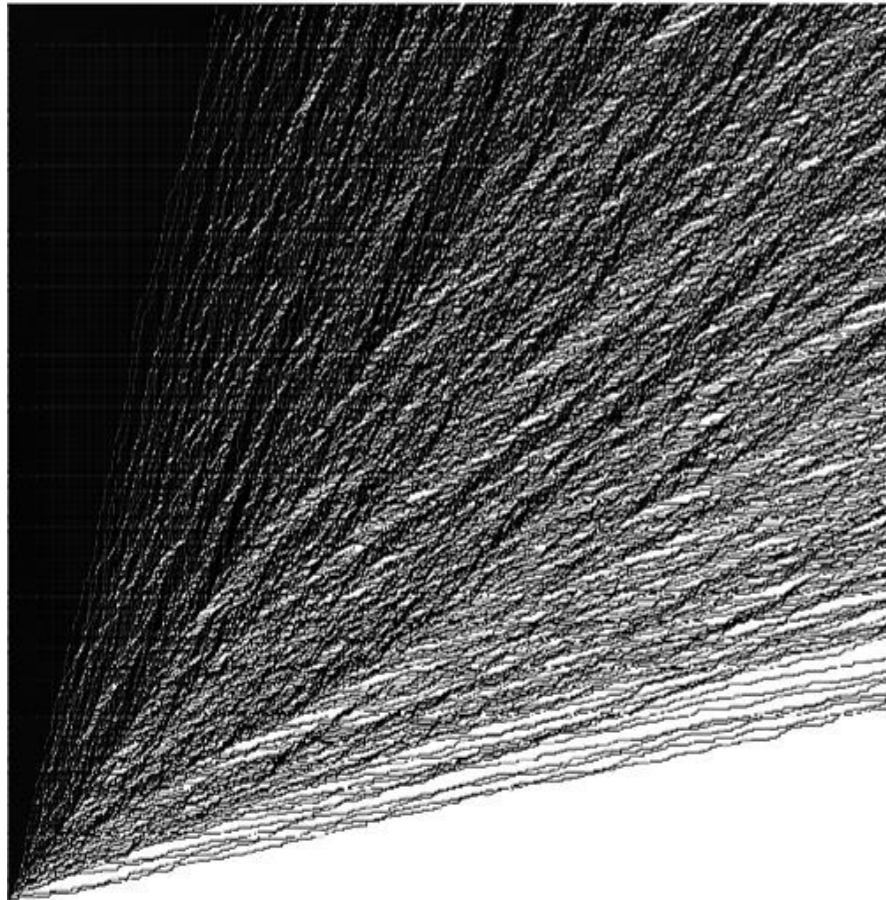
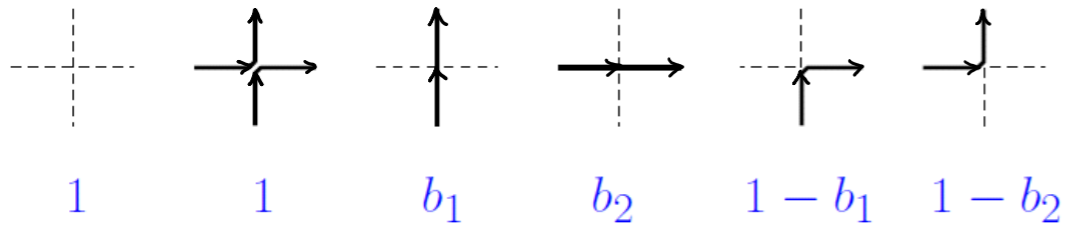
[Gwa-Spohn 1992]:

The stochastic six vertex model is a member of the KPZ universality class. This class was related to TW distributions in the late 1990's.

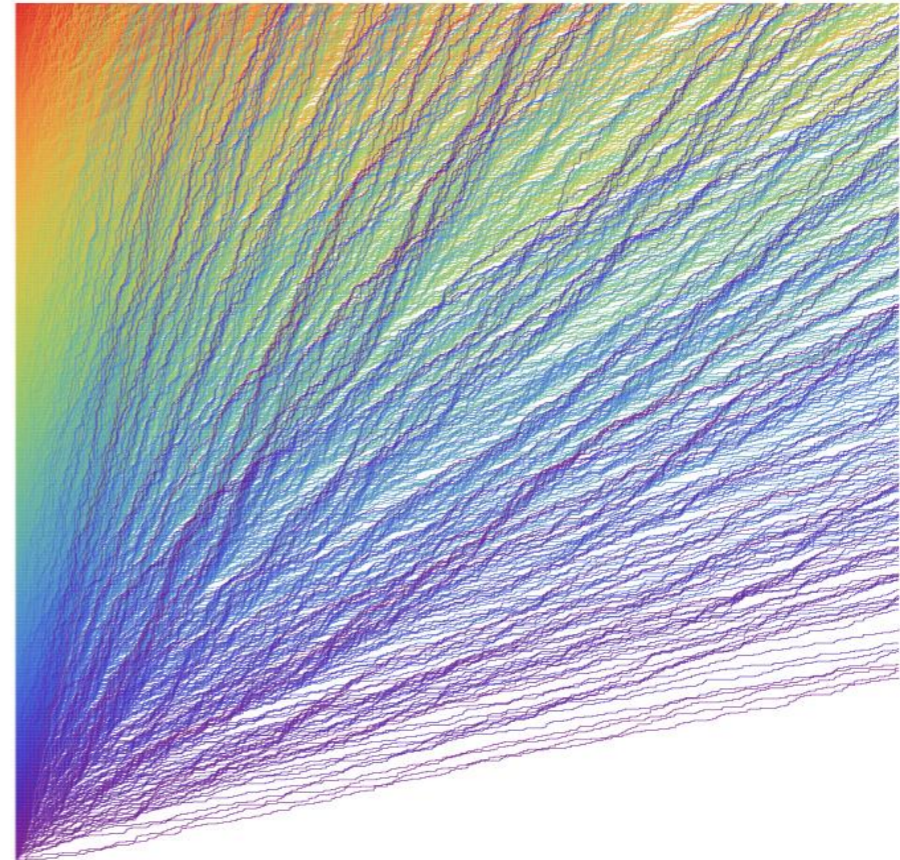
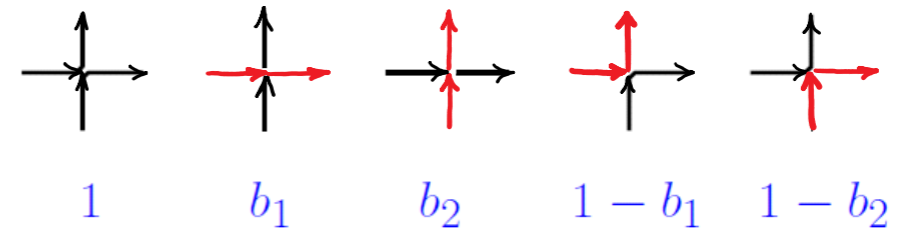


# Colored (higher rank) models

Stochastic six vertex model



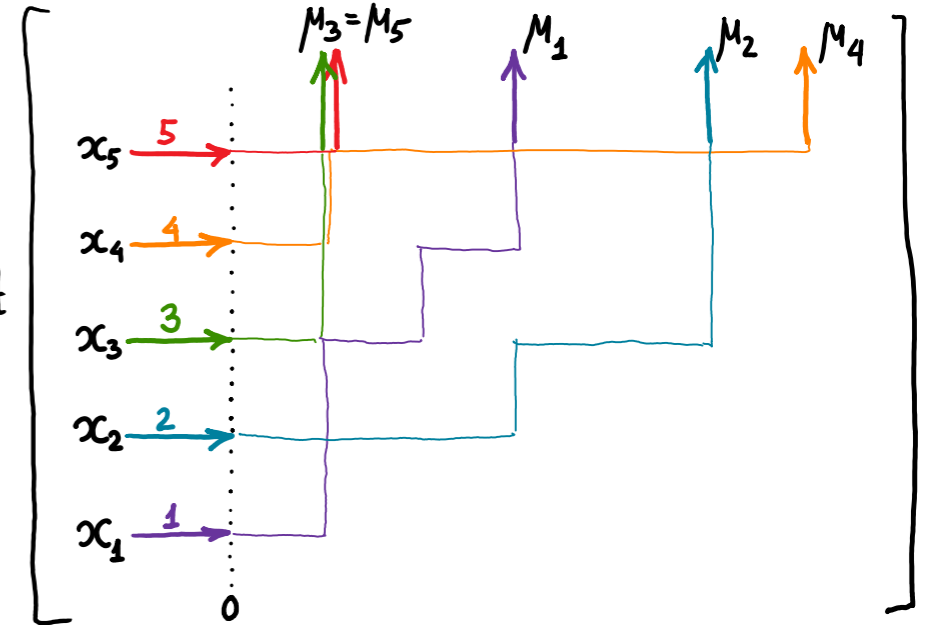
Colored stochastic six vertex model



# Nonsymmetric spin HL functions

$  \begin{array}{c}  I \\  \uparrow \\  0 \text{ --- } \rightarrow 0 \\  \downarrow \\  I  \end{array}  $ $\frac{1 - sxq^{I[1,n]}}{1 - sx}$	$  \begin{array}{c}  I \\  \uparrow \\  i \text{ --- } \rightarrow i \\  \downarrow \\  I  \end{array}  $ $\frac{(x - sq^{I_i})q^{I[i+1,n]}}{1 - sx}$	$  \begin{array}{c}  I_i^- \\  \uparrow \\  0 \text{ --- } \rightarrow i \\  \downarrow \\  I  \end{array}  $ $\frac{x(1 - q^{I_i})q^{I[i+1,n]}}{1 - sx}$
$  \begin{array}{c}  I_i^+ \\  \uparrow \\  i \text{ --- } \rightarrow 0 \\  \downarrow \\  I  \end{array}  $ $\frac{1 - s^2q^{I[1,n]}}{1 - sx}$	$  \begin{array}{c}  I_{ij}^{+-} \\  \uparrow \\  i \text{ --- } \rightarrow j \\  \downarrow \\  I  \end{array}  $ $\frac{x(1 - q^{I_j})q^{I[j+1,n]}}{1 - sx}$	$  \begin{array}{c}  I_{ji}^{+-} \\  \uparrow \\  j \text{ --- } \rightarrow i \\  \downarrow \\  I  \end{array}  $ $\frac{s(1 - q^{I_i})q^{I[i+1,n]}}{1 - sx}$

$$f_{\mu}(x_1, \dots, x_5) = \text{weight}$$



This is a complete basis of eigenfunctions of a transfer-matrix

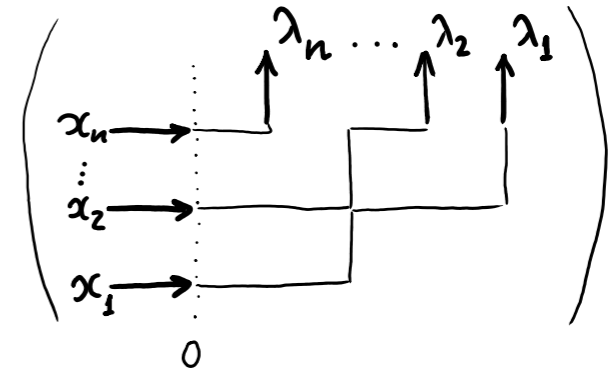
$$X_{\mu/\lambda}(u) = \text{weight}_u \left( \begin{array}{c} \lambda_3 \\ \lambda_5 \\ \lambda_1 \\ \lambda_2 \\ \lambda_4 \end{array} \right)$$

# Nonsymmetric spin HL functions

## Color-blindness

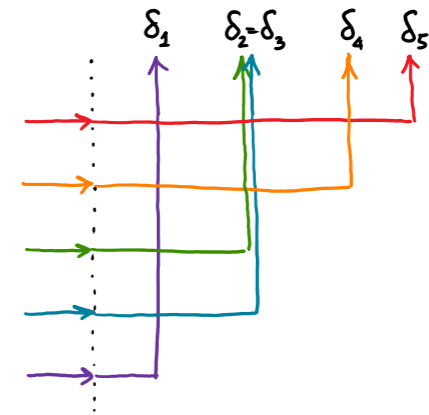
$$\sum_{\mu: \mu^+ = \lambda} f_{\mu}(x_1, \dots, x_n) = \frac{c(\lambda)}{c(0)} \cdot F_{\lambda}(x_1, \dots, x_n) \sim \text{weight}$$

( )<sup>+</sup> is ordering the coordinates



## Factorization for anti-dominant indices $\delta = (\delta_1 \leq \dots \leq \delta_n)$

$$f_{\delta}(x_1, \dots, x_n) = \frac{\prod_{j \geq 0} (s^2, q)_{m_j(\delta)}}{\prod_{i=1}^n (1 - s x_i)} \cdot \prod_{i=1}^n \left( \frac{x_i - s}{1 - s x_i} \right)^{\delta_i}$$



Unique path configuration

## AHA exchange relations

$$T_i f_{\mu}(x_1, \dots, x_n) = f_{s_i \circ \mu}(x_1, \dots, x_n)$$

for  $\mu_i < \mu_{i+1}$ ,  $s_i = (i \ i+1)$

$$T_i = q - \frac{x_i - q x_{i+1}}{x_i - x_{i+1}} (1 - s_i), \quad 1 \leq i \leq n-1$$

$$(T_i - q)(T_i + 1) = 0$$

# Nonsymmetric spin HL functions

Relation to off-shell nested Bethe vectors

Under the specialization

$$x_1^{(1)} = x_1^{(2)} = \dots = x_1^{(n)} = x_1$$

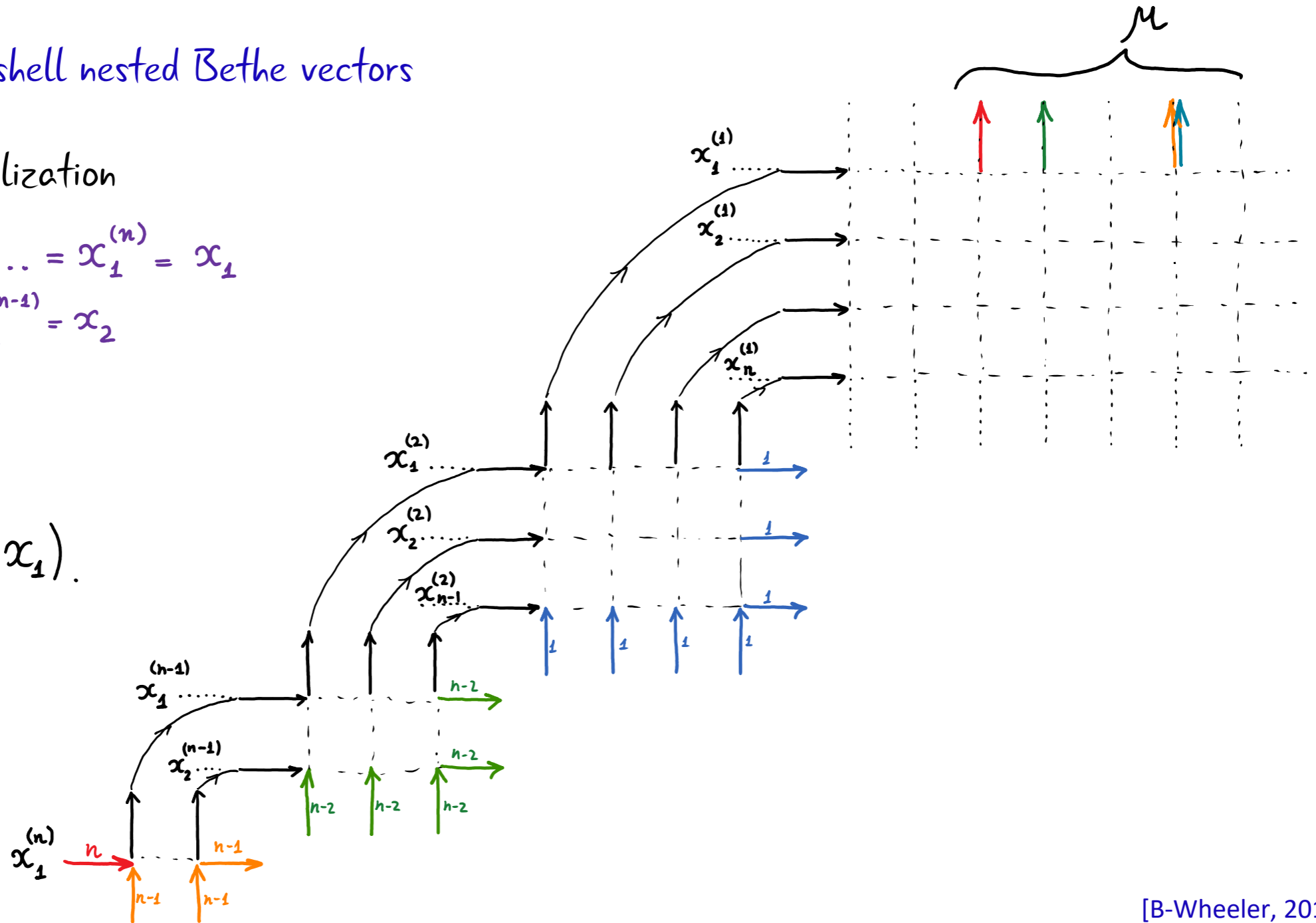
$$x_2^{(1)} \dots = x_2^{(n-1)} = x_2$$

$$\dots$$

$$x_n^{(1)} = x_n$$

one obtains

$$f_\mu(x_n, \dots, x_1).$$



# Nonsymmetric spin HL functions

Cauchy type summation identity

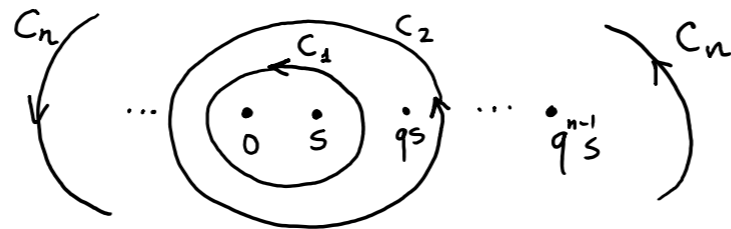
$$\sum_{\mu \geq 0} f_{\mu}(x_1, \dots, x_n) g_{\mu}^*(x_1, \dots, x_n) = \prod_{i=1}^n \frac{1}{1-x_i y_i} \cdot \prod_{i>j} \frac{1-q x_i y_j}{1-x_i y_j}$$

Orthogonality

$$g_{\mu}^*(x_1, \dots, x_n; s, q) = \text{const} \cdot f_{\tilde{\mu}}(x_n^{-1}, \dots, x_1^{-1}; s^{-1}, q^{-1})$$

$$\tilde{\mu} = (\mu_n, \dots, \mu_1)$$

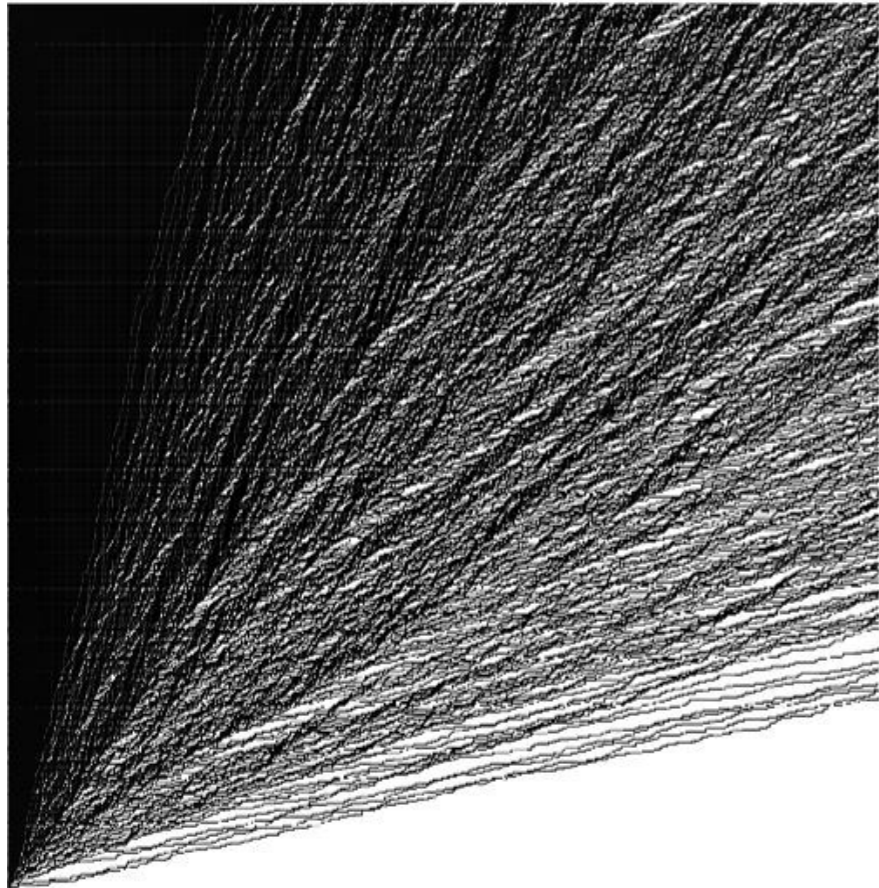
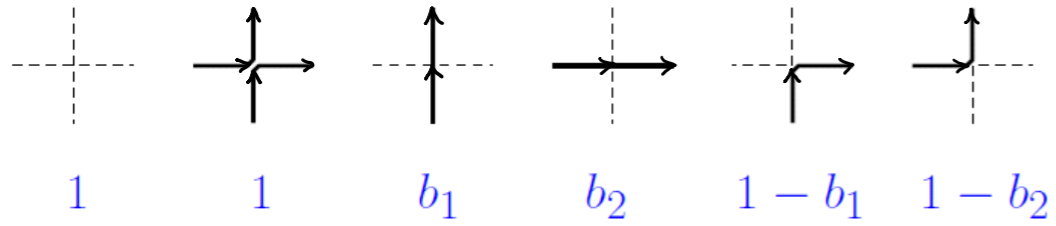
$$\frac{1}{(2\pi i)^n} \oint_{C_1} \dots \oint_{C_n} \prod_{i<j} \frac{x_j - x_i}{x_j - q x_i} f_{\mu}(x_1^{-1}, \dots, x_n^{-1}) g_{\nu}^*(x_1, \dots, x_n) dx = \mathbb{1}_{\mu=\nu}$$



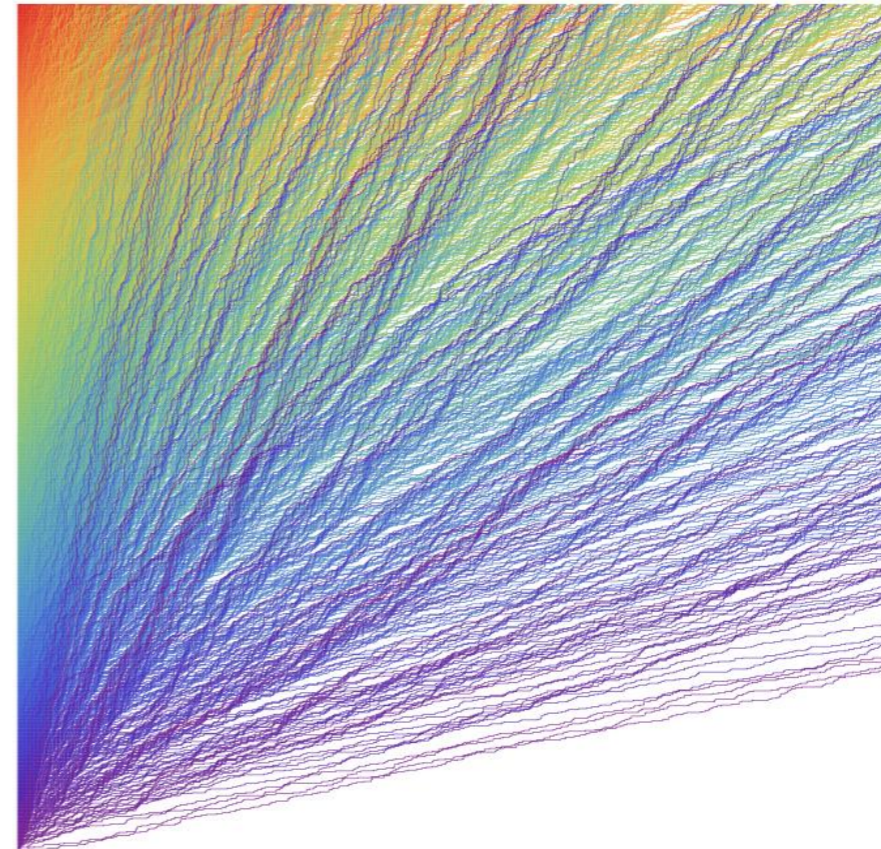
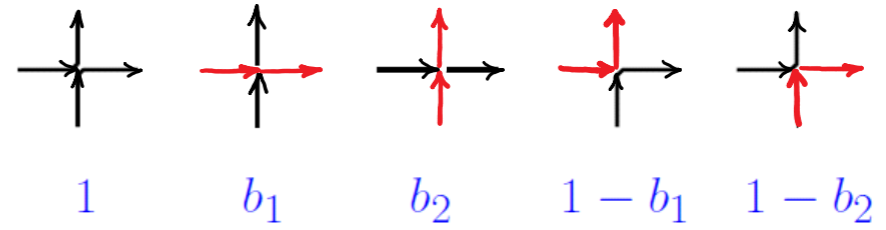


# A result about colored stochastic vertex models

Stochastic six vertex model

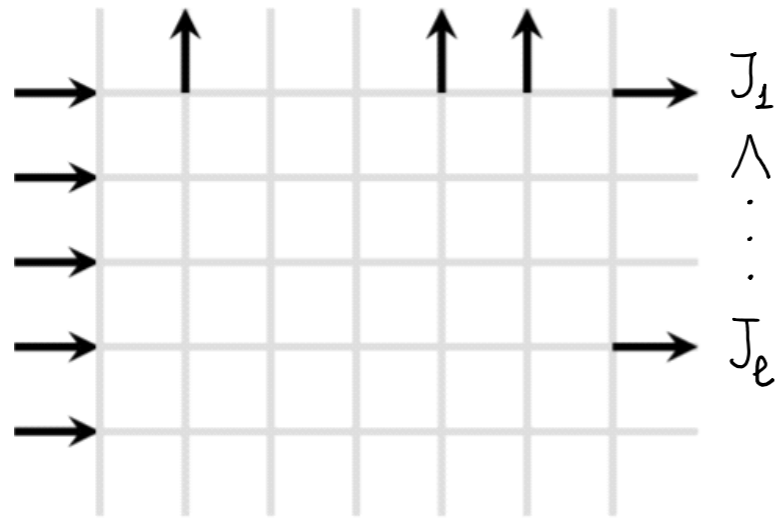
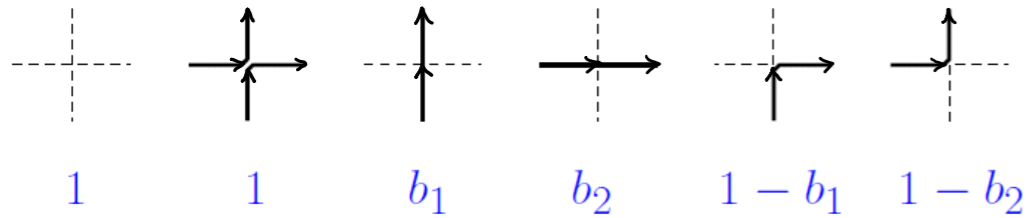


Colored stochastic six vertex model

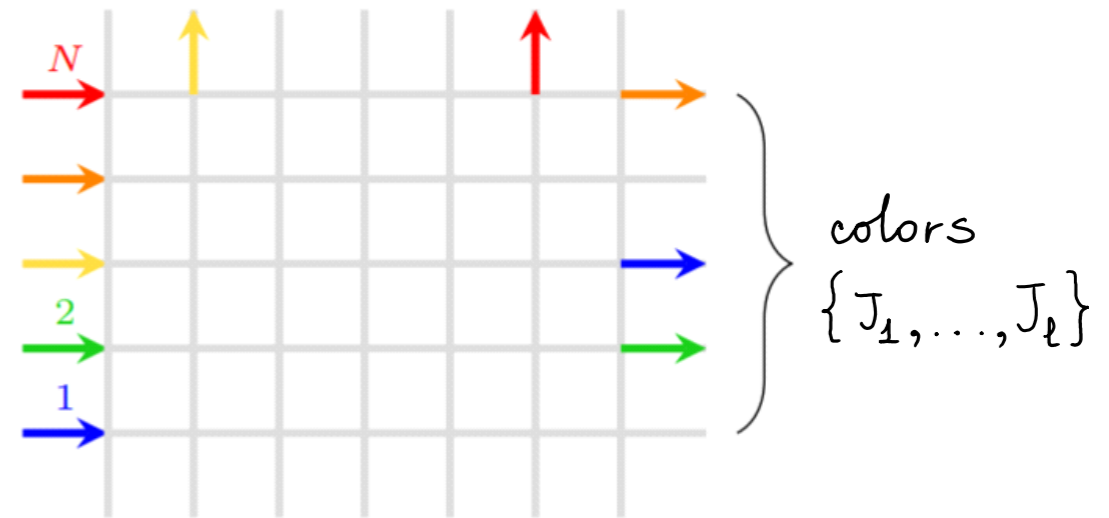
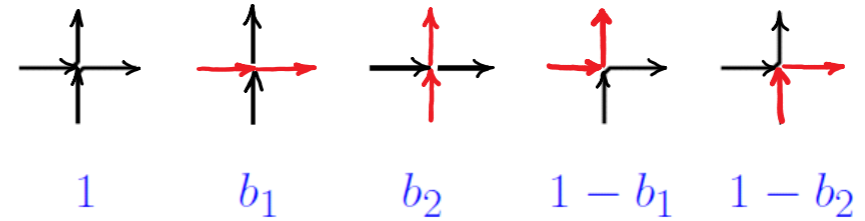


# A result about colored stochastic vertex models

Stochastic six vertex model



Colored stochastic six vertex model



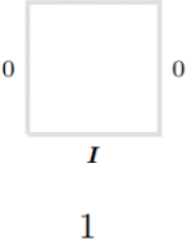
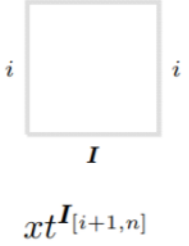
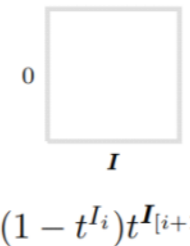
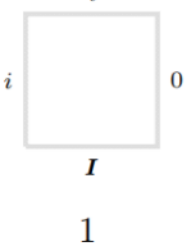
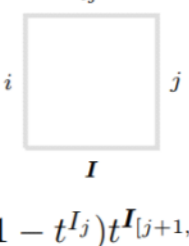

Theorem For any set  $\{J_1, \dots, J_\ell\}$  the following two probabilities coincide:

- (a) In the color-blind model, paths exit on the right exactly at those positions;
- (b) In the colored model, paths exiting on the right have exactly these colors.

Also works for inhomogeneous and fused models.

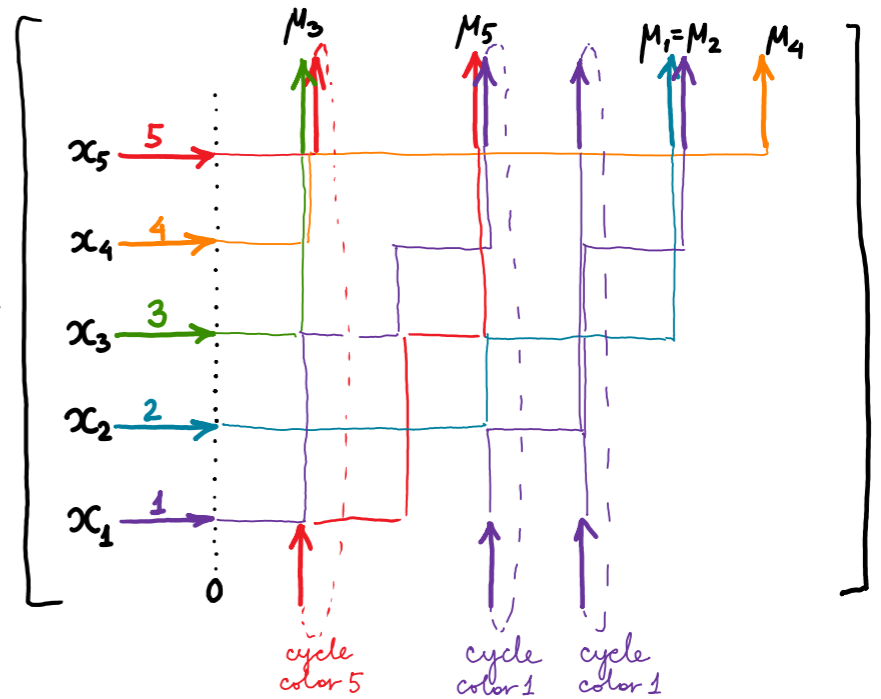
[B-Wheeler 2018]

# Nonsymmetric Macdonald polynomials

$I$  $1$	$I$  $x t^{I[i+1,n]}$	$I_i^-$  $x(1 - t^i) t^{I[i+1,n]}$
$I_i^+$  $1$	$I_{ij}^{+-}$  $x(1 - t^j) t^{I[j+1,n]}$	$I_{ji}^{+-}$  $0$

These are the same vertex weights with  $s=0$  and  $q$  replaced by  $t$ .

$$f_{\mu}(x_1, \dots, x_5) = \text{weight}$$



Theorem [B-Wheeler, 2019] If each cycle of color  $i$  at position  $j$  carries the additional factor of

$$q^{M_i-j} t^{-\#\{k < i : M_k > M_i\} + \#\{k > i : j \leq M_k < M_i\}}$$

then the partition function equals the nonsymmetric Macdonald polynomial indexed by  $\mu$ , up to an explicit multiplicative constant.