

Combinatorics of the ASEP on a ring and Macdonald polynomials

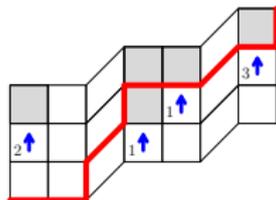
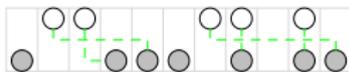
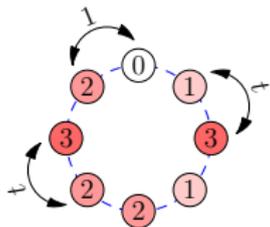
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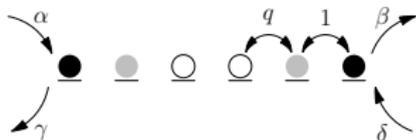
based on joint work with Sylvie Corteel (Paris Diderot) and Lauren Williams (UC Berkeley)

April 19, 2018

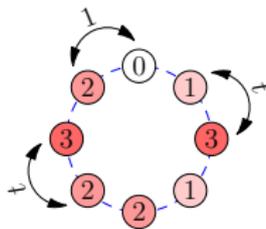


overview: asymmetric simple exclusion process (ASEP) and orthogonal polynomials

ASEP with open boundaries

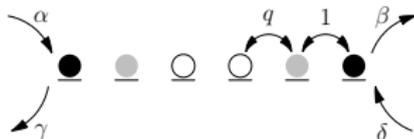


ASEP on a ring



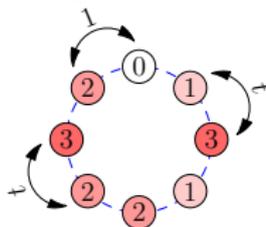
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Koornwinder-Macdonald polynomials
(Corteel-Williams, Cantini, 2015)

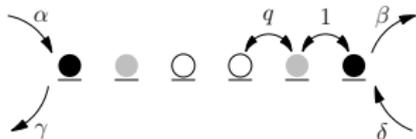
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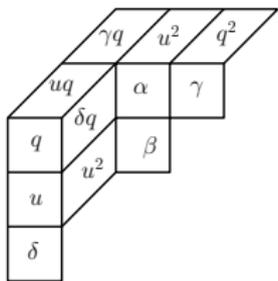
Macdonald polynomials (type A)
(Prohac-Evans-Mallick 2009)

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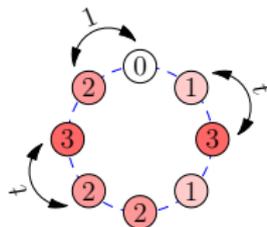


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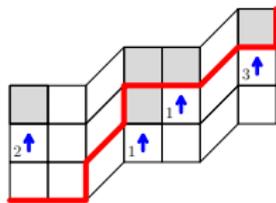


- (Corteel-M.-Williams, 2016)

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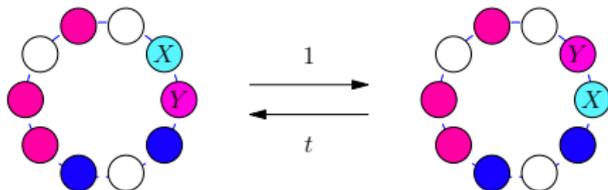
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2-species ASEP on a ring

- fix a ring with n sites and k, r, ℓ such that $n = k + r + \ell$.
- ASEP(k, r, ℓ) is a Markov chain on states $X \in \{2, 1, 0\}^n$ with k 2's, r 1's, and ℓ 0's
- possible transitions are swaps of adjacent particles:



with rate 1 if particle X has a bigger label than particle Y , and rate t otherwise. ($0 \leq t \leq 1$)

- we wish to compute the steady state probabilities $\Pr(X)$ of states of the ASEP

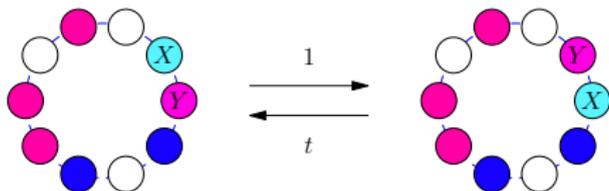
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$$\Pr(2, 1, 0) = \Pr(1, 0, 2) = \Pr(0, 2, 1) = \frac{1}{Z}(1 + 2t)$$

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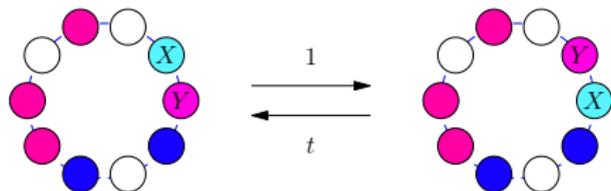
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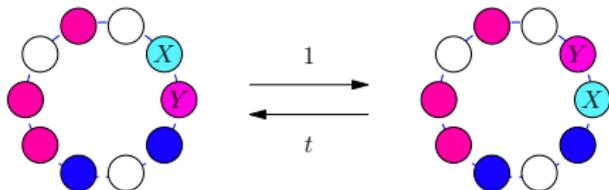
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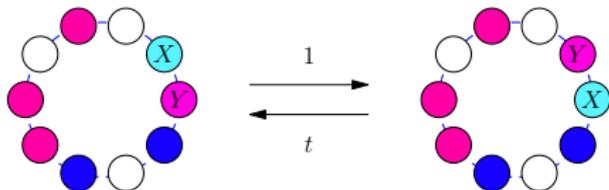
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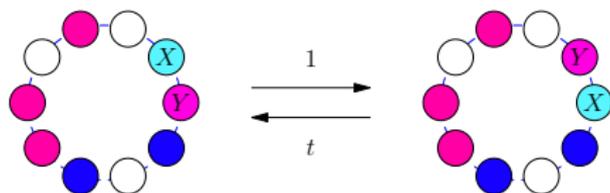
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Matrix Ansatz: 1-species ASEP with open boundaries

Theorem (Derrida-Evans-Hakim-Pasquier 1993)

Let $X = X_1 \dots X_n$, with $X_i \in \{2, 0\}$, be a word representing a state of the 1-species ASEP. If there exist matrices D and E and vectors $\langle w|$ and $|v\rangle$, that satisfy:

$$DE = tED + D + E,$$

$$\langle w|E = \frac{1}{\alpha}\langle w|, \quad D|v\rangle = \frac{1}{\beta}|v\rangle,$$

then

$$\Pr(X) = \frac{1}{Z_n} \langle w| \text{Mat}(X) |v\rangle$$

where $\text{Mat}(X) = \text{Mat}(X_1) \dots \text{Mat}(X_n)$ is a map defined by $\text{Mat}(2) = D$, $\text{Mat}(0) = E$, and $Z_n = \langle w|(D + E)^n|v\rangle$.

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$\Pr(X)$ can be expressed in terms of a product of non-commuting operators D and E , where $2 \mapsto D$, $1 \mapsto A$, and $0 \mapsto E$

Example:

$$\Pr(200020) = \frac{1}{Z_6} \langle w|DEEEDE|v\rangle$$

combinatorial interpretation of the quadratic algebra

$$DE = qED + D + E$$

Given a word in $\text{Mat}(X)$ in D 's and E 's, we can apply the relation $DE = qED + D + E$ to get

$$\text{Mat}(X) = \sum_{i,j} c_{i,j}(t) E^i D^j$$

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For example, let $X = 2202$. Then

$$\begin{aligned} \text{Mat}(X) &= DDED = D(tED + D + E)D \\ &= t(tED + D + E)DD + DDD + (tED + D + E)D \\ &= t^2EDDD + (t + 1)DDD + 2tEDD + DD + ED. \end{aligned}$$

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Define a Q-tableau from this algebra by assigning D to vertical edge and E to horizontal edge and applying local edge rewriting rules. (due to X. Viennot)

2-species ASEP generalization: Matrix Ansatz

For ASEP with particles of type 2, 1, 0, add third operator A to get the quadratic algebra (Uchiyama '09):

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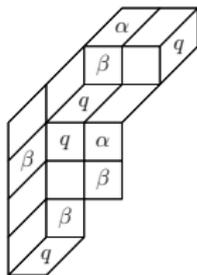
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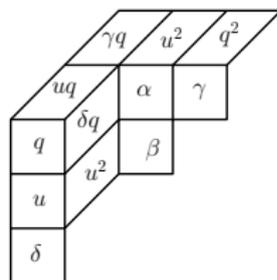
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Generalization of the alternative tableaux:



(M.-Viennot '15)



(Corteel-M.-Williams '16)

Get formula for Koornwinder-Macdonald polynomials (for partitions consisting of a single row/column) at $q = t$.

2-ASEP on a ring

Matrix ansatz (DEHP):

$$DE = tED + (1 - t)D + E,$$

$$DA = tAD + (1 - t)A,$$

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then

$$\Pr(X) = \frac{1}{Z_{(k,r,\ell)}} \text{tr}(\text{Mat}(X))$$

where $\text{Mat} : \{2, 1, 0\} \mapsto \{D, A, E\}$ and X has k 2's, r 1's, and ℓ 0's.

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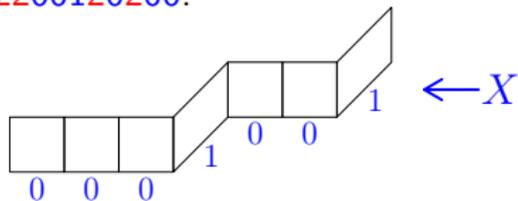
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we will split the infinite tableaux into classes and define finite **cylindric tableaux**, with each tableau representing a class and having a weight which is a generating function (instead of a polynomial)

cylindric rhombic tableaux (CRT)

Let $X \in ASEP(k, r, \ell)$.

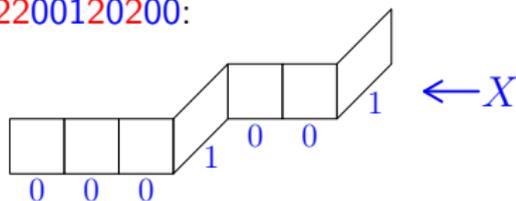
An X -strip is a strip of squares and rhombi corresponding to the 0,1-sub-word of X . For example, $X = 12200120200$:



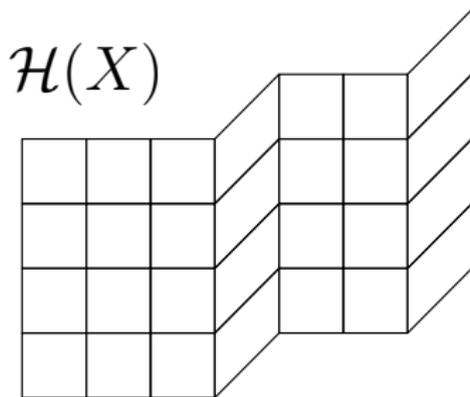
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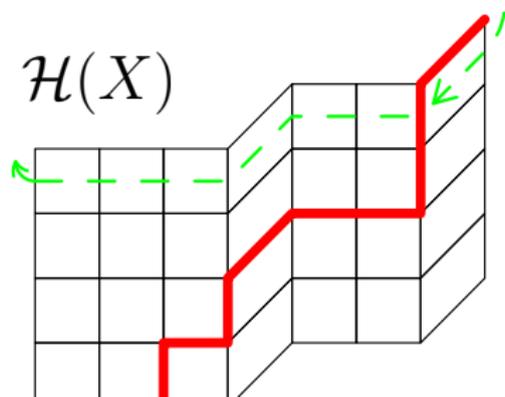


Join k X -strips to obtain the shape $\mathcal{H}(X)$ and superimpose a path $P(X)$:



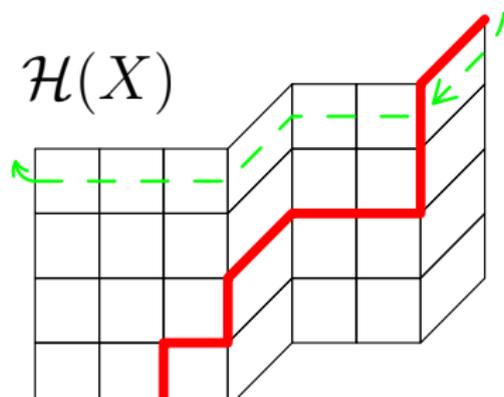
cylindric rhombic tableaux

- Identify the left and right boundaries of $\mathcal{H}(X)$, making it a cylinder.
- Horizontal strips start from $P(X)$ and wrap around to the left.
- An up-arrow can be in a square tile, killing the cells above it. (So, there can be at most one per column.)
- at $t = 0$, these tableaux are in bijection with **multiline queues**, which are an object due to Ferrari-Martin from 2005 that has been used to study the m-TASEP (M. 2018)
- choose any placement of up-arrows and some order σ assigned to the arrows in each row.



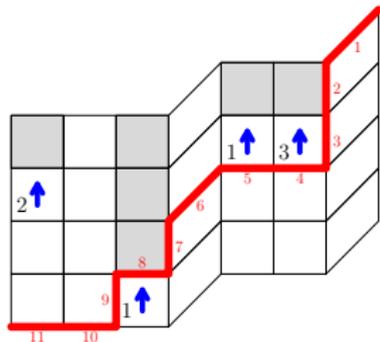
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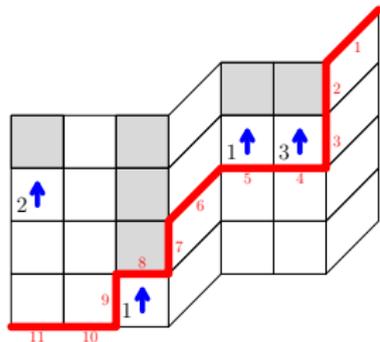
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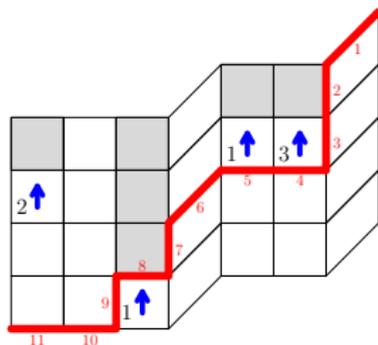


cylindric rhombic tableaux

- choose any placement of up-arrows and some order σ assigned to the arrows in each row.
- t keeps track of how many available locations are “skipped” with a **disorder** statistic
- Example:

$$\text{wt}(T) = \frac{t^{10}}{[7]_t [6]_t [5]_t [4]_t},$$

where $[j]_t = (1 - t^j)/(1 - t)$.

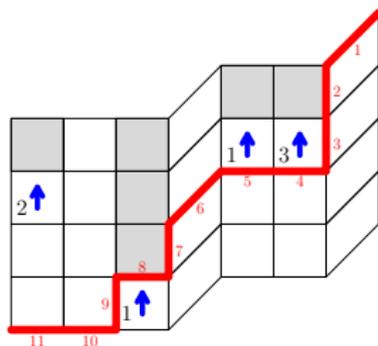


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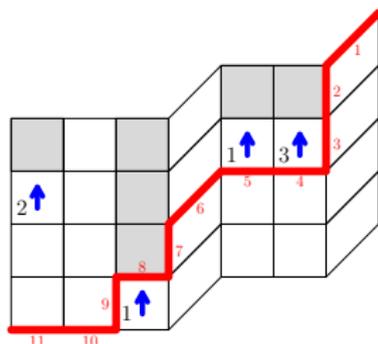


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probabilities of the 2-ASEP on a ring

Theorem (Corteel-M.-Williams 2018)

Let $X \in \text{States}(k, r, \ell)$ be a state of the 2-ASEP on a ring. Then

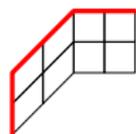
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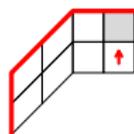
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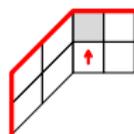
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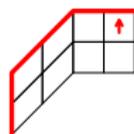
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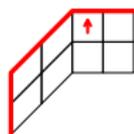
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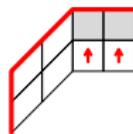
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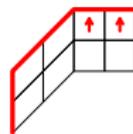
$\frac{1}{[4]}$



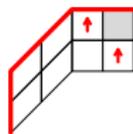
$\frac{t}{[4]}$



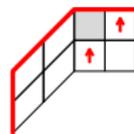
$\frac{(1+t^3)}{[4][3]}$



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$\frac{1}{[4][3]}$



$\frac{t}{[4][3]}$

Example: $\pi(001122) = \frac{1}{Z_{(2,2,2)} [4]_t [3]_t} (6 + t + 6t^2 + t^3 + t^4)$

Nonsymmetric Macdonald polynomials

- symmetric Macdonald polynomials $P_\lambda(\mathbf{x}; q, t)$ are q, t generalizations of Schur polynomials. Can be defined via symmetrization of *nonsymmetric Macdonald polynomials* E_μ .
- E_μ introduced in 1995 (Opdam, Cherednik), can be computed from Yang-Baxter graphs, with combinatorial formula due to Haiman-Haglund-Loehr '05
- recent work of Cantini-de Gier-Wheeler computes matrix product for polynomials $f_\mu(\mathbf{x}; q, t)$ which are solutions to a certain affine Hecke algebra.
- $f_\mu(1, \dots, 1; 1, t)$ is the steady state probability of state μ of the ASEP on a ring.
- f_μ are related to the E_λ 's:

$$E_\lambda = \sum_{\mu \geq \lambda} a_{\lambda, \mu}(q, t) f_{\mu^c}$$

and the P_λ 's, summed over permutations μ of the parts in λ :

$$P_\lambda(\mathbf{x}; q, t) = \sum_{\mu} f_{\mu}(\mathbf{x}; q, t).$$

(in particular, when λ is a partition, we have $f_{\lambda^c} = E_\lambda$.)

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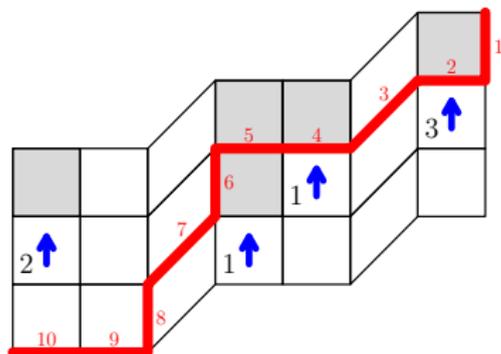
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enhanced weights on CRT to compute the f_μ

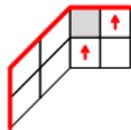
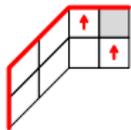
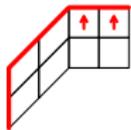
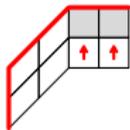
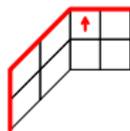
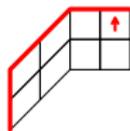
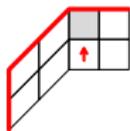
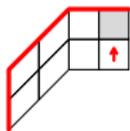
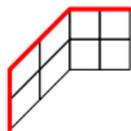
To a CRT with arrow placement and order (T, σ) , we can assign a weight $\text{wt}_{qtx}(T, \sigma)(\mathbf{x}; q, t)$ as follows:

- t 's count the disorder statistic of (T, σ) , and $[j]_{qt} = (1 - qt^j)/(1 - t)$
- associate x_i 's to the edges of $P(X)$ and have them keep track of which strips contain the *last arrow* in a given row according to σ
- have q count the number of times we “cross the boundary” of the CRT to traverse the arrows in order σ

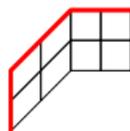


$$\text{wt}_{qtx}(T, \sigma) = \frac{t^{12} q^3}{[7]_{qt}^{(4)}!} x_2 x_3 x_4^2 x_5 x_6 x_7 x_8 x_9^2 x_{10}^2$$

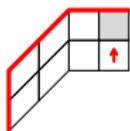
example for $X = 001122$



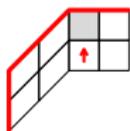
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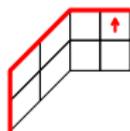
$$x_1^2 x_2^2 x_3 x_4$$



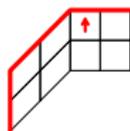
$$\frac{q x_1 x_2^2 x_3 x_4 x_6}{[4]_{qt}}$$



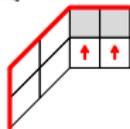
$$\frac{q t x_1^2 x_2 x_3 x_4 x_6}{[4]_{qt}}$$



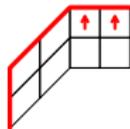
$$\frac{q x_1 x_2^2 x_3 x_4 x_5}{[4]_{qt}}$$



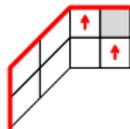
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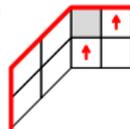
$$\frac{(q x_1 + q^2 t^3 x_2) x_1 x_2 x_3 x_4 x_6}{[4]_{qt} [3]_{qt}}$$



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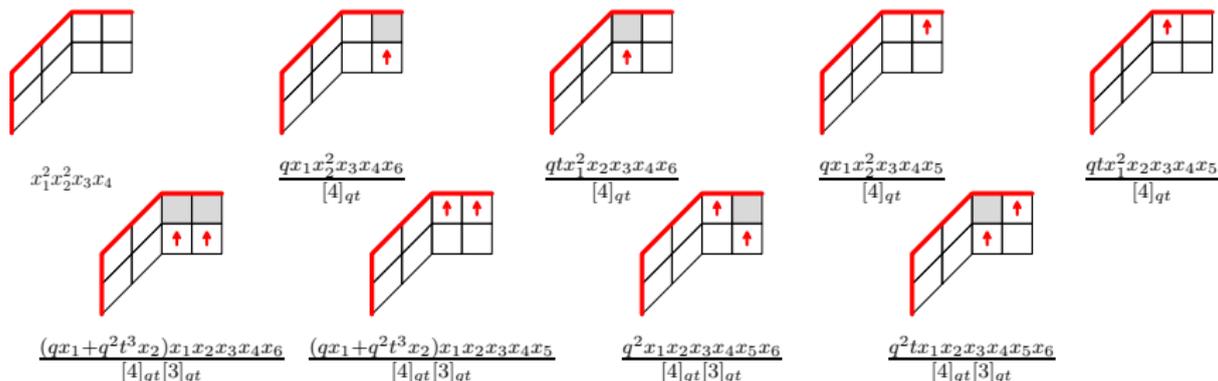
$$\frac{q^2 x_1 x_2 x_3 x_4 x_5 x_6}{[4]_{qt} [3]_{qt}}$$



$$\frac{q^2 t x_1 x_2 x_3 x_4 x_5 x_6}{[4]_{qt} [3]_{qt}}$$

$$f_{(001122)} = x_1^2 x_2^2 x_3 x_4 + \frac{q(t x_1 + x_2)(x_5 + x_6) x_1 x_2 x_3 x_4}{[4]_{qt}} + \frac{q(x_1 + q t^3 x_2)(x_5 + x_6) x_1 x_2 x_3 x_4}{[4]_{qt} [3]_{qt}} + \frac{q^2(1+t) x_1 x_2 x_3 x_4 x_5 x_6}{[4]_{qt} [3]_{qt}}$$

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In particular, when λ is a partition with parts of size at most 2, we obtain:

Theorem (Corteel-M.-Williams '18)

$$E_\lambda = \sum_{(T, \sigma) \in \text{CRT}(\lambda^C)} \text{wt}_{qt x} (T, \sigma).$$

e.g. nonsymmetric Macdonald poly. $E_{(2,2,1,1,0,0)}$ equals $f_{(0,0,1,1,2,2)}$ above.

summary

For 2-ASEP, we defined cylindric rhombic tableaux (CRT) with q, t, x weights that give formula $f_\mu(x_1, \dots, x_n; q, t)$ such that:

- $\Pr(\mu) = f_\mu(1, \dots, 1; 1, t)$ is the probability of the 2-ASEP on a ring
- $P_\lambda = \sum_\mu f(x_1, \dots, x_n; q, t)$ is the partition function over the CRT when λ has parts of size 0, 1, 2
- when μ is a partition, $f_{\mu^C}(x_1, \dots, x_n; q, t) = E_\mu$, the non-symmetric Macdonald polynomial (e.g. if $\mu = 211100$, then $\mu^C = 011122$).
- there is a bijection to *multiline queues* and a solution-in-progress for higher species.

Happy birthday Kolya!!!

