

Markov Dynamics on Macdonald Processes

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Main goal

Present axiomatic approach to known models of Markov dynamics, and get new examples out of it

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A motivating example from **random matrices**: appearance of Dyson's Brownian motion (originally, from GUE random matrices [Dyson '62]):

- ① Path transformation of independent Brownian motions related to Robinson–Schensted–Knuth correspondence [O'Connell '03]
- ② Warren's construction '07

Both extend to **hierarchies** of diffusions which are **different**, but have the **same** fixed-time distributions (corresponding to GUE corners). Why? Are there any other diffusions with these properties?

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I will explain a discrete version of the problem, and a solution.

Outline

① “Schur level”

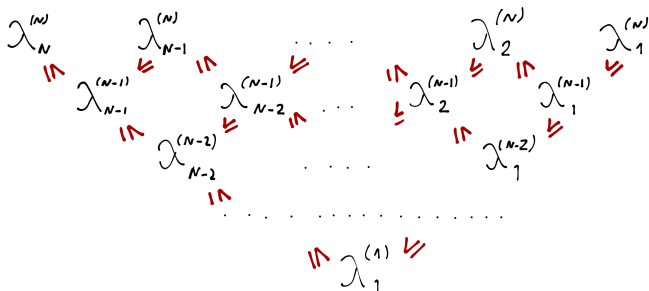
- Push-block dynamics (“Warren”)
- RSK dynamics (“Path transformation”)
- **Unifying axioms**
- **New RSK correspondences**

② “Macdonald level”

- From Schur to Macdonald
- q -deformed 1d particle systems: **new examples**
- **Randomized insertion algorithm for triangular matrices over a finite field**

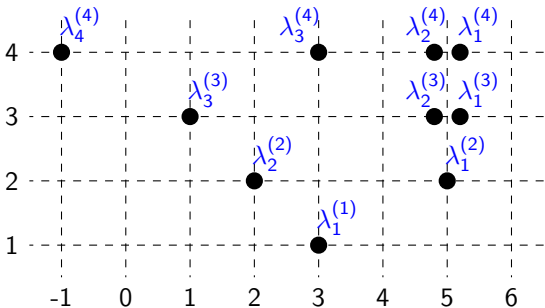
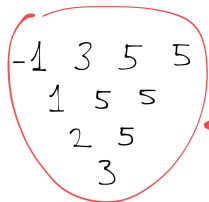
Interlacing integer arrays

Interlacing integer arrays (= Gelfand-Tsetlin schemes)



Main object: **continuous-time Markov dynamics** on the space of interlacing integer arrays.

Interlacing integer arrays \longleftrightarrow particles in 2 dimensions



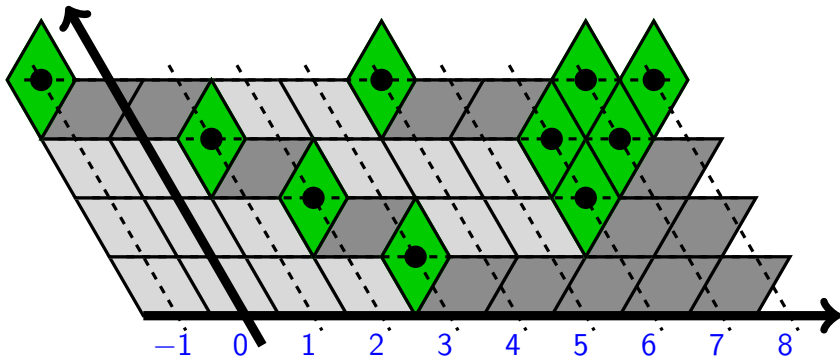
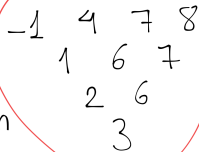
1 particle at level 1,
2 particles at level 2, etc.

Interlacing integer arrays \longleftrightarrow lozenge tilings

+0 ... +(N-2) +(N-1)



shift coordinates so that in each row they become distinct

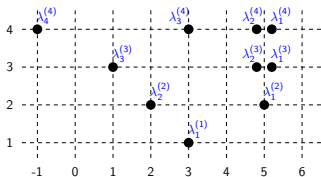


Two examples of dynamics on interlacing arrays:

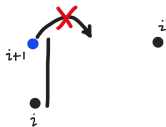
- Push-block dynamics
- Robinson–Schensted–Knuth (RSK) dynamics
- Common properties of the two dynamics

Push-block dynamics [Borodin–Ferrari '08], [Gorin–Shkolnikov '12]

1. Each particle $\lambda_j^{(k)}$ jumps to the right by one according to an independent exponential clock of rate 1.



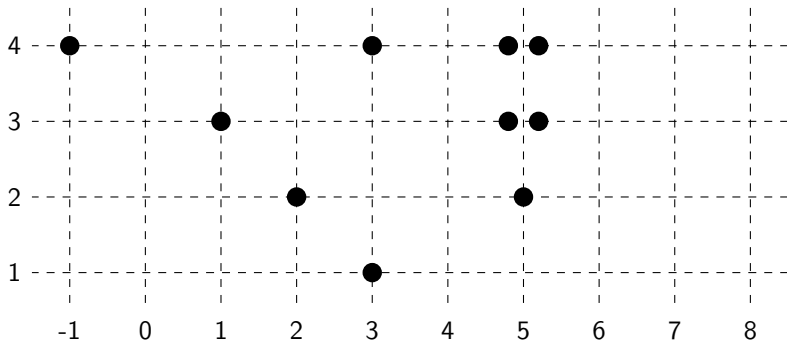
2. If it is **blocked** from below, there is no jump



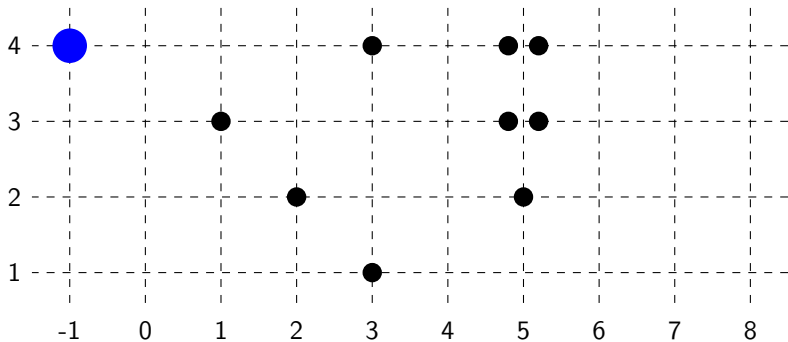
3. If violates interlacing with above, it **pushes** the above particles



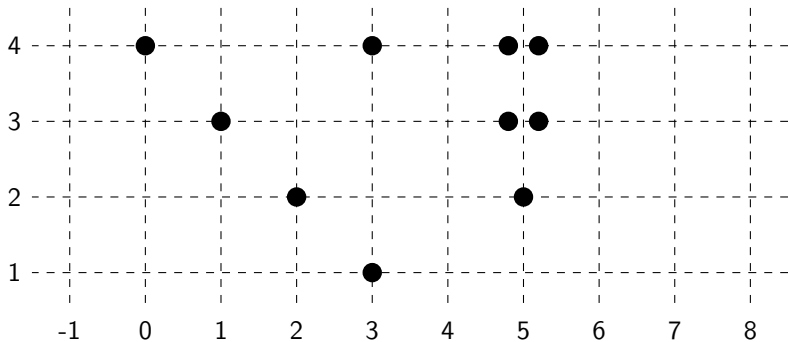
Push-block dynamics



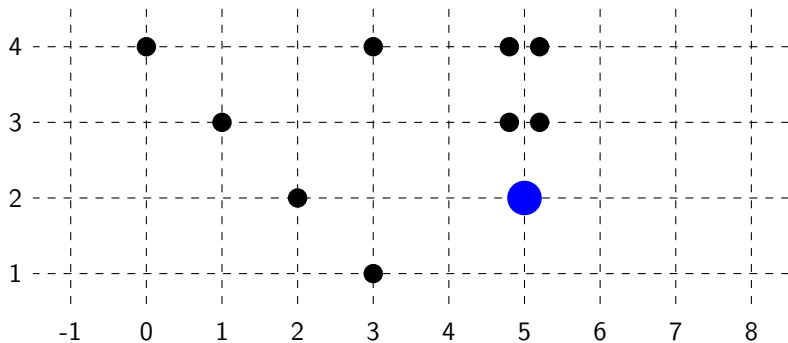
Push-block dynamics



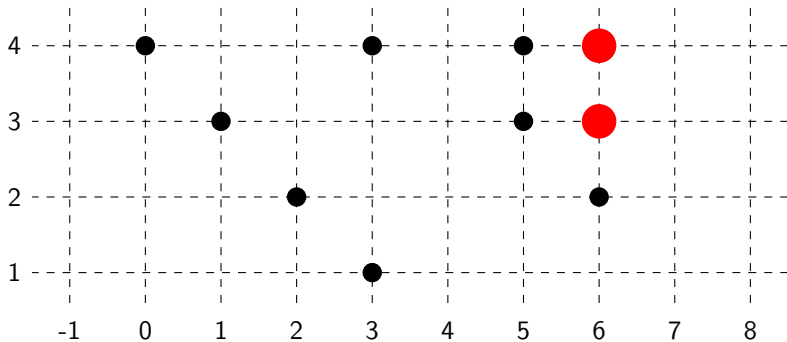
Push-block dynamics



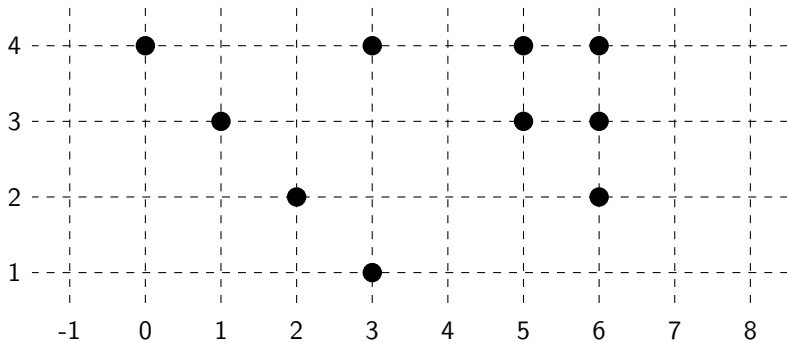
Push-block dynamics



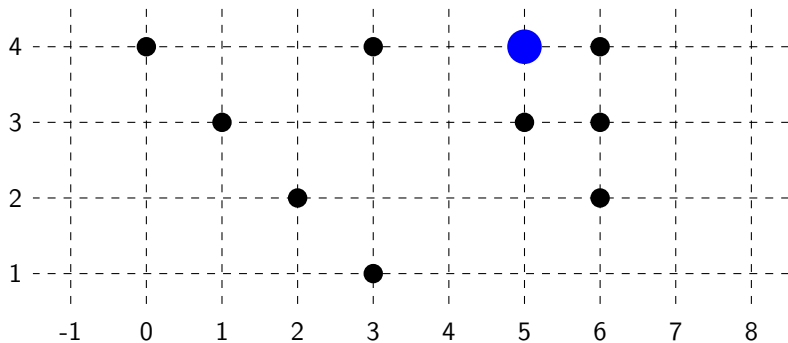
Push-block dynamics



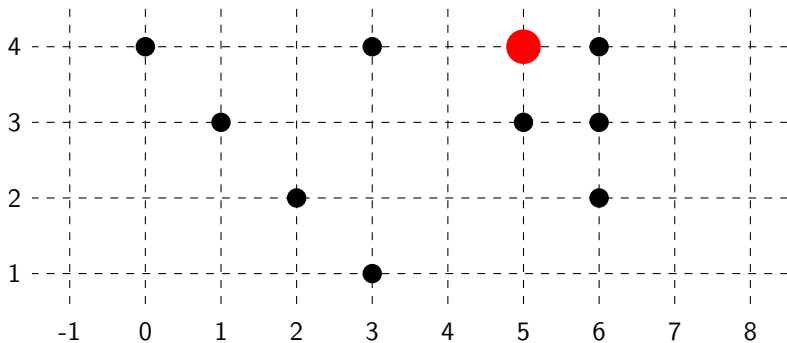
Push-block dynamics



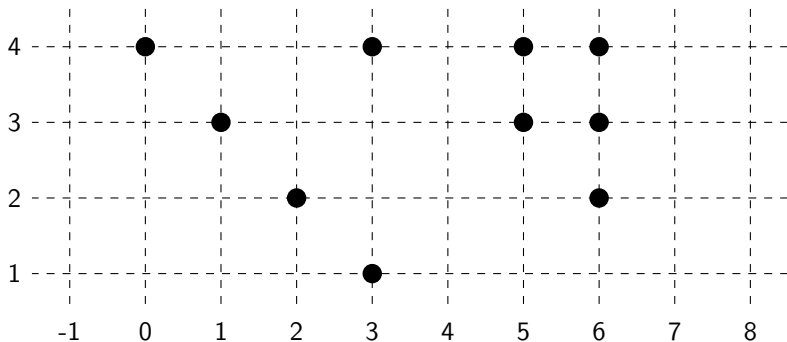
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Push-block dynamics

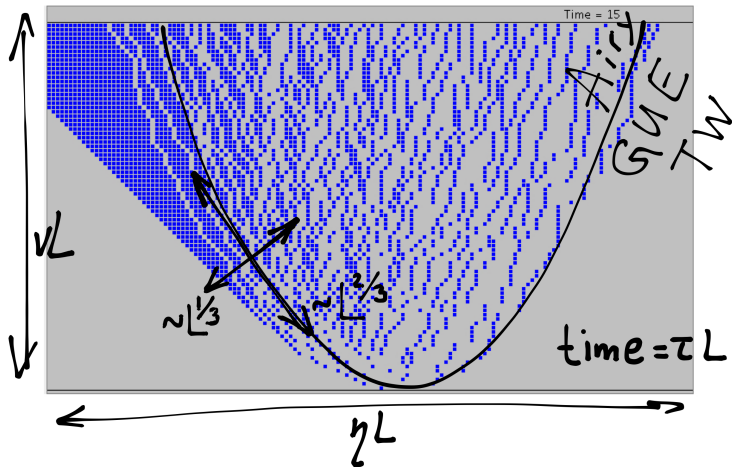


Push-block dynamics



Simulation

Push-block dynamics: KPZ universality [BF '08]



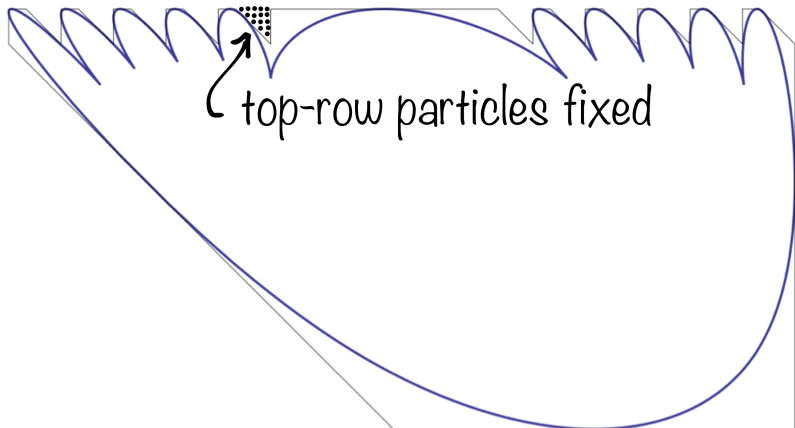
+ fluctuations $\sim L^{1/3}$ with time
(L — large parameter)

Remark: Other tiling models

Scaling orders $L^{1/3}$ – $L^{2/3}$, GUE Tracy–Widom distribution and Airy process found in other models of random lozenge tilings:

[Okounkov–Reshetikhin '07],

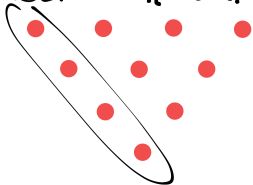
[Baik–Kriecherbauer–McLaughlin–Miller '07], **[P. '12]**



1d projections of the push-block dynamics

TASEP and PushTASEP

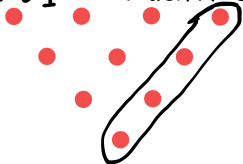
$$\text{TASEP: } x_n = \lambda_1^{(n)} - n$$



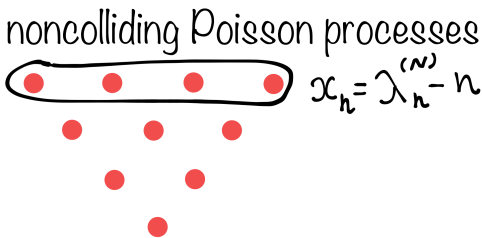
Markovian projection to the leftmost particles — TASEP

Markovian projection to the rightmost particles — PushTASEP

$$x_n = \lambda_1^{(n)} + n : \text{PushTASEP}$$



“Discrete version” of Dyson’s Brownian motion



Started from the empty initial state $\lambda_j^{(k)} = 0$, the evolution of the particles in each N th row is **Markovian**:

- Rate 1 Poisson processes conditioned never to intersect;
- Equivalently, Doob's h -transform of independent Poisson processes, with $h(x_1, \dots, x_N) = \prod_{i < j} (x_i - x_j)$.

Two examples of dynamics on interlacing arrays:

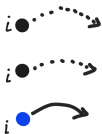
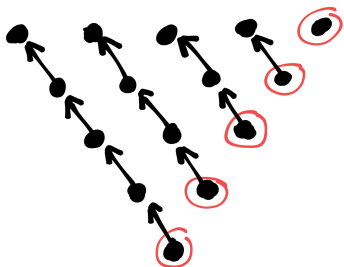
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RSK dynamics [Johansson '99,'02], [O'Connell '03]

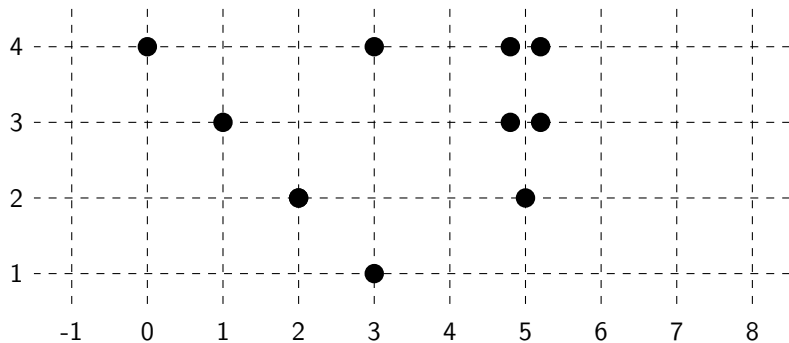
1. Each **rightmost** particle $\lambda_1^{(k)}$ jumps to the right by one according to an independent exponential clock of rate 1.

2. When any particle $\lambda_j^{(h)}$ moves, it triggers either the move $\lambda_j^{(h+1)} \mapsto \lambda_j^{(h+1)} + 1$, or $\lambda_{j+1}^{(h+1)} \mapsto \lambda_{j+1}^{(h+1)} + 1$ (exactly one of them).

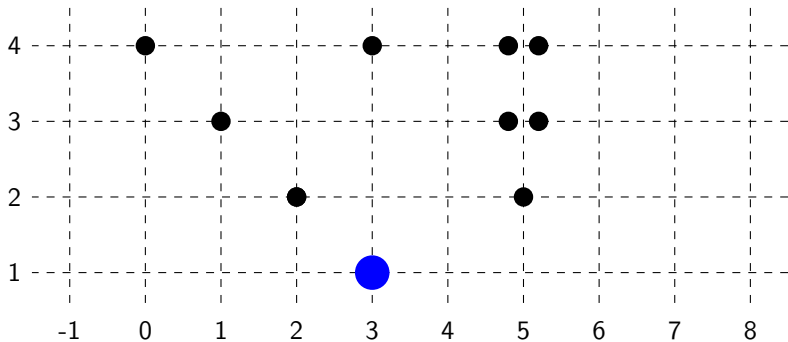
The second one is chosen generically, while the first one is chosen only if $\lambda_j^{(h+1)} = \lambda_j^{(h)}$, i.e., if the move $\lambda_j^{(h)} \mapsto \lambda_j^{(h)} + 1$ violated the interlacing constraint (**push rule**).



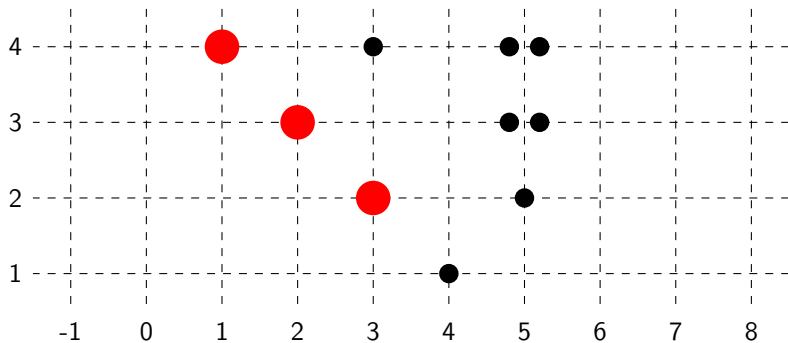
RSK dynamics



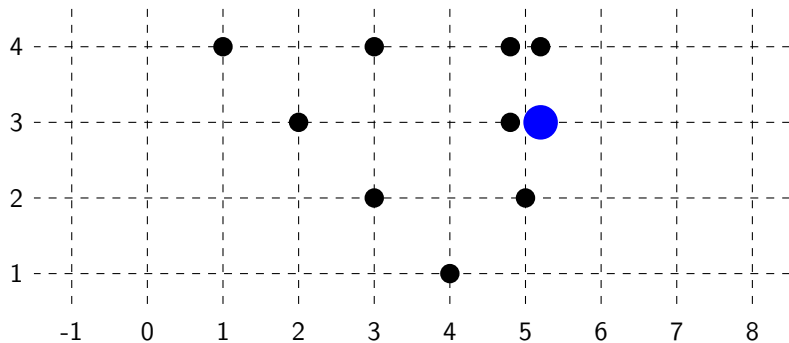
RSK dynamics



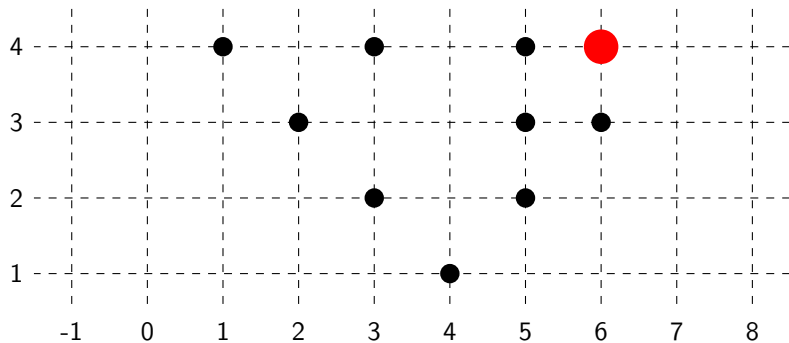
RSK dynamics



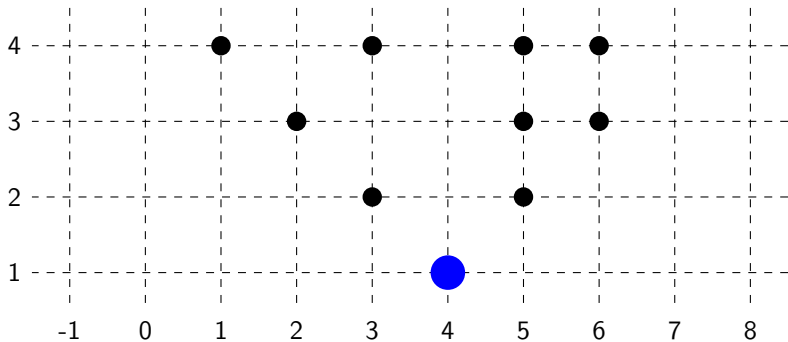
RSK dynamics



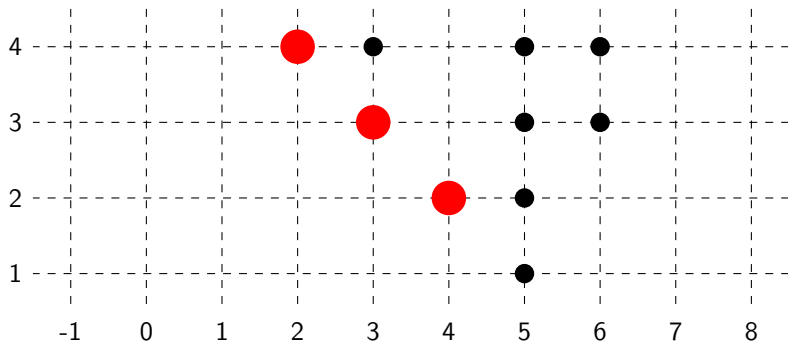
RSK dynamics



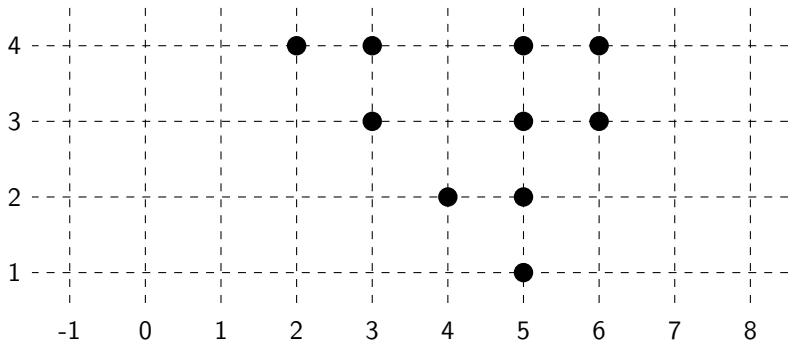
RSK dynamics



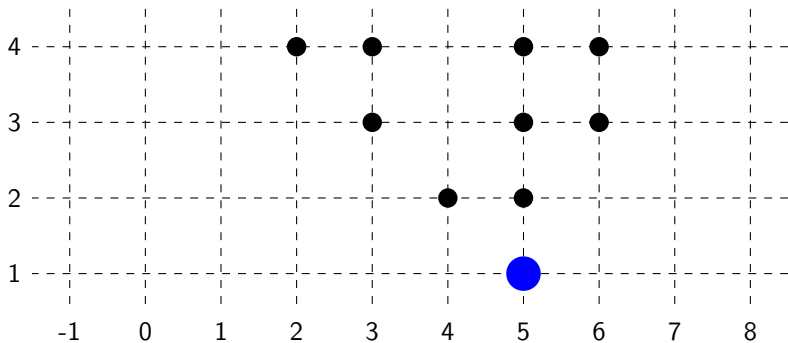
RSK dynamics



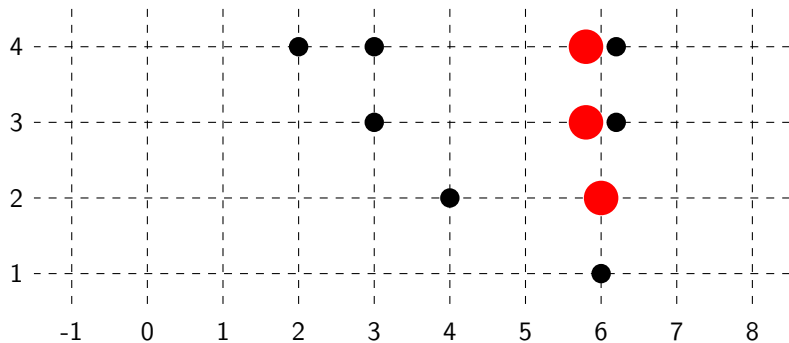
RSK dynamics



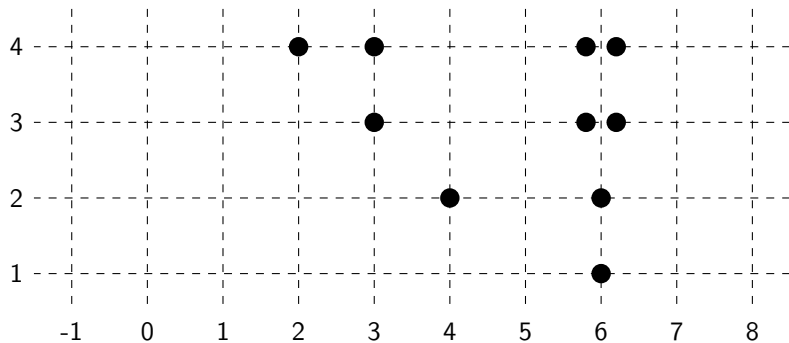
RSK dynamics



RSK dynamics



RSK dynamics



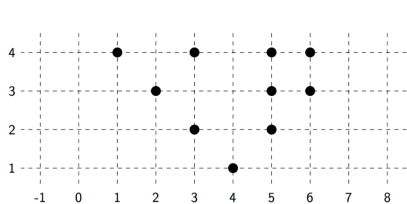
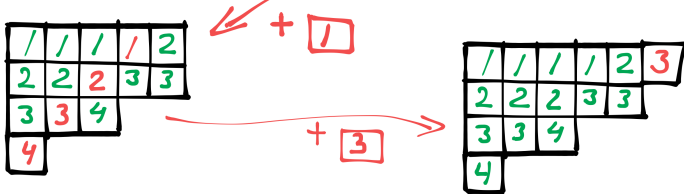
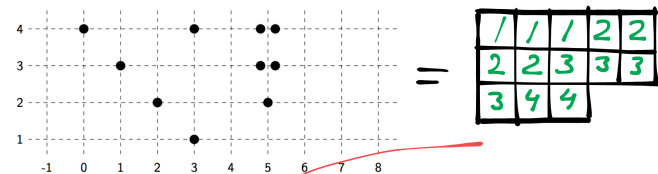
Remark: Classical RSK insertion

RSK is a bijection between words in the alphabet $\{1, 2, \dots, N\}$ and pairs of Young tableaux (one semistandard and one standard) of the same shape.

Interlacing arrays \longleftrightarrow
semistandard Young tableaux (“ P -tableaux”)

Independent jump at level h \longleftrightarrow
RSK-insert letter h into the tableau.

Remark: Classical RSK insertion

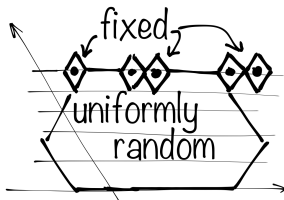


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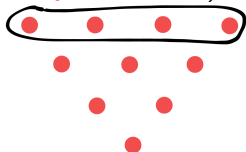
- ① Push-block dynamics
- ② Robinson–Schensted–Knuth (RSK) dynamics
- ③ Common properties of the two dynamics

Common properties of both dynamics

- ① “Interaction goes up”: evolution of $\{\lambda^{(1)}, \dots, \lambda^{(h)}\}$ is independent of $\{\lambda^{(h+1)}, \dots, \lambda^{(N)}\}$.
- ② Preserve the class of Gibbs measures

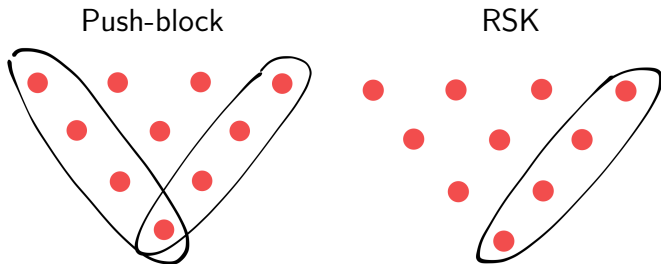


- ③ Started from a Gibbs measure, each row evolves according to the dynamics of noncolliding Poisson processes (“discrete Dyson’s BM”)



Common properties of both dynamics

- Started from $\lambda_j^{(k)} = 0$, at time t both dynamics have the same distribution.
- Markovian projections:

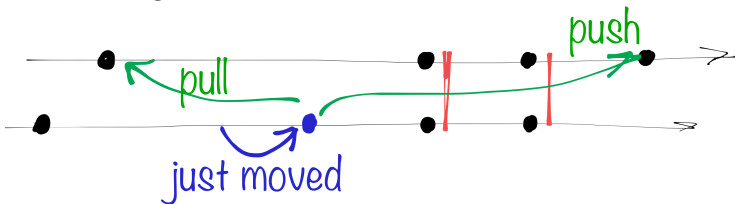


Nearest neighbor dynamics

Nearest neighbor dynamics

We look for other dynamics which satisfy:

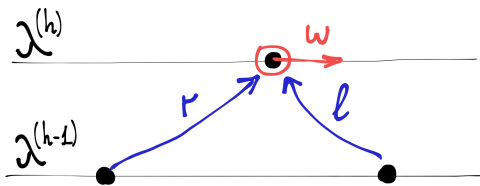
- ① “Interaction goes up”
- ② Preserve Gibbs measures
- ③ “discrete Dyson’s BM” on floors
- ④ Nearest neighbor interactions:



(push/pull with some probabilities, do nothing with the complementary probability)

[Borodin-P. '13] — introduce these axioms, and obtain complete classification of nearest neighbor dynamics.

Nearest neighbor dynamics



independent jump rate
 $w = w(\lambda^{(h-1)}, \lambda^{(h)})$

pushing probabilities (after lower particle jumped)
 $r = r(\lambda^{(h-1)}, \lambda^{(h)})$
and $l = l(\lambda^{(h-1)}, \lambda^{(h)})$

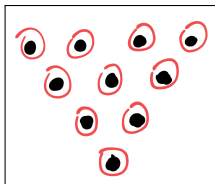
Theorem [Borodin-P. '13]. Nearest neighbor dynamics correspond to solutions of the equations

$$r(\lambda^{(h-1)}, \lambda^{(h)}) + l(\lambda^{(h-1)}, \lambda^{(h)}) + w(\lambda^{(h-1)}, \lambda^{(h)}) = 1$$

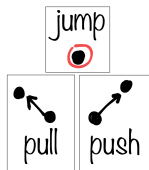
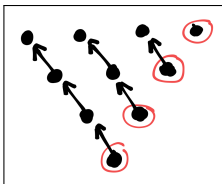
written for all states $\lambda^{(1)}, \dots, \lambda^{(N)}$ of the array and each particle in it.

“Basis” dynamics are encoded by pictures such as:

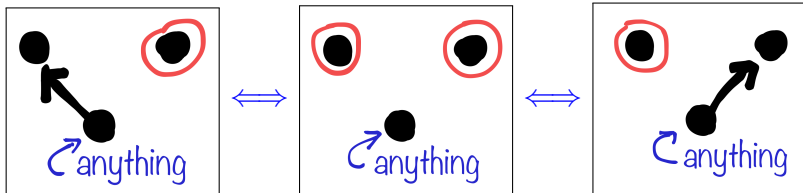
Push-block:



RSK:



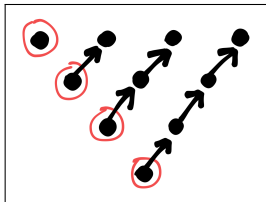
plus local flips



All other dynamics are linear combinations of “basis” ones

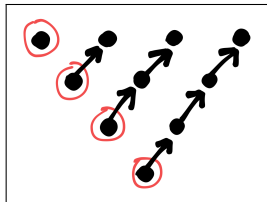
“Basis” nearest neighbor dynamics examples

Column (= dual) RSK [O’C ’03]

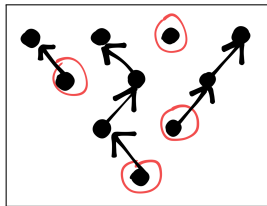


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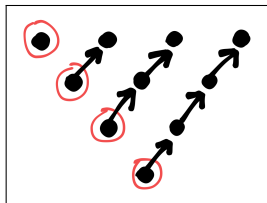


RSK-type (\Rightarrow we obtain $N!$ bijections between words and pairs of tableaux)

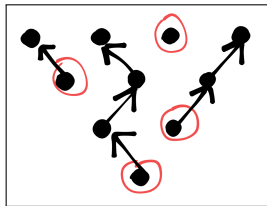


“Basis” nearest neighbor dynamics examples

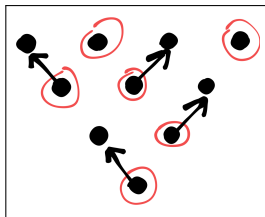
Column (= dual) RSK [O’C ’03]



RSK-type (\Rightarrow we obtain $N!$ bijections between words and pairs of tableaux)



Another example



From Schur to Macdonald

Schur polynomials in dynamics on interlacing arrays

Schur polynomials:

$$s_{\mu}(x_1, \dots, x_k) = \frac{\det \left[x_i^{\mu_j + N - j} \right]_{i,j=1}^k}{\det \left[x_i^{N-j} \right]_{i,j=1}^k}, \text{ where } \mu_1 \geq \dots \geq \mu_k.$$

$s_{\lambda^{(k)}/\lambda^{(k-1)}}$ — skew Schur polynomials.

Distribution of the dynamics — **Schur process**

[Okounkov–Reshetikhin '03]:

$$\mathbb{P}_t \left[\begin{array}{ccc} \circ & \circ & \dots & \circ \\ \circ & \circ & \dots & \circ \\ \vdots & & & \\ \circ & & & \end{array} \rightarrow \begin{array}{ccc} \lambda_N^{(N)} & \dots & \lambda_1^{(N)} \\ \vdots & & \vdots \\ \lambda_1^{(1)} & & \end{array} \right] = \frac{1}{Z} S_{\lambda^{(1)}}(a_1) S_{\lambda^{(2)}/\lambda^{(1)}}(a_2) \dots S_{\lambda^{(N)}/\lambda^{(N-1)}}(a_N) S_{\lambda^{(N)}}(\rho_t)$$

$a_1 = \dots = a_k = 1$
 $(a_k = \text{speed at level } k)$

„Plancherel specialization”,
 $t = \text{time}$

It is a **determinantal point process**, which is the source of integrability of the model.

Macdonald polynomials

$P_\lambda(x_1, \dots, x_N) \in \mathbb{Q}(q, t)[x_1, \dots, x_N]^{S(N)}$ labeled by partitions $\lambda = (\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_N \geq 0)$ form a basis in symmetric polynomials in N variables over $\mathbb{Q}(q, t)$. They diagonalize

$$\mathcal{D}_1 = \sum_{i=1}^N \prod_{j \neq i} \frac{tx_i - x_j}{x_i - x_j} T_{q, x_i}, \quad (T_q f)(z) := f(zq),$$

with (generically) pairwise different eigenvalues

$$\mathcal{D}_1 P_\lambda = (q^{\lambda_1} t^{N-1} + q^{\lambda_2} t^{N-2} + \dots + q^{\lambda_N}) P_\lambda.$$

Macdonald polynomials have many remarkable properties (similar to those of **Schur polynomials** corresponding to $q = t$) including orthogonality, simple reproducing kernel (Cauchy identity), Pieri and branching rules, index/variable duality, etc. There are also simple higher order Macdonald difference operators commuting with \mathcal{D}_1 .

From Schur to Macdonald

In short, replace all Schur polynomials by Macdonald polynomials. All previous constructions of dynamics work.

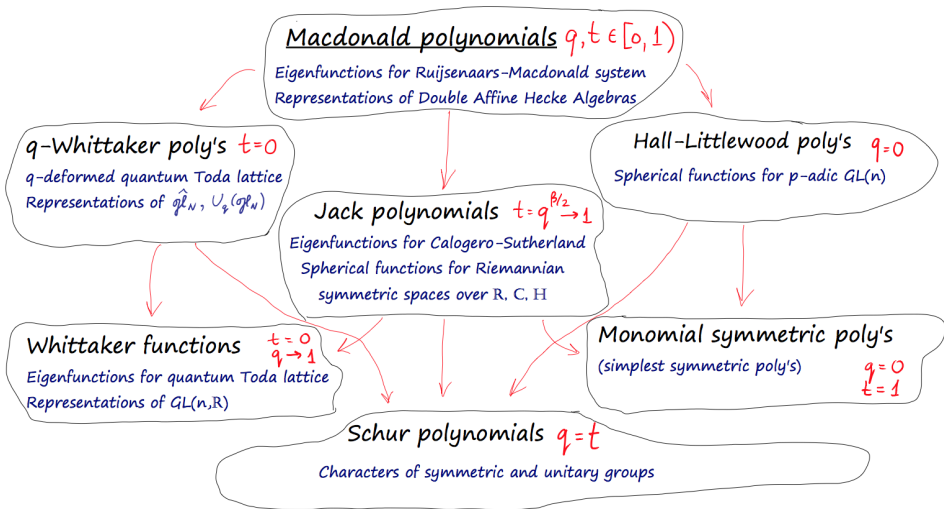
Get Markov dynamics on interlacing arrays whose distributions are **Macdonald processes** [Borodin–Corwin '11], [Borodin–Corwin–Gorin–Shakirov '13]:

$$\mathbb{P}_t \left[\begin{array}{ccc} 0 & 0 & \dots & 0 \\ & 0 & \dots & 0 \\ & & \ddots & \\ & & & 0 \end{array} \rightarrow \begin{array}{ccc} \lambda_N^{(N)} & \dots & \lambda_1^{(1)} \\ \vdots & & \vdots \\ \lambda_1^{(1)} & & \end{array} \right] = \frac{1}{Z} \mathbf{P}_{\lambda}^{(N)}(a_1) \mathbf{P}_{\lambda^{(2)}/\lambda^{(1)}}(a_2) \dots \mathbf{P}_{\lambda^{(N)}/\lambda^{(N-1)}}(a_N) \mathbf{Q}_{\lambda^{(N)}}(\rho_{\pm})$$

Macdonald polynomials, ordinary and skew
dual basis
Plancherel specialization, t = time
 $a_1 = \dots = a_k = 1$

[Borodin–Corwin '11], [O'Connell–Pei '12],
[Borodin–P. '13] (complete classification of these dynamics)

Symmetric polynomials and related objects

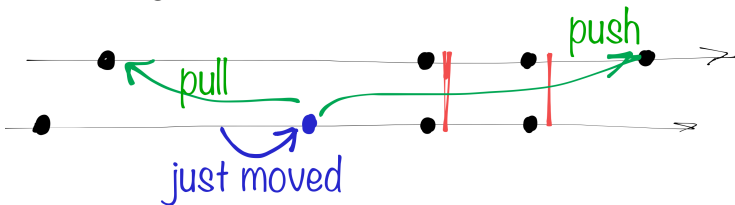


Nearest neighbor dynamics on Macdonald processes:

- ① Axioms
- ② ($t = 0$) Push-block dynamics and q -TASEP
- ③ ($t = 0$) q -PushTASEP — a new q -deformed 1d particle system
- ④ ($q = 0$) Random triangular matrices over a finite field

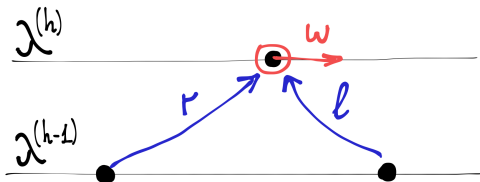
Nearest neighbor dynamics on Macdonald processes

- ① “Interaction goes up”
- ② Preserve (q, t) -Gibbs measures
- ③ (q, t) -“discrete Dyson’s BM” on floors
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Nearest neighbor dynamics on Macdonald processes



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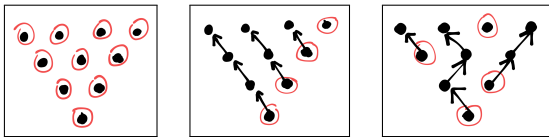
Theorem [Borodin-P. '13]. Nearest neighbor dynamics on Macdonald processes correspond to solutions of the equations

$$T \cdot r(\lambda^{(h-1)}, \lambda^{(h)}) + \tilde{T} \cdot l(\lambda^{(h-1)}, \lambda^{(h)}) + w(\lambda^{(h-1)}, \lambda^{(h)}) = S.$$

Here T, \tilde{T}, S are certain coefficients depending on q, t , and also on $\lambda^{(h-1)}, \lambda^{(h)}$.

Nearest neighbor dynamics on Macdonald processes

The “basis” nearest neighbor dynamics are encoded by **the same** pictures as before.



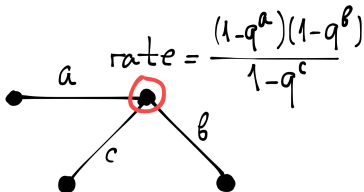
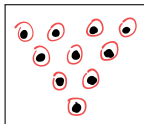
Not all of the “Schur level” pictures lead to dynamics with nonnegative jump rates. We have to speak about *formal* Markov processes.

Nearest neighbor dynamics on Macdonald processes:

- ① Axioms
- ② ($t = 0$) Push-block dynamics and q -TASEP
- ③ ($t = 0$) q -PushTASEP — a new q -deformed 1d particle system
- ④ ($q = 0$) Random triangular matrices over a finite field

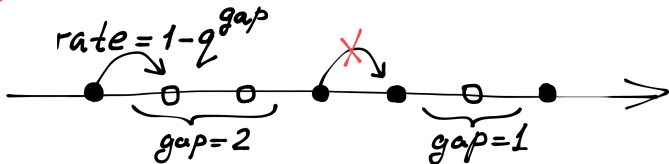
Push-block dynamics [Borodin–Corwin '11]

Let the second Macdonald parameter $t = 0$. The push-block dynamics gives:



Markovian projection — q -TASEP [BC '11], [BC–Sasamoto '12], [O'Connell–Pei '12], [BC–P.–Sasamoto '13], [Povolotsky '13]

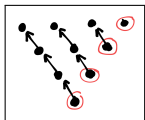
q -TASEP: $x_n = \mathcal{X}_n^{(n)} - n$



Nearest neighbor dynamics on Macdonald processes:

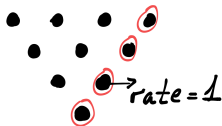
- ① Axioms
- ② ($t = 0$) Push-block dynamics and q -TASEP
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RSK-type dynamics [Borodin–P. '13]



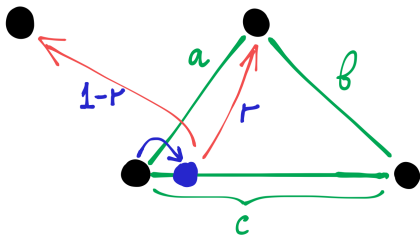
Let the second Macdonald parameter $t = 0$.
Then the q -deformation of the classical RSK is:

1. Only the rightmost particles make independent jumps with rate 1



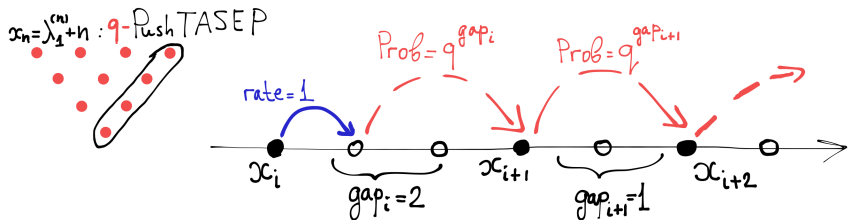
2. If a particle moves, it pushes its immediate upper neighbors with probabilities r and $1 - r$, where

$$r = q^a \frac{1 - q^b}{1 - q^c}$$

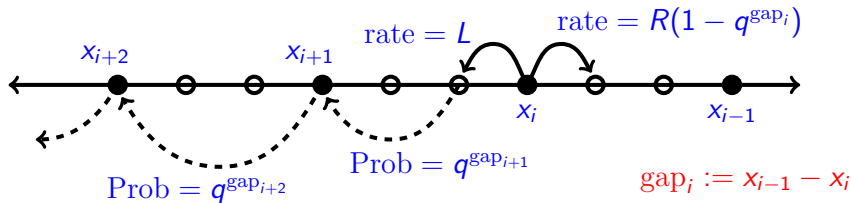


q -PushTASEP [Borodin–P. '13], [Corwin–P. '13]

Another **Markovian** projection:



Remark: q -PushASEP [Corwin–P. '13]



$R * (q\text{-TASEP, to the right}) + L * (q\text{-PushTASEP, to the left})$

Traffic model (relative to a time frame moving to the right)

- Right jump = a car *accelerates*. Chance $1 - q^{gap}$ is lower if another car is in front.
- Left jump = a car *slows down*. The car behind sees the **brake lights**, and may also quickly slow down, with probability q^{gap} (chance is higher if the car behind is closer).

Nearest neighbor dynamics on Macdonald processes:

- ① Axioms
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- ④ ($q = 0$) Random triangular matrices over a finite field

Random triangular matrices over a finite field

Through Jordan normal form of truncations of matrices from \mathbf{U} , the problem reduces to measures μ_n on Young diagrams $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_\ell$ with fixed number n of boxes. The measures μ_n are related to Hall–Littlewood polynomials (these are Macdonald polynomials with $q = 0$; and t as in F_{t-1}).

Conjectural classification of measures μ on \mathbf{U} [Kerov '03]:
measures depend on parameters

$$\begin{aligned} \alpha_1 &\geq \alpha_2 \geq \dots \geq 0; & \sum_{i=1}^{\infty} \left(\alpha_i + \frac{\beta_i}{1-t} \right) &\leq 1. \\ \beta_1 &\geq \beta_2 \geq \dots \geq 0; \end{aligned}$$

These measures $\mu^{\alpha;\beta}$ exist and are ergodic.

The problem is to show the completeness of classification.

See [Gorin–Kerov–Vershik '12].

Random triangular matrices over a finite field

We construct a randomized RSK to sample these ergodic measures. Input of the RSK is a random Bernoulli word.

Using this RSK, we prove another conjecture of Vershik–Kerov — a law of large numbers for the measures $\mu_n^{\alpha;\beta}$ ($t = 0$ — infinite symmetric group)

Theorem

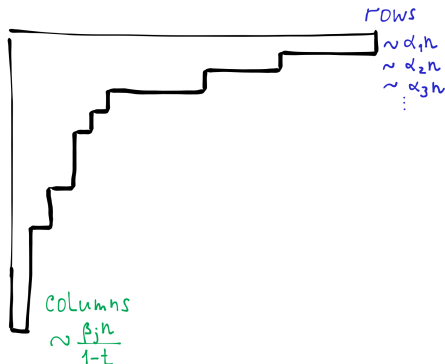
[Bufetov–P., in progress].

For random Young diagrams distributed according to $\mu_n^{\alpha;\beta}$,

as $n \rightarrow \infty$:

$$\frac{\text{row}(i)}{n} \rightarrow \alpha_i$$

$$\frac{\text{column}(j)}{n} \rightarrow \frac{\beta_j}{1-t}$$



Conclusion

