## The multispecies zero range process and modified Macdonald polynomials

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## Macdonald polynomials and hopping particles

- Macdonald polynomials (Macdonald '88) $P_{\lambda}(X ; q, t)$ are a remarkable family of symmetric functions with coefficients in $\mathbb{Q}[q, t]$ that simultaneously generalize Schur functions $(q=t)$, Hall-Littlewood polynomials $(q=0)$, Jack polynomials ( $t=q^{\alpha}$ and $q \rightarrow 1$ ), others..


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- $P_{\lambda}\left(x_{1}, \ldots, x_{n} ; q, t\right)$ specializes to the partition function of the multispecies ASEP at $x_{1}=\cdots=x_{n}=q=1$ (Cantini-de Gier-Wheeler '15)


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\begin{aligned}
n & =6 \\
\lambda & =(3,3,2,1,1) \\
\lambda^{\prime} & =(5,3,2)
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\mathrm{wt}(M)=x^{M} t^{\text {skipped }} \prod_{\text {pairings }} q^{(\ell-r+1) \delta_{\text {wrap }}} \frac{1-t}{1-q^{\ell-r+1} t^{\text {free }}}
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- Can be represented by a tableau, where each string is mapped to a column


## multiline queues, the ASEP, and Macdonald polynomials

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Theorem (Cantini-deGier-Wheeler '15)
At $x_{1}=\cdots=x_{n}=q=1, P_{\lambda}$ specializes to the partition function of $\operatorname{ASEP}(\lambda, n)$ :

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Fix $\lambda, n$. The (unnormalized) stationary probability of $\tau \in \operatorname{ASEP}(\lambda, n)$ is

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## Theorem (Corteel-M-Williams '18)

The Macdonald polynomial is given by

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P_{\lambda}\left(x_{1}, \ldots, x_{n} ; q, t\right)=\sum_{M \in \operatorname{MLQ}(\lambda, n)} w t(M)(X ; q, t)
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## modified Macdonald polynomials

- modified Macdonald polynomials $\widetilde{H}_{\lambda}(X ; q, t)$ (Garsia-Haiman '96) are a combinatorial form of $P_{\lambda}(X ; q, t)$, obtained via plethystic substitution:

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\widetilde{H}_{\lambda}(X ; q, t)=t^{n(\lambda)} J_{\lambda}\left[\frac{X}{1-t^{-1}} ; q, t^{-1}\right]
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& \tilde{H}_{(2,1)}=q t s_{(1,1,1)}+(q+t) s_{(2,1)}+s_{3} \\
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Our goal is to get a multiline queue-esque construction for $\widetilde{H}_{\lambda}(X ; q, t)$ by interpreting plethysm through multiline queues

From multiline queues to $\widetilde{H}_{\lambda}$

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\widetilde{H}_{\lambda}(X ; \boldsymbol{q}, t) & =f_{\lambda}(\boldsymbol{q}, t) P_{\lambda}\left[\frac{X}{1-t^{-1}} ; \boldsymbol{q}, t^{-1}\right] \\
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this can be represented by a multiline queue with columns labeled $x_{1}, x_{1} t^{-1}, x_{1} t^{-2}, \ldots, x_{2}, x_{2} t^{-1}, x_{2} t^{-2}, \ldots$

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we get multiline diagrams with no restriction on the number of particles at each site!

## From multiline diagrams to tableaux

Each string is mapped to a column in the tableau:

multiline diagram of type $(\lambda, n) \rightarrow$ a tableau $\sigma: \lambda^{\prime} \rightarrow[n]$

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| $x$ |  |  | where $x<y<z($ cyclically $\bmod n)$ |
| :--- | :--- | :--- | :--- |
| $y$ | $\cdots$ | $z$ |  |

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## Theorem (Ayyer-M-Martin '21)

Let $\lambda$ be a partition. The modified Macdonald polynomial equals

$$
\widetilde{H}_{\lambda}\left(x_{1}, \ldots, x_{n} ; q, t\right)=\sum_{\sigma: \lambda^{\prime} \rightarrow[n]} q^{\operatorname{maj}(\sigma)} t^{\text {quinv }(\sigma)} x^{\sigma}
$$



The particle process corresponding to multiline diagrams is the multispecies totally asymmetric zero range process!


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The state of the particle process is read off the bottom row of the diagram:

$$
\tau=(4,1|3| 4,1)
$$

## multispecies totally asymmetric zero range process

- a zero range process (ZRP) is continuous-time 1D stochastic process (Spitzer '70). Each site can contain any number of particles, and particles hop from site $j$ to site $j \pm 1$ with rates that depend only on the content of site $j$.


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- Kuniba-Maruyama-Okado (2015+) (and others) have studied many variants of the TAZRP (all of which are integrable!). The version we study was first studied by Takayama '15


## TAZRP probabilities and tableaux

## Theorem (Ayyer-M-Martin '21)

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

$$
\widetilde{\operatorname{Pr}}(\tau)=\sum_{\substack{\sigma: \lambda^{\prime} \rightarrow[n] \\ \sigma \text { has type } \tau}} x^{\sigma} t^{\mathrm{quinv}(\sigma)}
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## Corollary

The partition function of $\operatorname{TAZRP}(\lambda, n)$ is

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\mathcal{Z}_{\lambda, n}\left(x_{1}, \ldots, x_{n} ; t\right)=\widetilde{H}_{\lambda}\left(x_{1}, \ldots, x_{n} ; 1, t\right)
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Proof: construction of a Markov chain on tableaux that lumps to the TAZRP.

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- if $c=(1, j)$ is in the bottom row, the rate $f(\sigma, c)$ matches the transition rate $f_{\sigma(c)}\left(\lambda_{j}\right)$ of the corresponding particle in the TAZRP.
- (when $\lambda$ has repeated parts, we need to do some more work!)


## Observables

Observables are macroscopic properties of the TAZRP process that can be measured e.g. through simulation. These include:

- stationary probabilities/the partition function
- the current
- densities of particle content at individual sites
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- correlations of behaviors of tuples of particles or tuples of sites
what can we learn about the observables using the enhanced Markov chain on multiline diagrams/tableaux and the connection with modified Macdonald polynomials?


## Current

The current of particle $\ell$ across the edge $j$ is defined as the number of particles of type $\ell$ traversing the edge $j$ per unit of time in the large time limit.

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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{\widetilde{H}_{\left(1^{m-1}\right)}\left(x_{1}, \ldots, x_{n} ; 1, t\right)}{\widetilde{H}_{\left(1^{m}\right)}\left(x_{1}, \ldots, x_{n} ; 1, t\right)}
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This comes from the stationary probability of the 1 -species TAZRP:

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\pi(\tau)=\frac{1}{\widetilde{H}_{\left(1^{m}\right)}\left(x_{1}, \ldots, x_{n} ; 1, t\right)}\left[\begin{array}{c}
m \\
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## Theorem (Ayyer-M-Martin '22+)

Let $\lambda=\left(1^{m_{1}}, \ldots, k^{m_{k}}\right)$, and let $1 \leq j \leq k$. The current of the particle of type $j$ of the TAZRP of type $\lambda$ on $n$ sites is given by

$$
J=\left[m_{j}+\cdots+m_{k}\right]_{t} \frac{\widetilde{H}_{\left(1^{m_{j}+\cdots+m_{k}-1}\right)}}{\widetilde{H}_{\left(1^{m_{j}+\cdots+m_{k}}\right)}}-\left[m_{j+1}+\cdots+m_{k}\right]_{t} \frac{\widetilde{H}_{\left(1^{m_{j+1}+\cdots+m_{k}-1}\right)}}{\widetilde{H}_{\left(1^{m_{j+1}+\cdots+m_{k}}\right)}}
$$

## Particle densities

- Take $\operatorname{TAZRP}(\lambda, n)$ with content $\lambda=\left(1^{m_{1}}, 2^{m_{2}}, \ldots, k^{m_{k}}\right)$
- $z_{j}^{(\ell)}$ : random variable counting the number of particles of type $\ell$ at site $j$ in a configuration of $\operatorname{TAZRP}(\lambda, n)$
- $\left\langle z_{j}^{(\ell)}\right\rangle$ : the expectation in the stationary distribution


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## Theorem (Ayyer-M-Martin '22)

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

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\left\langle z_{1}^{(\ell)}\right\rangle=x_{1} \partial_{x_{1}} \log \left(\frac{\widetilde{H}_{\left(1^{m_{\ell}}+\cdots+m_{k}\right)}\left(x_{1}, \ldots, x_{n} ; 1, t\right)}{\widetilde{H}_{\left(1^{m_{\ell+1}+\cdots+m_{k}}\right)}\left(x_{1}, \ldots, x_{n} ; 1, t\right)}\right) .
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This generalizes to probabilities for fixed content on an interval of sites $1, \ldots, k$ : these probabilities are symmetric in the variables $\left\{x_{k+1}, \ldots, x_{n}\right\}$ !

## Future directions

- a few other particle models have been found to be connected to Macdonald polynomials:
- inhomogeneous ASEP and Schubert polynomials Lam-Williams '11
- long-range ASEP and $P_{\lambda}\left(x_{1}, \ldots, x_{n} ; 1, t\right)$ Angel-Ayyer-Martin '23+
- multi-hopping multispecies TAZRP and $\widetilde{H}_{\lambda}\left(x_{1}, \ldots, x_{n} ; 1,0\right), \widetilde{H}_{\lambda}(1, \ldots, 1 ; 1, t)$ Esipova-M '23+, Corteel-Keating '23+

What other particle processes can be described through Macdonald (or related) polynomials?

- Using multiline queues, Corteel-Haglund-M-Mason-Williams FPSAC ' 20 defined quasisymmetric Macdonald polynomials which refine the symmetric Macdonald polynomial $P_{\lambda}(X ; q, t)$. Can we use a parallel construction to define an interesting family of quasisymmetric polynomials that refine $\widetilde{H}_{\lambda}(X ; q, t)$ ?


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

## Symmetries in local correlations

- Fix $\lambda, n$, and $0 \leq \ell \leq n$, and let $w$ be a configuration of the TAZRP on the first $\ell$ sites of type $\mu$, where $\mu \subseteq \lambda$.


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- Example: let $\lambda=(2,2,1,1), n=4, \ell=2$, and $w=(2 \mid 1)$.

Configurations contributing to $\mathbb{P}_{\lambda, n}(\bar{w})$ are

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