

Symmetric functions and interacting particle processes

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joint with [Arvind Ayyer](#) and [James Martin](#),
arXiv:2011.06117, arXiv:2209.09859

particle models and symmetric functions

- ① the **asymmetric simple exclusion process (ASEP)** → combinatorial formula for Macdonald polynomials and some nice specializations
- ② **modified Macdonald polynomials** → the multispecies totally asymmetric zero range process (mTAZRP) and observables

exactly solvable interacting particle models

- **integrable systems**: a class of dynamical systems with a certain restricted structure, in particular making them *solvable*

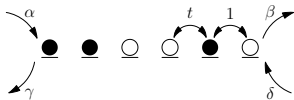
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- we are interested in studying integrable systems whose exact solutions (e.g. **stationary distributions**) can be expressed in terms of combinatorial formulas or special functions (e.g. **Macdonald polynomials**)

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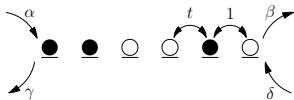
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- we are interested in studying integrable systems whose exact solutions (e.g. **stationary distributions**) can be expressed in terms of combinatorial formulas or special functions (e.g. **Macdonald polynomials**)
- the field was initiated by **Spitzer** in his 1970 paper where he defined the **ASEP** (**A**symmetric **S**imple **E**xclusion **P**rocess) and the **ZRP** (**Z**ero **R**ange **P**rocess)

Asymmetric Simple Exclusion Process (ASEP)



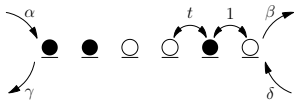
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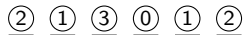
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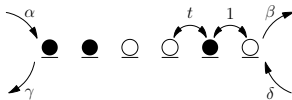


single species ASEP

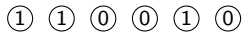


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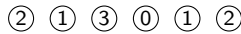
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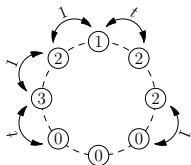


multispecies ASEP

- **dynamics:** any two adjacent particles may swap with some predetermined rate (in our case, fixed by a parameter $0 \leq t \leq 1$):



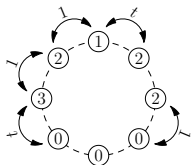
our setting: ASEP on a circle



$$n = 8, \quad \lambda = (3, 2, 2, 2, 1, 0, 0, 0)$$

- Fix a circular lattice on n sites, and choose n nonnegative integer weights. **Collection of weights: a vector $\lambda = (\lambda_1 \geq \dots \geq \lambda_n \geq 0)$.**

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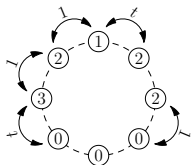


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$$\alpha = (1, 2, 2, 0, 0, 0, 3, 2) \in \text{ASEP}(\lambda)$$

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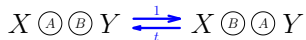
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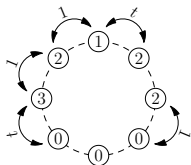
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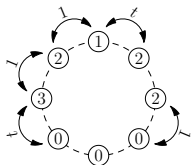
- For example, ASEP($(2, 2, 1, 0)$) has 12 states:

$$(2, 2, 1, 0), (2, 1, 2, 0), (2, 1, 0, 2), (2, 2, 0, 1), (2, 0, 2, 1), (2, 0, 1, 2), (0, 2, 2, 1), \dots$$

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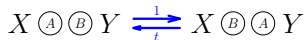
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- Goal: compute the stationary probabilities**

Example for $\lambda = (2, 1)$ and $n = 3$

- the elements in the state space are:

② ① ① ② ① ① ① ② ① ① ① ② ① ② ① ① ① ②

- the (row stochastic) transition matrix is:

$$\begin{pmatrix} 1 - \frac{2+t}{3} & \frac{1}{3} & \frac{1}{3} & 0 & 0 & \frac{t}{3} \\ \frac{t}{3} & 1 - \frac{1+2t}{3} & 0 & \frac{t}{3} & \frac{1}{3} & 0 \\ \frac{t}{3} & 0 & 1 - \frac{1+2t}{3} & \frac{1}{3} & \frac{t}{3} & 0 \\ 0 & \frac{1}{3} & \frac{t}{3} & 1 - \frac{2+t}{3} & 0 & \frac{1}{3} \\ 0 & \frac{t}{3} & \frac{1}{3} & 0 & 1 - \frac{2+t}{3} & \frac{1}{3} \\ \frac{1}{3} & 0 & \frac{t}{3} & 0 & \frac{t}{3} & 1 - \frac{1+2t}{3} \end{pmatrix}$$

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- the (unnormalized) **stationary distribution** is:

$$\tilde{\text{Pr}}((2, 1, 0)) = \tilde{\text{Pr}}((1, 0, 2)) = \tilde{\text{Pr}}((0, 2, 1)) = 2 + t$$

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From ASEP to Macdonald polynomials

Define the **partition function** of ASEP(λ, n):

$$\mathcal{Z}_{\lambda, n}(t) = \sum_{\alpha \in \mathcal{S}_n \cdot \lambda} \tilde{\text{Pr}}(\alpha)(t).$$

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Theorem (Cantini–de Gier–Wheeler '15)

*The partition function of ASEP(λ, n) is a specialization of the **Macdonald polynomial**:*

$$P_{\lambda}(1, \dots, 1; 1, t) = \mathcal{Z}_{\lambda, n}(t)$$

Symmetric functions

- Let $X = x_1, x_2, \dots$ be a family of indeterminates, and let $\Lambda = \Lambda_{\mathbb{Q}}$ be the algebra of symmetric functions in X over \mathbb{Q}
 - $f(x_1, \dots, x_n) \in \Lambda$ is **symmetric** if $\forall \pi \in S_n, f(x_1, \dots, x_n) = f(x_{\pi(1)}, \dots, x_{\pi(n)})$

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- Λ has several nice bases: e.g. $\{m_\lambda\}, \{e_\lambda\}, \{h_\lambda\}, \{p_\lambda\}$, indexed by partitions λ .
E.g. $m_{(2,1)} = \sum_{i,j} x_i^2 x_j^1 = x_1^2 x_2 + x_1^2 x_3 + x_2^2 x_3 + x_1 x_2^2 + x_1 x_3^2 + x_2 x_3^2 + \dots$

Let \langle, \rangle be the **standard inner product** on Λ . $\{s_\lambda\}$ is the unique basis of Λ :

- orthogonal with respect to \langle, \rangle
- upper triangular with respect to $\{m_\lambda\}$:

$$s_\lambda = m_\lambda + \sum_{\mu < \lambda} c_{\mu, \lambda} m_\mu$$

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- $s_{\lambda} = \sum_{\sigma} x^{\sigma}$ where σ 's are **semi-standard fillings** of the Young diagram of shape λ

E.g. $s_{(2,1)} = x_1^2 x_2 + x_1^2 x_3 + x_1 x_2^2 + x_1 x_2 x_3 + x_1 x_2 x_3 + x_1 x_3^2 + x_2^2 x_3 + x_2 x_3^2 = m_{(2,1)} + m_{(1,1,1)}$



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Let $\langle \cdot, \cdot \rangle_{q,t}$ be the inner product on $\Lambda(q, t)$. Then $\{P_\lambda\}$ is the unique basis of $\Lambda(q, t)$ that is uniquely determined by:

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- **Example:**

$$P_{(2,1)}(X; q, t) = m_{(2,1)} + \frac{(1-t)(2+q+t+2qt)}{1-qt^2} m_{(1,1,1)}.$$

Combinatorial formulas

- Haglund-Haiman-Loehr '04 gave a formula for P_λ as a sum over tableaux with statistics **maj** and **(co)inv**:

$$P_\lambda(X; q, t) = \sum_{\substack{\sigma \in \text{dg}(\lambda) \\ \sigma \text{ non-attacking}}} x^\sigma q^{\text{maj}(\sigma)} t^{\text{coinv}(\sigma)} \prod_u \frac{1-t}{1-q^{\text{leg}(u)+1} t^{\text{arm}(u)+1}}$$

Combinatorial formulas

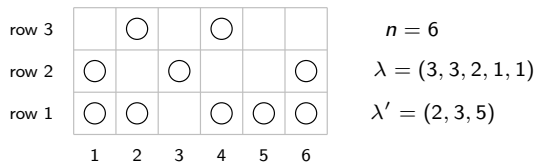
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- Corteel-M-Williams '18: a new formula for P_λ in terms of **multiline queues**, which also give formulas for the **stationary distribution of the ASEP**; this was inspired by the result of Cantini-de Gier-Wheeler '15

multiline queues

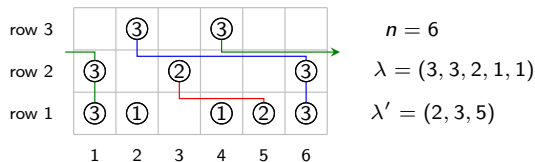
- a **multiline queue** (MLQ) of type λ , n is an **arrangement** and **pairing** of balls on a $n \times \lambda_1$ lattice, with λ'_j balls in **row j** .



Angel '08, Ferrari-Martin '07 ($t = 0$ case), Martin '18 (for $q = x_1 = \dots = x_n = 1$),
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multiline queues

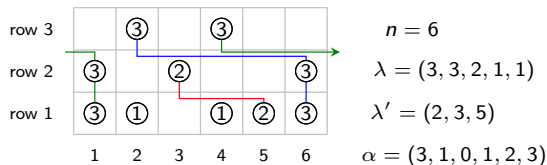
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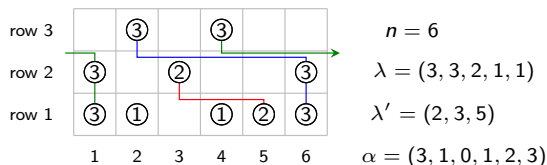


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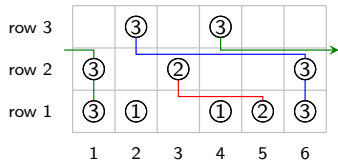
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- The weight $\text{wt}(M)$ of a multiline queue depends on the parameters t, q, x_1, \dots, x_n :

$$\text{weight} = x_1^2 x_2^2 x_3 x_4^2 x_5 x_6^2 q t^2 \frac{(1-t)^3}{(1-qt^3)^2(1-qt^2)}$$

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$$n = 6$$

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

$$\lambda' = (2, 3, 5)$$

$$\alpha = (3, 1, 0, 1, 2, 3)$$

2	4				
6	1	3			
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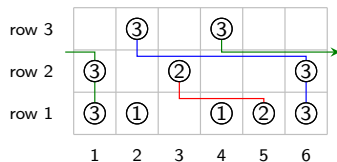
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- Can be represented by a **tableau**, where each **string** is mapped to a **column**

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- The **state** of a multiline queue is read off the bottom row
- The weight $\text{wt}(M)$ of a multiline queue depends on the parameters t, q, x_1, \dots, x_n :

$$\text{weight} = x_1^2 x_2^2 x_3 x_4^2 x_5 x_6^2 q t^2 \frac{(1-t)^3}{(1-qt^3)^2(1-qt^2)}$$

- Can be represented by a **tableau**, where each **string** is mapped to a **column**
- Can be represented by a queueing system and described as a **coupled system of 1-ASEPs**.
The pairing is a **projection map** onto the n -ASEP.

From ASEP to Macdonald polynomials

Theorem (Martin '18, Corteel-M-Williams '18)

The (unnormalized) *stationary probability* of state α of the m ASEP is

$$\tilde{\text{Pr}}(\alpha)(\mathbf{t}) = \sum_{M \in \text{MLQ}(\alpha)} \text{wt}(M)(1, \dots, 1; \mathbf{1}, \mathbf{t})$$

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The *partition function* of $\text{ASEP}(\lambda, n)$ is a specialization of the *Macdonald polynomial*:

$$P_\lambda(1, \dots, 1; 1, \mathbf{t}) = \mathcal{Z}_{\lambda, n}(\mathbf{t}) = \sum_{\alpha \in \mathcal{S}_n \cdot \lambda} \tilde{\text{Pr}}(\alpha)(\mathbf{t}).$$

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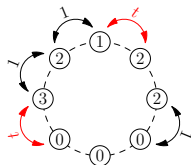
Theorem (Corteel-M-Williams '18)

$$P_\lambda(x_1, \dots, x_n; q, t) = \sum_{M \in \text{MLQ}(\lambda, n)} \text{wt}(M)(x_1, \dots, x_n; q, t)$$

(also Lenart '09 for λ with distinct parts.)

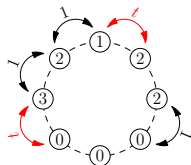
$t = 0$: Hall-Littlewood polynomials $P_\lambda(X; q, 0)$ via multiline queues

- at $t = 0$, the ASEP becomes the TASEP:



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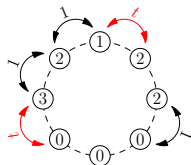
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- Define the Hall-Littlewood polynomial $P_\lambda(X; q) = P_\lambda(X; q, 0)$
- In the multiline queues, **pairings** become **forced**.

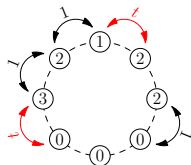
	○		○		
○		○			○
○	○		○	○	○
1	2	3	4	5	6

$$n = 6$$

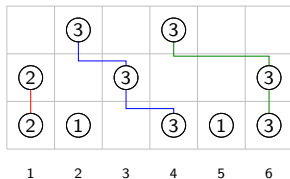
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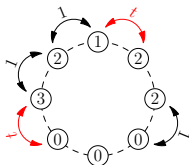
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	③		③		
②		③			③
②	①		③	①	③
1	2	3	4	5	6

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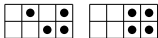
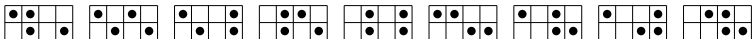
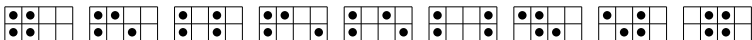
$$\alpha = (2, 1, 0, 3, 1, 3)$$

- call $MLQ(\lambda, n)$ the set of ball arrangements with λ_j^i balls in each row j , on a lattice of size $n \times \lambda_1$. (The labels can be recovered uniquely)

$$M = (\{1, 2, 4, 5, 6\}, \{1, 3, 6\}, \{2, 4\})$$

$q = t = 0$: Schur polynomials via multiline queues

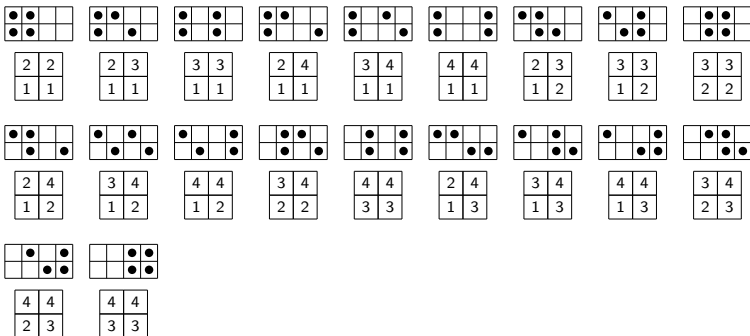
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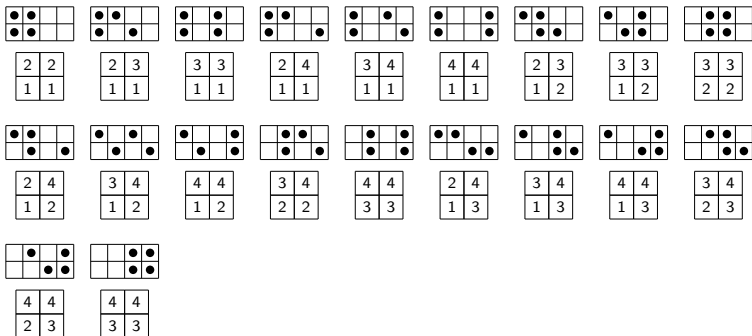
$$\sum_{M \in MLQ_0(\lambda, n)} x^M = s_\lambda$$



$q = t = 0$: Schur polynomials via multiline queues

- at $q = 0$, the multiline queues are **non-wrapping**, denote this set by $\text{MLQ}_0(\lambda, n)$:

$$\sum_{M \in \text{MLQ}_0(\lambda, n)} x^M = s_\lambda$$



- the map $\text{MLQ}_0(\lambda) \rightarrow \text{SSYT}(\lambda)$ is given by **column RSK** applied to the **row reading word** of the multiline queue. (bottom to top, left to right)

Lascoux–Schützenberger charge formula via MLQs

- For a permutation σ , define $\text{charge}(\sigma) := \text{maj}(\text{rev}(\sigma^{-1}))$. For a SSYT τ , $\text{charge}(\tau) = \text{charge}(\text{rw}(\tau))$.

Theorem (Lascoux–Schützenberger, '78)

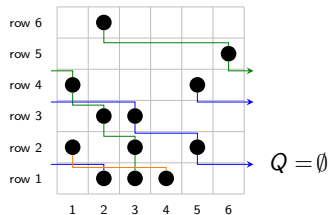
$$P_\lambda(X; q, 0) = \sum_{\mu \leq \lambda} K_{\mu' \lambda'}(q) s_\mu, \quad K_{\lambda \mu}(q) = \sum_{Q \in \text{SSYT}(\lambda, \mu)} q^{\text{charge}(Q)}$$

- Define **collapsing procedure** ρ (with Jerónimo Valencia '23+):

$$\begin{aligned} \text{MLQ}(\lambda) &\longrightarrow \bigcup_{\mu} \text{MLQ}_0(\mu) \times \text{SSYT}(\mu', \lambda') \\ M &\longrightarrow (M_0, Q) \end{aligned}$$

- $\text{charge}(Q) = \text{maj}(M)$ (the q -statistic, keeps track of wrapping pairings)
- can be described using **lowering operators** on the **column reading word** of M
- **lifting procedure** ρ^{-1} can be described using **raising operators**
- generalizes to a **quasisymmetric refinement** of $K_{\lambda \mu}(q)$.

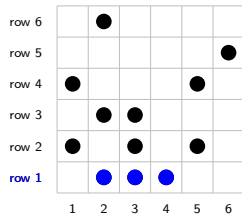
Collapsing procedure



$$M \in \text{MLQ}((6, 4, 2))$$

$$\text{maj}(M) = 2 + 1 + 3 = 6$$

Collapsing procedure

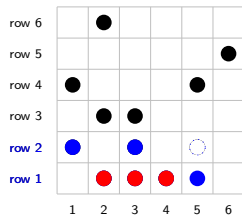


$$Q = \begin{array}{|c|c|c|} \hline 1 & 1 & 1 \\ \hline \end{array}$$

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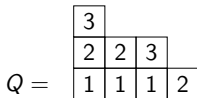
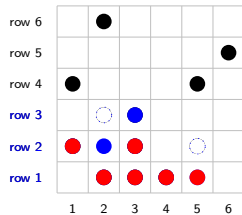


$$Q = \begin{array}{|c|c|c|c|} \hline 2 & 2 & & \\ \hline 1 & 1 & 1 & 2 \\ \hline \end{array}$$

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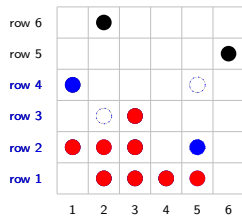
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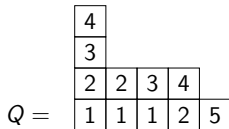
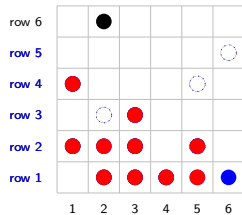
$Q =$

4			
3			
2	2	3	4
1	1	1	2

$$M \in \text{MLQ}((6, 4, 2))$$

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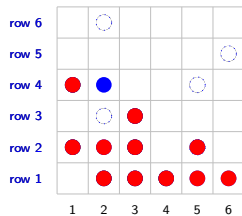
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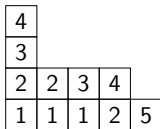
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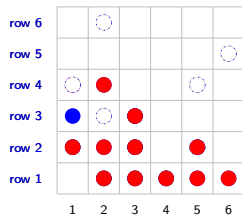
$Q =$



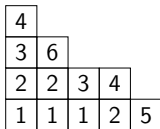
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Collapsing procedure



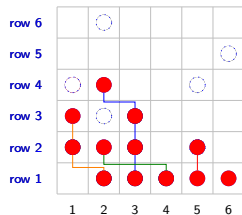
$Q =$



$$M \in \text{MLQ}((6, 4, 2))$$

$$\text{maj}(M) = 2 + 1 + 3 = 6$$

Collapsing procedure



$Q =$

4					
3	6				
2	2	3	4		
1	1	1	2	5	

$$M \in \text{MLQ}((6, 4, 2))$$

$$\text{maj}(M) = 2 + 1 + 3 = 6$$

$$M_0 \in \text{MLQ}_0((4, 3, 2, 2, 1))$$

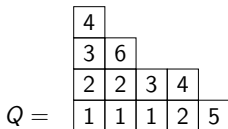
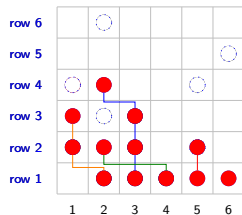
$$\text{charge}(Q) = \text{charge}(4|36|2234|11125)$$

$$= \text{charge}(436215) + \text{charge}(3412) + \text{charge}(21)$$

$$= 2 + (1 + 3) + 0$$

$$Q \in \text{SSYT}(\mu', \lambda')$$

Collapsing procedure



$$M \in \text{MLQ}((6, 4, 2))$$

$$\text{maj}(M) = 2 + 1 + 3 = 6$$

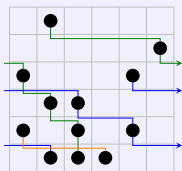
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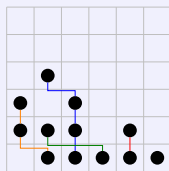
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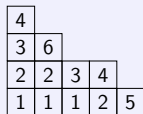
$$\lambda = (6, 4, 2)$$

$$\text{maj}(M) = 6$$

 $\rho \rightarrow$


$$\lambda = (4, 3, 2, 2, 1)$$

$$\text{maj}(M_0) = 0$$

 $,$


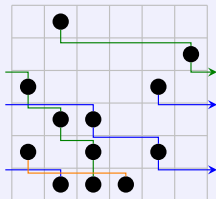
$$\text{charge}(Q) = 6$$

Collapsing procedure

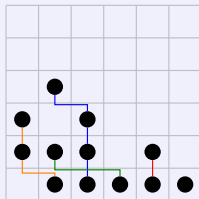
Lemma

$$\text{RSK}^{\text{col}} : (a_1, \dots, a_n) \rightarrow \text{SSYT}(\mu') \times \text{SSYT}(\mu)$$

$$\text{RSK}^{\text{col}}(\text{rw}(M)) = \text{RSK}^{\text{col}}(\text{rw}(\rho(M)))$$



and



$\xrightarrow{\text{RSK}^{\text{col}}}$

6				
5	5			
3	4			
2	2	3		
1	1	2	3	

$$\text{rw}(M) = 234|135|23|15|6|2$$

$$\text{rw}(\rho(M)) = 23456|1235|13|2$$

Littlewood-Richardson rule via MLQs

Theorem

$$s_\lambda s_\mu = \sum_{\nu} c_{\lambda\mu}^{\nu} s_{\nu} = \sum_{\nu} c_{\lambda'\mu'}^{\nu'} s_{\nu}$$

- Using the collapsing procedure:

$$\text{MLQ}_0(\lambda) \times \text{MLQ}_0(\mu) \rightarrow \text{MLQ}_0(\nu) \times \text{SSYT}^*(\nu'/\lambda', \mu)$$

row 3		●	●		
row 2	●		●		●
row 1			●	●	●
row 3	●				
row 2			●	●	
row 1	●	●	●		●

$Q = \emptyset$

$$M_1 \in \text{MLQ}(\lambda), \quad \lambda = (4, 2, 1)$$

$$M_2 \in \text{MLQ}(\mu), \quad \mu = (3, 3, 2)$$

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row 3		●	●		
row 2	●		●		●
row 1			●	●	●
row 3	●				
row 2			●	●	
row 1	●	●	●		●

 $Q =$

3			
2	2		
1	1	1	1

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row 3		●	●		
row 2	●		●		●
row 1			○	○	○
row 3	●		●		
row 2			●	○	●
row 1	●	●	●	●	●

$$Q =$$

3	1			
2	2	1		
1	1	1	1	1

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row 3		●	●		
row 2	○		○		○
row 1	●		●	○	○
row 3	●		●	○	●
row 2			●	●	●
row 1	●	●	●	●	●

$$Q =$$

2	2			
3	1	2		
2	2	1		
1	1	1	1	1

$$M_1 \in \text{MLQ}(\lambda), \quad \lambda = (4, 2, 1)$$

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row 3		○	○		
row 2	○		●		○
row 1	●	●	●	○	○
row 3	●		●	○	●
row 2			●	●	●
row 1	●	●	●	●	●

$Q =$

3					
2	2	3			
3	1	2			
2	2	1			
1	1	1	1	1	

$$M_1 \in \text{MLQ}(\lambda), \quad \lambda = (4, 2, 1)$$

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row 3		○	○		
row 2	○		●		○
row 1	●	●	●	○	○
row 3	●		●	○	●
row 2			●	●	●
row 1	●	●	●	●	●

Q =

3					
2	2	3			
3	1	2			
2	2	1			
1	1	1	1	1	

$$M_1 \in \text{MLQ}(\lambda), \quad \lambda = (4, 2, 1)$$

$$M_2 \in \text{MLQ}(\mu), \quad \mu = (3, 3, 2)$$

$$M \in \text{MLQ}(\nu), \quad \nu = (5, 4, 4, 1, 1)$$

$$Q \in \text{SSYT}^*(\nu'/\lambda', \mu')$$

- $\text{SSYT}^*(\nu'/\lambda', \mu')$ is the set of **Yamanouchi fillings** of ν'/λ' with content μ'

modified Macdonald polynomials $\tilde{H}_\lambda(X; q, t)$

Garsia and Haiman '96 introduced the modified Macdonald polynomials, denoted by $\tilde{H}_\lambda(X; q, t)$ as a combinatorial version of the P_λ 's

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Garsia and Haiman '96 introduced the modified Macdonald polynomials, denoted by $\tilde{H}_\lambda(X; q, t)$ as a combinatorial version of the P_λ 's

- obtained from a normalized form of $P_\lambda(X; q, t)$ by plethystic substitution:

$$\tilde{H}_\lambda(X; q, t) = t^{n(\lambda)} J_\lambda \left[\frac{X}{1-t^{-1}}; q, t^{-1} \right]$$

where J_λ is a scalar multiple of P_λ .

Example: $\tilde{H}_{(2,1)}(X; q, t) = m_{(3)} + (1+q+t)m_{(2,1)} + (1+2q+2t+qt)m_{(1,1,1)}$

modified Macdonald polynomials $\tilde{H}_\lambda(X; q, t)$

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From multiline queues to a new formula for \tilde{H}_λ

- $\tilde{H}_\lambda(X; q, t)$ is obtained from the integral form of P_λ via **plethysm**:

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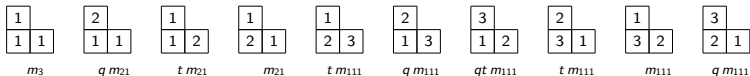
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- the resulting objects are of the same flavor as multiline queues, except that **multiple balls are allowed at each location**. (This translates to removing the “non-attacking” condition from the corresponding tableaux)

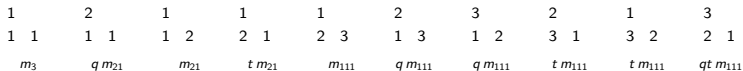
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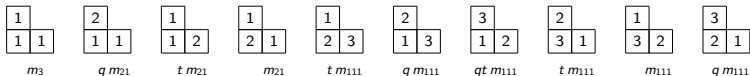
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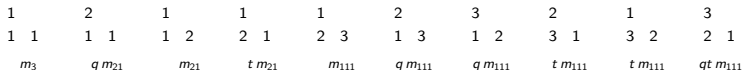
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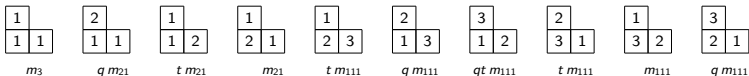


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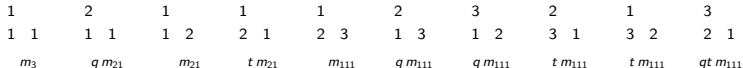
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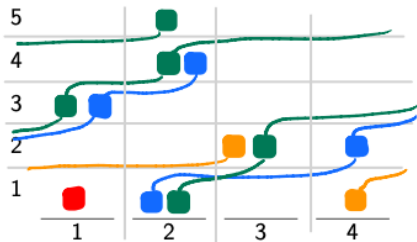


• while the **inv** and **quinv** statistics appear very similar, there does not seem to be an easy way to go from one to the other – is there a bijective proof? **Update! Yes there is due to Loehr '22**

motivation for queue inversions: multiline diagrams

The tableaux are actually representing a [queueing system](#) which is an arrangement of lattice paths/strings

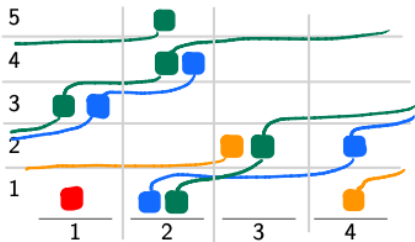
2				
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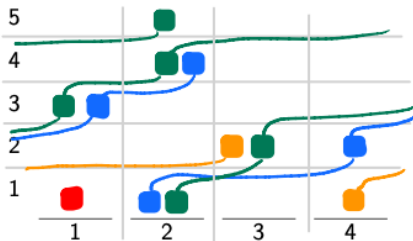
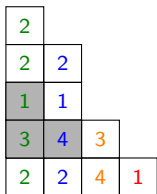
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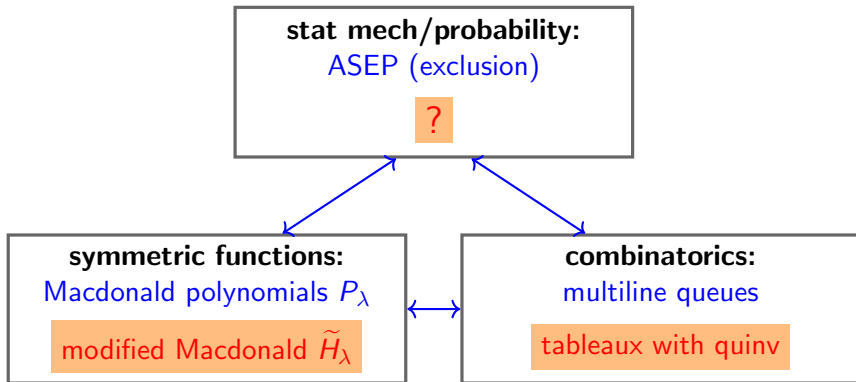


“plethystic version” of
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skipped particle

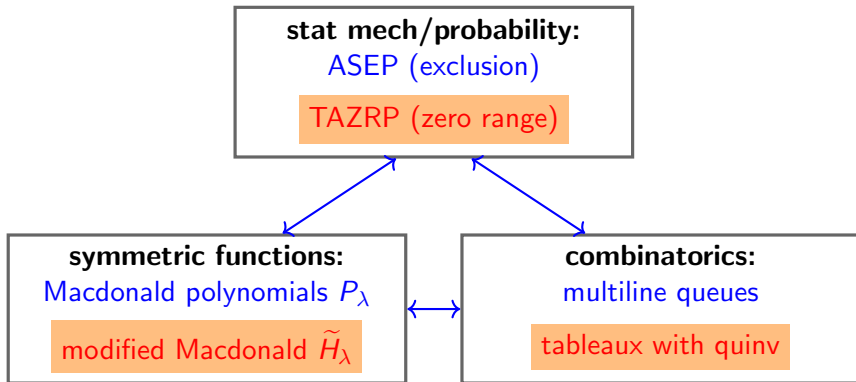


Big picture



What is the analogous **interacting particle system** whose partition function is a specialization of \tilde{H}_λ ?

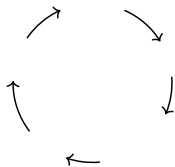
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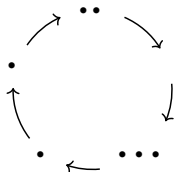
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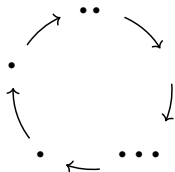
Here, $n = 5$, $k = 7$

$$\tau = (11 \mid \cdot \mid 111 \mid 1 \mid 1)$$

- simplest case: there are k indistinguishable particles, moving **clockwise**. A configuration $\tau = (\tau_1, \dots, \tau_n)$ is any allocation of the k particles on the n sites.

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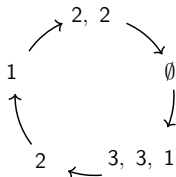
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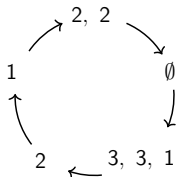
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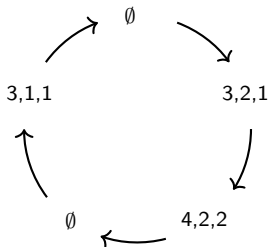
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- Kuniba–Maruyama–Okado (2015+)** (and others) have studied many multispecies variants of the TAZRP. **All of these are integrable!** The version we will describe was first studied by **Takayama '15**

the mTAZRP: states

- Fix a (circular 1D) lattice on n sites and a partition $\lambda = (\lambda_1 \geq \dots \geq \lambda_k > 0)$ for the particle types



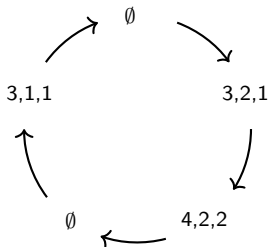
$$n = 5$$

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- Fix a (circular 1D) lattice on n sites and a partition $\lambda = (\lambda_1 \geq \dots \geq \lambda_k > 0)$ for the particle types
- TAZRP(λ, n) is a Markov chain whose states are *multiset compositions* τ of type λ , with n (possibly empty) parts



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$$f_j(k) = x_j^{-1} t^d \sum_{u=0}^{c-1} t^u$$

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For example: If site j contains the particles $\{4, 3, 3, 1, 1, 1\}$, then:

$$k = 1: \quad d = 3, \quad c = 3, \quad f_j(1) = x_j^{-1} t^3 (1 + t + t^2).$$

$$k = 3: \quad d = 1, \quad c = 2, \quad f_j(3) = x_j^{-1} t (1 + t).$$

$$k = 4: \quad d = 0, \quad c = 1, \quad f_j(4) = x_j^{-1}.$$

Lumping of tableaux to mTAZRP

Very similar projection map as for the ASEP.

- Given a filling σ , read the state $\tau \in \text{TAZRP}(\lambda, n)$ from the bottom row of σ as follows:

τ_j is the multiset $\{\lambda_i : \sigma(1, i) = j\}$

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- For example, for $\lambda = (2, 1, 1)$ and $n = 3$, the following are all the tableaux that correspond to the state $\tau = (21 \mid \cdot \mid 1)$:

1				2				3				1				2				3			
1	1	3		1	1	3		1	1	3		1	3	1		1	3	1		1	3	1	

TAZRP probabilities and tableaux

Theorem (Ayyer–M–Martin '21)

Fix λ, n . The (unnormalized) stationary probability of $\tau \in \text{TAZRP}(\lambda, n)$ is

$$\tilde{\text{Pr}}(\tau) = \sum_{\substack{\sigma: \text{dg}(\lambda) \rightarrow [n] \\ \sigma \text{ has type } \tau}} x^\sigma t^{\text{quinv}(\sigma)}.$$

Corollary

The so-called *partition function* of $\text{TAZRP}(\lambda, n)$ is

$$\mathcal{Z}_{\lambda, n}(x_1, \dots, x_n; t) = \tilde{H}_\lambda(x_1, \dots, x_n; 1, t).$$

Proof: construction of a Markov chain on tableaux that lumps to the TAZRP.

an example for $\lambda = (2, 1, 1)$ and $n = 2$

The stationary distribution is:

$$\begin{array}{l|l} \left(\begin{array}{c|c} 211 & \cdot \\ 11 & 2 \end{array} \right) & x_1^3(x_1 + x_2) \\ \left(\begin{array}{c|c} 11 & 2 \end{array} \right) & x_1^2 x_2(t^2 x_2 + x_1) \\ \left(\begin{array}{c|c} 21 & 1 \end{array} \right) & x_1^2 x_2(tx_1 + x_2)(1 + t) \\ \left(\begin{array}{c|c} 1 & 21 \end{array} \right) & x_1 x_2^2(x_1 + tx_2)(1 + t) \\ \left(\begin{array}{c|c} 2 & 11 \end{array} \right) & x_1 x_2^2(t^2 x_1 + x_2) \\ \left(\begin{array}{c|c} \cdot & 211 \end{array} \right) & x_2^3(x_1 + x_2) \end{array}$$

Example computation for $(21 | 1)$:

$$\begin{array}{|c|} \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline 2 \\ \hline \end{array} : t^2, \quad \begin{array}{|c|} \hline 2 \\ \hline \end{array} \begin{array}{|c|} \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline 2 \\ \hline \end{array} : t, \quad \begin{array}{|c|} \hline 1 \\ \hline \end{array} \begin{array}{|c|} \hline 2 \\ \hline \end{array} \begin{array}{|c|} \hline 1 \\ \hline \end{array} : t, \quad \begin{array}{|c|} \hline 2 \\ \hline \end{array} \begin{array}{|c|} \hline 2 \\ \hline \end{array} \begin{array}{|c|} \hline 1 \\ \hline \end{array} : 1$$

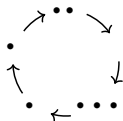
the total is: $\tilde{\text{Pr}}(21|1) = x_1^2 x_2(tx_1 + x_2)(1 + t)$.

Current

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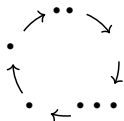
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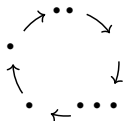
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Proposition (Current for the single species TAZRP)

For the **single-species TAZRP** on n sites with m particles, the current is given by

$$J = [m]_t \frac{\tilde{H}_{(1^{m-1})}(x_1, \dots, x_n; \mathbf{1}, t)}{\tilde{H}_{(1^m)}(x_1, \dots, x_n; \mathbf{1}, t)}.$$

Current

Theorem (Ayyer-M-Martin '22+)

Let $\lambda = (1^{m_1}, \dots, k^{m_k})$, and let $1 \leq j \leq k$. The current of the particle of type j of the TAZRP of type λ on n sites is given by

$$J = [m_j + \dots + m_k]_t \frac{\tilde{H}_{(1^{m_j + \dots + m_k - 1})}}{\tilde{H}_{(1^{m_j + \dots + m_k})}} - [m_{j+1} + \dots + m_k]_t \frac{\tilde{H}_{(1^{m_{j+1} + \dots + m_k - 1})}}{\tilde{H}_{(1^{m_{j+1} + \dots + m_k})}}$$

Densities

- Take TAZRP(λ, n) with content $\lambda = (1^{m_1}, 2^{m_2}, \dots, k^{m_k})$.
- Define $z_j^{(\ell)}$ to be the random variable counting the number of particles of type ℓ at site j in a configuration of TAZRP(λ, n).
- Denote the expectation in the stationary distribution by $\langle z_j^{(\ell)} \rangle$.

Theorem (Ayyer-M-Martin '22+)

For $1 \leq \ell \leq k$, the density of the ℓ 'th species at site 1 is given by

$$\langle z_1^{(\ell)} \rangle = x_1 \partial_{x_1} \log \left(\frac{\tilde{H}_{(1^{m_\ell + \dots + m_k})}(x_1, \dots, x_n; 1, t)}{\tilde{H}_{(1^{m_{\ell+1} + \dots + m_k})}(x_1, \dots, x_n; 1, t)} \right).$$

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Corollary

$\langle z_1^{(\ell)} \rangle$ is symmetric in the variables $\{x_2, \dots, x_n\}$.

Local correlations

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$\mathbb{P}_{\lambda, n}(\overline{w})$ is *tcb*symmetric in the variables $\{x_{\ell+1}, \dots, x_n\}$.

final remarks

- Explicit bijection from the **inv** to the **quinv** statistic?

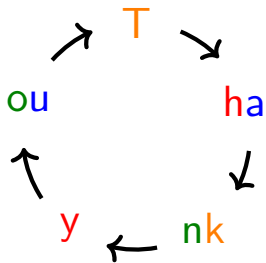


(recently found by Loehr)

- Can we find a dynamical process that incorporates the q as a parameter?

This seems difficult because

- We lose factorization of \tilde{H}_λ
 - We lose translation invariance
- Suitable quasisymmetric version of modified Macdonald polynomials?
Nonsymmetric version?



Modified Macdonald polynomials and the multispecies zero range process:
arXiv:2011.06117, arXiv:2209.09859