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# Asymptotic determinism of Robinson-Schensted-Knuth algorithm joint work with Dan Romik 

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## Robinson-Schensted-Knuth algorithm — induction step


$\mathrm{x}=(23,53,74,5,99,69,82,37,41)$

## Robinson-Schensted-Knuth algorithm — induction step



$$
\mathbf{x}=(23,53,74,5,99,69,82,37,41,18)
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$$
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## Robinson-Schensted-Knuth algorithm

 ($x=\emptyset$

## Robinson-Schensted-Knuth algorithm


$x=(23)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53)$

## Robinson-Schensted-Knuth algorithm


$\mathrm{x}=(23,53,74)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69,82)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69,82,37)$

## Robinson-Schensted-Knuth algorithm


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$\mathrm{x}=(23,53,74,5,99,69,82,37,41,18,39)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69,82,37,41,18,39,61)$

## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69,82,37,41,18,39,61,73)$

## Robinson-Schensted-Knuth algorithm


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## Robinson-Schensted-Knuth algorithm


$x=(23,53,74,5,99,69,82,37,41,18,39,61,73,66,22)$

## outlook

- $x_{1}, x_{2}, \ldots$ independent random variables with uniform distribution on the interval $[0,1]$;
- insertion tableau $P_{m}=P\left(x_{1}, \ldots, x_{m}\right)$;


## General problem

What can we say about (the time evolution of) the insertion tableau $P_{m}$ ?
"with the right scaling of time and space, the answer is deterministic (asymptotically)"

## diffusion of a box

- $x_{n}\left(P_{m}\right)$ denotes the location of the box containing $x_{n}$ in the insertion tableau $P_{m}$, for $m \geq n$;


## Concrete problem 1

Suppose that $n$ and $x_{n}$ are known; what can we say about the time evolution of $x_{n}\left(P_{m}\right)$ for $m=n, n+1, \ldots$ ?
asymptotic determinism of this and that
the key result: new box
proof of the key result 0 000 00
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## diffusion of a box

- $x_{n}\left(P_{m}\right)$ denotes the location of the box containing $x_{n}$ in insertion tableau $P_{m}$, for $m \geq n$;


## Theorem

There exists an explicit function $G: \mathbb{R}_{+} \rightarrow \mathbb{R}_{+}^{2}$ such that

$$
\frac{x_{n}\left(P_{\left\lfloor n e^{\tau}\right\rfloor}\right)}{\sqrt{n x_{n}}} \frac{\text { in probability }}{n \rightarrow \infty} G_{\tau} \quad \text { for } \tau \geq 0 \text {. }
$$

## hydrodynamic limit of RSK

## bumping routes



$$
x=(23,53,74,5,99,69,82,37,41, \underbrace{18}_{x_{n}})
$$

## bumping routes




## Theorem

Bumping route (scaled by factor $\frac{1}{\sqrt{n x_{n}}}$ ) obtained by adding entry $x_{n}$ to the tableau $P_{n-1}$ converges in probability (as $n \rightarrow \infty$ ) to a deterministic curve $G_{\tau}$.

## new box

$$
P\left(x_{1}, \ldots, x_{n}, x_{n+1}\right) \backslash P\left(x_{1}, \ldots, x_{n}\right)=\{\square\}
$$



## Theorem

$$
\left\|\frac{\square}{\sqrt{n}}-\left(\mathrm{RSK} \cos x_{n+1}, \mathrm{RSK} \sin x_{n+1}\right)\right\| \underset{\text { in probability }}{n \rightarrow \infty} 0
$$

## new box



## new box


the key result explains the behavior of bumping routes

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## proof, part 1 - reduction of problem

instead of (for deterministic $x_{n+1}$ )

$$
P\left(x_{1}, \ldots, x_{n}, x_{n+1}\right) \backslash P\left(x_{1}, \ldots, x_{n}\right)=\{\square\}
$$



## proof, part 1 - reduction of problem

we study (for random $0<t_{1}<\cdots<t_{k}<1$ )

$$
P\left(x_{1}, \ldots, x_{n}, t_{1}, \ldots, t_{k}\right) \backslash P\left(x_{1}, \ldots, x_{n}\right)=\{1, \ldots, k
$$



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$$


if $x_{n+1}<t_{i}$ then $\square$ is north-west from $\quad \mathrm{i}$
for $\frac{i}{k} \approx x_{n+1}+\epsilon$, this happens with high probability, as $k \rightarrow \infty$

## representations of the symmetric groups

representation $\rho$ of a group $G$ is a homomorphism to matrices

$$
\rho: G \rightarrow \mathrm{GL}_{k}
$$

irreducible representation $\rho^{\lambda}$ of the symmetric group $S_{n}$

Young diagram $\lambda$ with $n$ boxes


Littlewood-Richardson coefficients

$$
\left(\rho^{\lambda} \otimes \rho^{\mu}\right) \uparrow_{S_{|\lambda|} \times S_{|\mu|}}^{S_{|\lambda|+|\mu|}}=\bigoplus_{\nu} c_{\lambda, \mu}^{\nu} \rho^{\nu}
$$

## RSK and Littlewood-Richardson coefficients

if $0 \leq x_{1}, \ldots, x_{n} \leq 1$ is a random sequence, such that

$$
\text { shape of } P\left(x_{1}, \ldots, x_{n}\right)=\lambda ;
$$

and $0 \leq t_{1}, \ldots, t_{k} \leq 1$ is a random sequence, such that

$$
\text { shape of } P\left(t_{1}, \ldots, t_{k}\right)=\mu
$$

then the random Young diagram

$$
\text { shape of } P\left(x_{1}, \ldots, x_{n}, t_{1}, \ldots, t_{k}\right)
$$

has the same distribution as random irreducible component of

$$
V^{\lambda} \otimes V^{\mu} \uparrow \begin{aligned}
& S_{n+k} \\
& S_{n} \times S_{k}
\end{aligned}
$$

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$$
\text { shape of } P\left(t_{1}, \ldots, t_{k}\right)=(k)=\begin{array}{|l|l|l|l|l|}
\hline & & & & \\
\hline
\end{array}
$$

then the random Young diagram

$$
\text { shape of } P\left(x_{1}, \ldots, x_{n}, t_{1}, \ldots, t_{k}\right)
$$

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## content of the box

$$
\text { content }(\square)=(x \text {-coordinate })-(y \text {-coordinate })
$$

## Example


c
content of Young diagram $=(-2,-1,0,0,1,1,2,3)$

## Jucys-Murphy elements

$$
X_{i}=(1, i)+(2, i)+\cdots+(i-1, i) \quad \text { for } i \in\{1, \ldots, n\}
$$

$X_{1}, \ldots, X_{n}$ are elements of the symmetric group algebra $\mathbb{C}\left(S_{n}\right)$
for any Young diagram $\lambda$ with contents $\left(c_{1}, \ldots, c_{n}\right)$ and a symmetric polynomial $P\left(x_{1}, \ldots, x_{n}\right)$

$$
\chi^{\lambda}\left(P\left(X_{1}, \ldots, X_{n}\right)\right)=\frac{\operatorname{Tr} \rho^{\lambda}\left(P\left(X_{1}, \ldots, X_{n}\right)\right)}{\operatorname{Tr} \rho^{\lambda}(1)}=?
$$

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\chi^{\lambda}\left(P\left(X_{1}, \ldots, X_{n}\right)\right)=\frac{\operatorname{Tr} \rho^{\lambda}\left(P\left(X_{1}, \ldots, X_{n}\right)\right)}{\operatorname{Tr} \rho^{\lambda}(1)}=P\left(c_{1}, \ldots, c_{n}\right)
$$

## growth of Young diagrams and Jucys-Murphy elements

let $\lambda \vdash n, \mu \vdash k$ be fixed Young diagrams
let $\Gamma$ be a random irreducible component of $V^{\lambda} \otimes V^{\mu} \uparrow{ }_{S_{n} \times S_{k}}^{S_{n+k}}$
let $c_{n+1}, \ldots, c_{n+k}$ be the contents of boxes of $\Gamma \backslash \lambda$
then for any symmetric polynomial $P\left(x_{n+1}, \ldots, x_{n+k}\right)$ we have

$$
\begin{gathered}
\left(\chi^{\lambda} \otimes \chi^{\mu}\right)\left(P\left(X_{n+1}, \ldots, X_{n+k}\right) \downarrow_{S_{n+k}}^{S_{n}}, \begin{array}{c}
S_{k}
\end{array}\right) \\
=\mathbb{E} P\left(c_{n+1}, \ldots, c_{n+k}\right)
\end{gathered}
$$

## proof, part 2

if $k \approx \sqrt[4]{n}$

$$
\frac{1}{k}\left(\delta_{\frac{c_{1}}{\sqrt{n}}}+\cdots+\delta_{\frac{c_{k}}{\sqrt{n}}}\right) \frac{\text { in probability }}{n \rightarrow \infty} \mu_{\mathrm{SC}}=\underbrace{}_{-2}
$$

where $c_{i}=c(i)$
Hint: $p$-th moment of the left-hand-side

$$
\frac{1}{k} \sum_{j}\left(\frac{c_{j}}{\sqrt{n}}\right)^{p}
$$

is a random variable, show that the mean converges to $p$-th moment of $\mu_{\mathrm{SC}}$ show that the variance converges to zero

## proof, part 2

if $k \approx \sqrt[4]{n}$

$$
\frac{1}{k}\left(\delta_{\frac{c_{1}}{\sqrt{n}}}+\cdots+\delta_{\frac{c_{k}}{\sqrt{n}}}\right) \frac{\text { in probability }}{n \rightarrow \infty} \mu_{\mathrm{SC}}=\prod_{-2}
$$

where $c_{i}=c(\boxed{\mathrm{i}})$
since $c_{1}<\cdots<c_{k}$, this implies that if $\frac{i}{k} \rightarrow x_{n+1}$ then

$$
\frac{c(\sqrt{\mathrm{i}})}{\sqrt{n}} \stackrel{\text { in probability }}{\longrightarrow} F_{\mu_{\mathrm{SC}}}^{-1}\left(x_{n+1}\right)
$$

## proof, part 3



shape of $P_{n}$ (scaled by factor $\frac{1}{\sqrt{n}}$ ) with high probability concentrates around some explicit shape<br>Logan, Shepp, Vershik, Kerov

$\frac{\square}{\sqrt{n}}$ is with high probability close to the boundary of this limit shape

## further reading



Dan Romik, Piotr Śniady
Jeu de taquin dynamics on infinite Young tableaux and second class particles
Annals of Probability, to appear arXiv:1111.0575
围 Dan Romik, Piotr Śniady
Limit shapes of bumping routes in the Robinson-Schensted correspondence
arXiv:1304.7589

