

# Random Young diagrams and tableaux

Valentin Féray

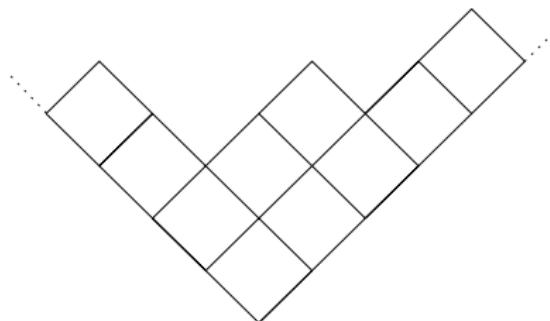
(based on joint work with J. Borga, C. Boutillier and P.-L. Méliot)

CNRS, Institut Élie Cartan de Lorraine (IECL)

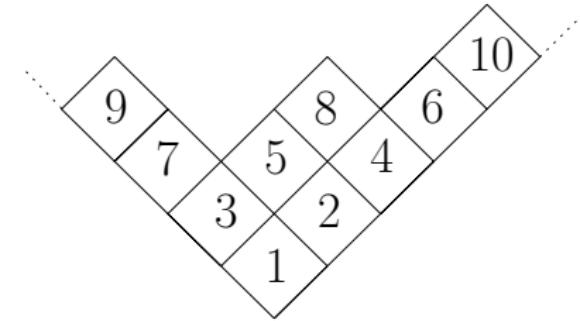
Deuxièmes rencontres LOUCCOUUM  
Poitiers, novembre 2025



# Young diagrams and tableaux



Young diagram

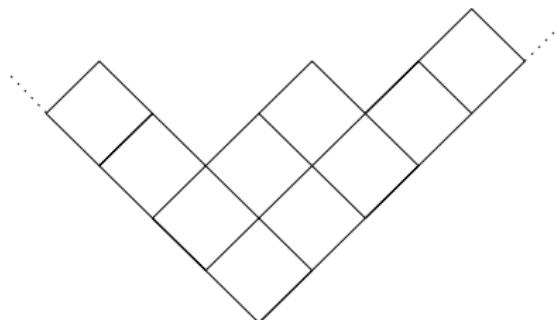


(Standard) Young tableau

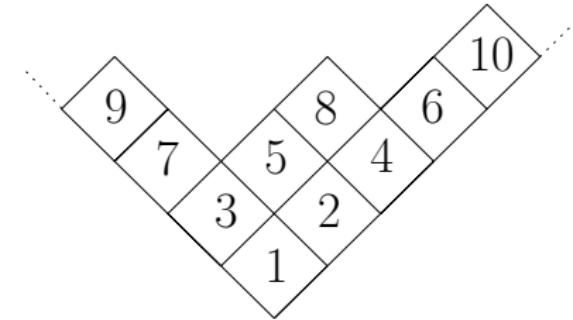
**Young diagram**: stack of boxes in the upper quarter-plane (encodes an integer partition).

**Young tableau**: filling of a Young diagram with integers from 1 to  $n$ , increasing upwards (encodes a growing sequence of tableaux).

# Young diagrams and tableaux



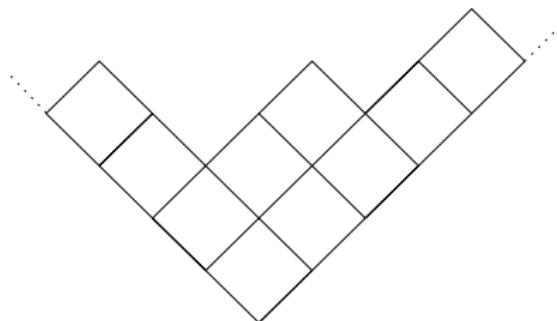
Young diagram



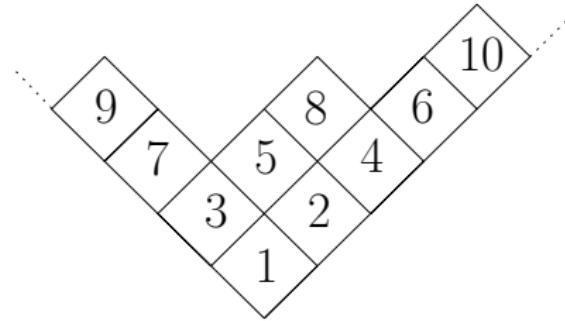
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Standard object in algebraic combinatorics (symmetric group representation, symmetric functions, . . . )  
→ yields tractable models of random walks and random surfaces.

# Young diagrams and tableaux



Young diagram



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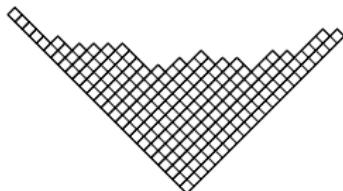
→ yields tractable models of **random walks and random surfaces**.

**First part:** survey some connections with random matrices/random walks

**Second part:** I'll focus on some recent work with Borga–Boutillier–Méliot on random tableaux.

# A well-known connection with random matrices: edge asymptotics

Plancherel measure on diagrams



For a partition  $\lambda$ , we take

$$\mathbb{P}(\lambda) = \frac{\dim(\lambda)^2}{n!}$$

$\dim(\lambda)$ : number of tableaux of shape  $\lambda$ .

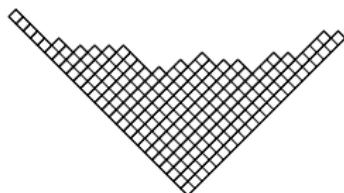
GUE model of random matrices

$$\begin{pmatrix} a_{1,1} & a_{1,2} & \cdots \\ \overline{a_{1,2}} & \ddots & \vdots \\ \vdots & \cdots & a_{n,n} \end{pmatrix}$$

Hermitian matrix with independent complex Gaussian entries above the diagonal and real Gaussian entries on the diagonal.

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**Theorem (Borodin–Okounkov–Olshanski, Okounkov, Johansson, ~'00)**

*Suitably renormalized, for all  $k$ , the first rows  $(\lambda_1, \lambda_2, \dots, \lambda_k)$  of a random Plancherel Young diagram have the same fluctuations as the largest eigenvalues of a GUE matrix (they both converge to the “Airy ensemble”).*

## Other analogies between random diagrams and random matrices

- Fluctuation of **linear statistics** ( $\sum_{i=1}^n P(\lambda_i)$  for GUE,  $\sum_{i=1}^n P(\lambda_i - i)$  for diagrams, where  $P$  is a polynomial) are described by similar Gaussian processes (Johansson '98, Kerov–Ivanov–Olshanski '03).

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- **Bulk fluctuations** are described by the sine (resp. discrete sine) processes (Dyson '70, Borodin–Okounkov–Olshanski '00).

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- **Bulk fluctuations** are described by the sine (resp. discrete sine) processes (Dyson '70, Borodin–Okounkov–Olshanski '00).
- "Fixed dimension version" (next slides).

## Fixed dimension

Fix an integer  $d \geq 1$  and consider a Plancherel random Young diagram **conditioned to have at most  $d$  rows**.

**Permutation interpretation:** we look at the RS shape of a uniform random permutation without decreasing subsequence of length  $d+1$ .

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**Theorem (Śniady, '06)**

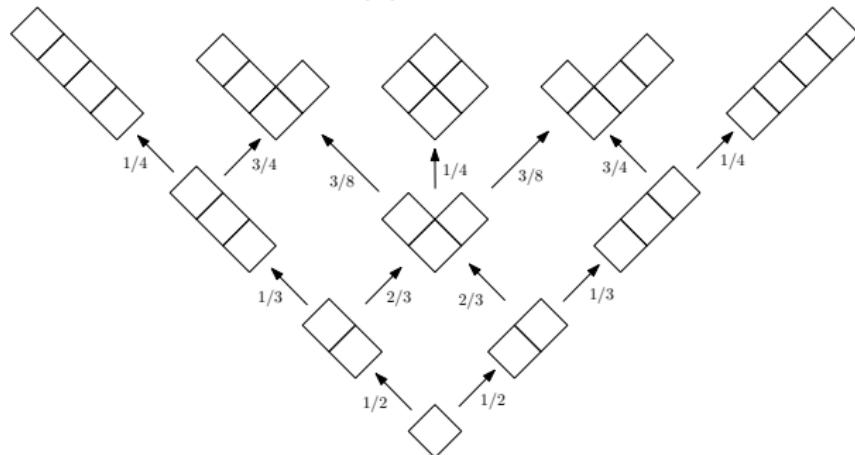
Let  $\lambda_n = (\lambda_{n,1}, \dots, \lambda_{n,d})$  be a Plancherel random Young diagram **conditioned to have at most  $d$  rows**. Then

$$\left( \sqrt{\frac{d}{n}} \left( \lambda_{n,i} - \frac{n}{d} \right) \right)_{1 \leq i \leq d}$$

converges in distribution to the eigenvalues of a **traceless GUE**  $d \times d$  random matrix.

# Introducing dynamics – the Plancherel growth process

Let  $\lambda^{(1)}, \dots, \lambda^{(n)}$  be a Markov chain of Young diagrams with  $|\lambda^{(k)}| = k$  and  $\mathbb{P}(\lambda^{(n)} = \lambda | \lambda^{(n-1)} = \mu) = \frac{\dim(\lambda)}{n \dim(\mu)}$ .



## Lemma

For each  $n$ ,  $\lambda^{(n)}$  is Plancherel distributed.

We call the sequence  $(\lambda^{(1)}, \dots, \lambda^{(n)})$  a **Plancherel random tableau**.

## Fixed dimension, dynamic version

Fix an integer  $d \geq 1$  and consider a Plancherel random tableau  $(\lambda^{(1)}, \dots, \lambda^{(n)})$  **conditioned to have at most  $d$  rows**.

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Theorem (Rizzolo, '19)

$$\left( \sqrt{\frac{d}{n}} \left( \lambda_{n,i}^{(tn)} - \frac{nt}{d} \right) \right)_{1 \leq i \leq d, 0 \leq t \leq 1}$$

converges to a traceless  $d$ -dimensional **Dyson Brownian motion**.

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Dyson Brownian motion

$\begin{cases} \text{eigenvalues of GUE dynamic matrices (entries are Brownian motions)} \\ \text{universal limit object for random walks in the cone } \{x_1 \geq \dots \geq x_d\} \end{cases}$

## $\beta$ -deformation (matrix side)

The eigenvalues of GUE random matrices have the following **density** w.r.t. **Lebesgue measure** on  $\{x_1 \geq x_2 \geq \dots \geq x_d\}$ :

$$\frac{1}{C_d} e^{-(x_1^2 + \dots + x_d^2)} \prod_{i < j} (x_i - x_j)^2.$$

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We **define  $G\beta E$  ensemble** as having the following density w.r.t. Lebesgue measure on  $\{x_1 \geq x_2 \geq \dots \geq x_d\}$ :

$$\frac{1}{C_d(\beta)} e^{-\frac{\beta}{2}(x_1^2 + \dots + x_d^2)} \prod_{i < j} (x_i - x_j)^\beta.$$

$\beta = 1, 4$ : these are eigenvalues of natural models of matrices with real/quaternionic entries.

→ huge literature on this model...

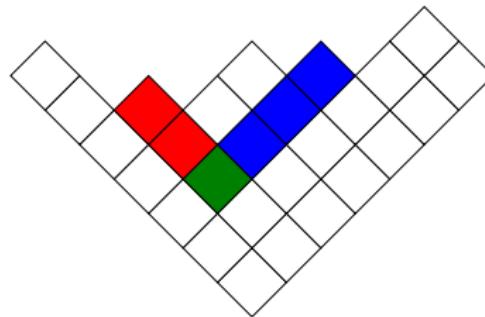
## $\beta$ -deformation (partition side)

The usual Plancherel measure is defined by

$$\mathbb{P}(\lambda) = \frac{\dim(\lambda)^2}{n!} = \frac{n!}{h_\lambda^2},$$

where

$$h_\lambda = \prod_{(i,j) \in \lambda} ((\lambda_i - j) + (\lambda'_j - i) + 1)$$



## $\beta$ -deformation (partition side)

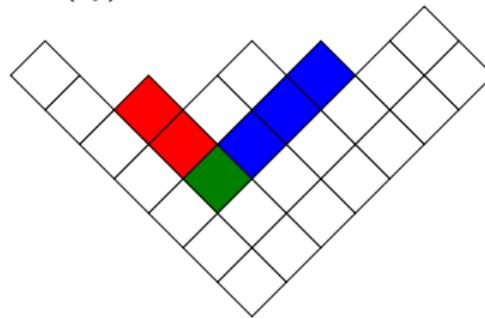
The Jack-Plancherel measure is defined by

$$\mathbb{P}(\lambda) = \frac{\alpha^n n!}{h_\lambda^{(\alpha)} h_\lambda'^{(\alpha)}},$$

where

$$h_\lambda^{(\alpha)} = \prod_{(i,j) \in \lambda} (\alpha(\lambda_i - j) + (\lambda'_j - i) + 1)$$

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## $\beta$ -deformation (partition side)

The **Jack**-Plancherel measure is defined by

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Note:  $\mathbb{P}(\lambda) = [J_\lambda^{(\alpha)}] p_1^n$ , where  $J_\lambda^{(\alpha)}$  is the (integral) Jack symmetric function indexed by  $\lambda$ . This allows in particular to define a **Jack-Plancherel growth process**.

# Analges random matrices - random partitions (general $\beta$ )

- Edge asymptotics of  $G\beta E$  and Jack-Plancherel diagrams are both described by the  $\beta$ -Tracy–Widom distributions (Valkó–Virág, '09, Guionnet–Huang '19).
- Fluctuations of linear statistics are non-centered Gaussian processes (Dimitriu – Edelman, '06, F. – Dołęga, '16).
- Fixed dimension analogies (next slides).

# A $\beta$ version of Śniady's result

Theorem (Matsumoto, '08)

Let  $\lambda_n^{(\alpha)} = (\lambda_{n,1}^{(\alpha)}, \dots, \lambda_{n,d}^{(\alpha)})$  be a *Jack-Plancherel random Young diagram* conditioned to have at most  $d$  rows. Then

$$\left( \sqrt{\frac{\alpha d}{n}} \left( \lambda_{n,i}^{(\alpha)} - \frac{n}{d} \right) \right)_{1 \leq i \leq d}$$

converges to a  $d$ -dimensional traceless *G $\beta$ E ensemble*, where  $\beta = 2/\alpha$ .

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converges to a  $d$ -dimensional traceless  $G\beta E$  ensemble, where  $\beta = 2/\alpha$ .

Permutation interpretation: for  $\alpha = 2$ ,  $\lambda_{n,1}^{(\alpha)}$  has the same distribution as the LIS of a uniform random fixed-point free involution conditionned to have no decreasing subsequence of length  $> 2d$ .

# A dynamic $\beta$ version?

## Conjecture

Let  $\lambda_n^{(\alpha),(1)}, \dots, \lambda_n^{(\alpha),(n)}$  be a Jack–Plancherel random Young tableau conditioned to have at most  $d$  rows. Then

$$\left( \sqrt{\frac{\alpha d}{n}} \left( \lambda_{n,i}^{(\alpha),(tn)} - \frac{tn}{d} \right) \right)_{1 \leq i \leq d, 0 \leq t \leq 1}$$

converges to a to a  $d$ -dimensional traceless  $\beta$ -Dyson Brownian motion.

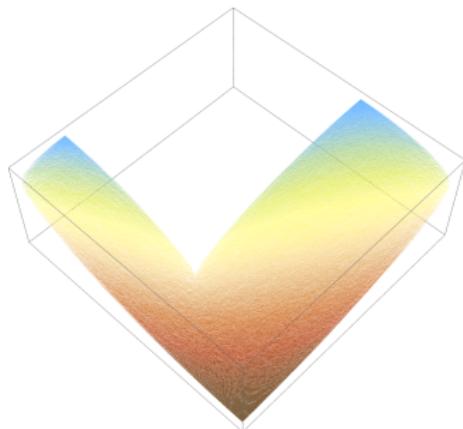
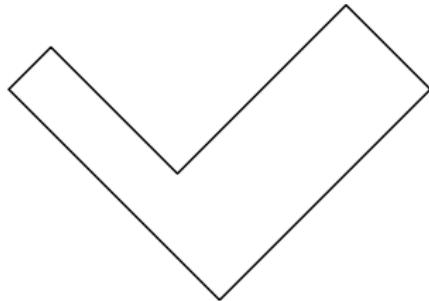
**Take away message:** algebra provides explicit models with interesting asymptotic behaviour!

# Interlude



## Second part: random tableau of fixed shape

Our model: fix a (large) Young diagram  $\lambda$  (on the left), and take a uniform random Young tableau  $T$  of shape  $\lambda$  (on the right).



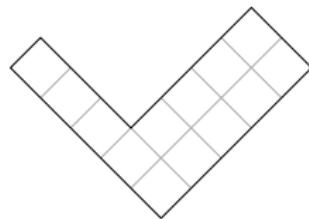
Also studied by Biane, Pittel, Romik, Angel, Holroyd, Virág, Gorin, Rahman, Linusson, Potka, Sulzgruber, Sun, Banderier, Marchal, Wallner, Śniady, Matsumoto, Maślanka, Gordenko, Xu, Prause, Raposo, ...

# Motivations

- **Bijection with other models:** constrained random permutations (RSK bijection), random sorting networks (Edelman–Greene bijection).
- **Asymptotic representation theory:** random tableaux encode some asymptotic information on restrictions of representations of large symmetric groups.
- Link with the well-studied **lozenge tiling models** (Young tableaux are in some sense a limit case of lozenge tilings);
- Tractable model of **random linear extensions** of 2-dimensional posets.

## Simulation (first example)

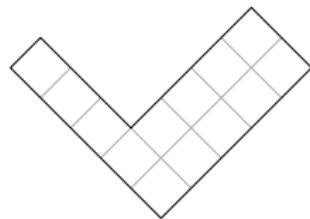
We consider the  $n$ -th dilatation  $n \cdot \lambda^0$  of the following diagram



i.e. we replace each box by a  $n \times n$  square of boxes.

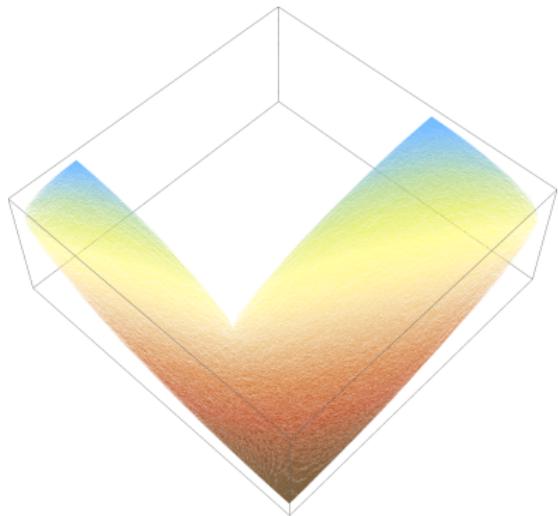
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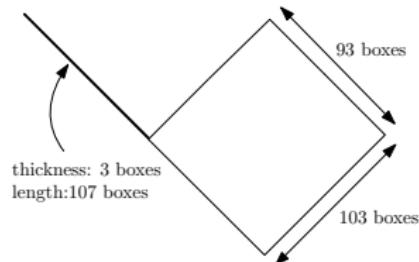
A uniform tableau  $T_N$  of shape  $n \cdot \lambda^0$ :



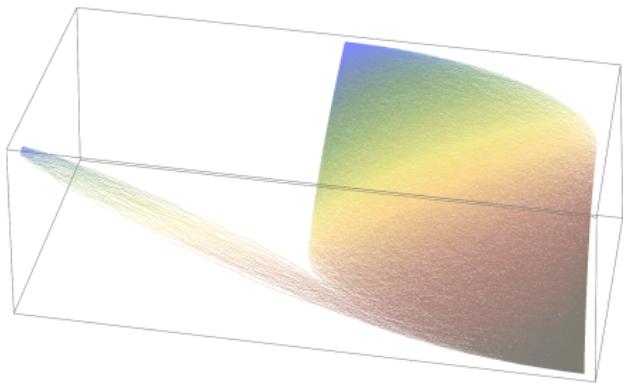
Here,  $n = 100$  so the tableau  $T_N$  has  $N = 130000$  boxes. There seems to be a smooth limit surface.

## Simulation (second example)

This time, take  $\lambda^0$  to be

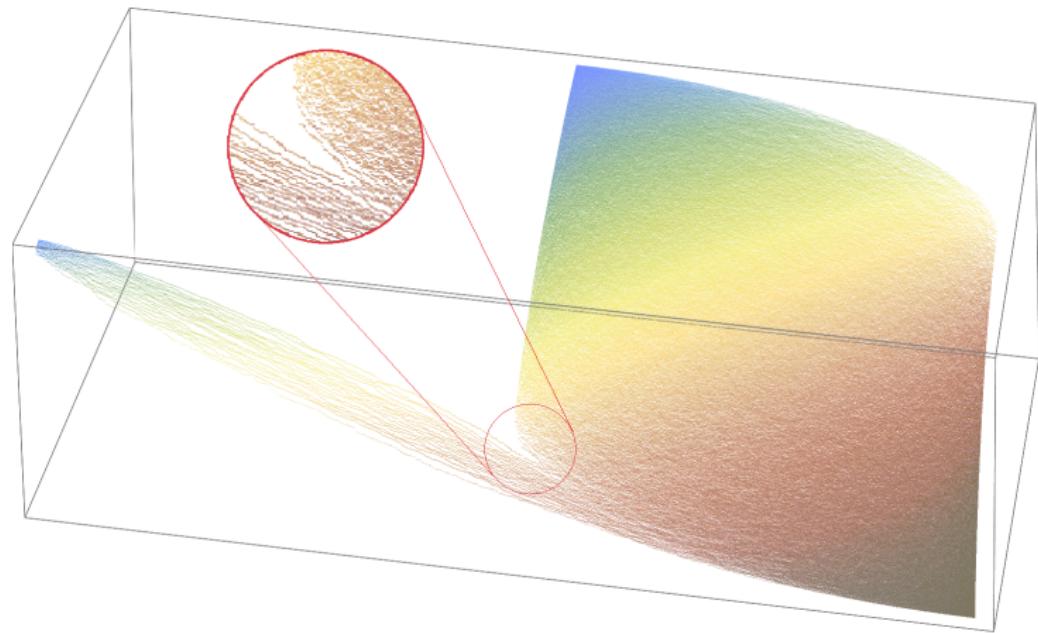


A uniform tableau  $T_N$  of shape  $n \cdot \lambda^0$ :



Here,  $n = 6$  so the diagram/tableau has  $N = 356400$  boxes.

## Simulation (second example, with a zoom)



There still seems to be a limiting surface, but this time it is discontinuous!

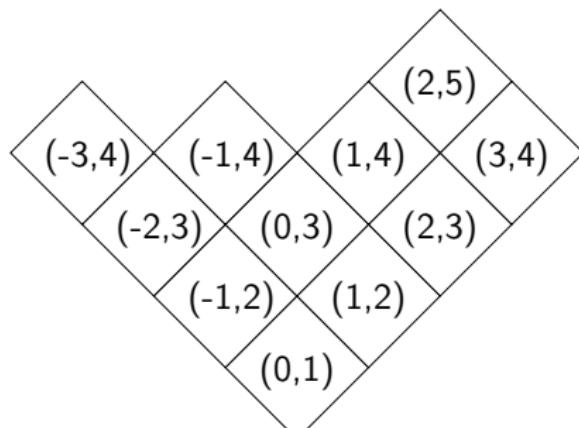
## Results (informally)

- Previous contributions (Biane '03, Sun '18): convergence to a limiting surface with some implicit description (via Markov–Krein correspondence and free compression or via a variational principle).
- Our results: a more explicit description of the limit surface in the multirectangular case (dilatation of a fixed diagram  $\lambda^0$ ) + characterization of the diagrams  $\lambda^0$  leading to discontinuous limit surfaces.

# Height function

Notation: if  $T$  is a tableau of size  $N$ , we let

- $T(x, y)$ : content of the box with coordinates  $(x, y)$  in  $T$ ;



Box coordinates

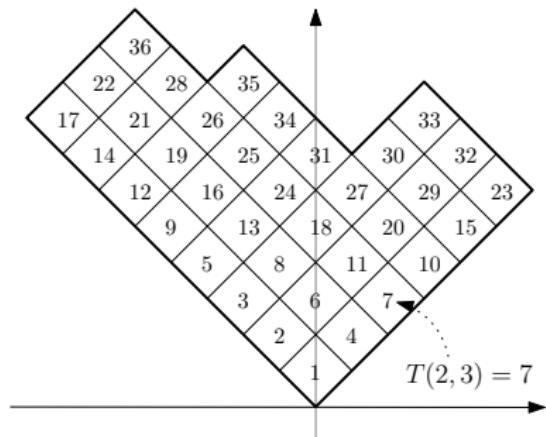
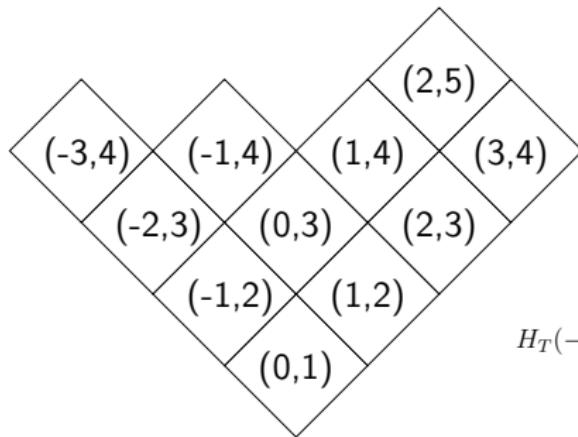


Tableau function

# Height function

Notation: if  $T$  is a tableau of size  $N$ , we let

- $T(x, y)$ : content of the box with coordinates  $(x, y)$  in  $T$ ;
- $H_T(x, t) = \#\{y : T(x, y) \leq Nt\}$ : number of entries on the vertical line  $x$  smaller than  $Nt$ .



Box coordinates

$$H_T(-3, t) = \begin{cases} 0 & \text{for } t \leq \frac{5}{36} \\ 1 & \text{for } \frac{5}{36} \leq t \leq \frac{16}{36} \\ 2 & \text{for } \frac{16}{36} \leq t \leq \frac{26}{36} \\ 3 & \text{for } \frac{26}{36} \leq t \end{cases}$$

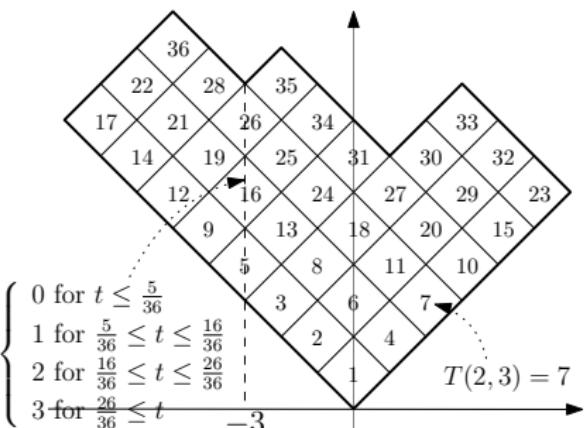


Tableau and height functions

$(y \mapsto T(x, y)$  and  $t \mapsto H_T(x, t)$  are roughly **inverses** of each other.)

# Existence of the limiting height function

Theorem (Biane '03, Sun '18)

Let  $\lambda^0$  be a fixed Young diagram. For  $n \geq 1$ , we let  $T_N$  be a uniform random Young tableau of shape  $\lambda_N := n \cdot \lambda^0$ . Then there exists a deterministic function  $H^\infty$  such that

$$\frac{1}{\sqrt{N}} H_{T_N} \left( \lfloor x\sqrt{N} \rfloor, t \right) \xrightarrow[N \rightarrow +\infty]{} H^\infty(x, t),$$

in probability, uniformly on  $(x, t)$ .

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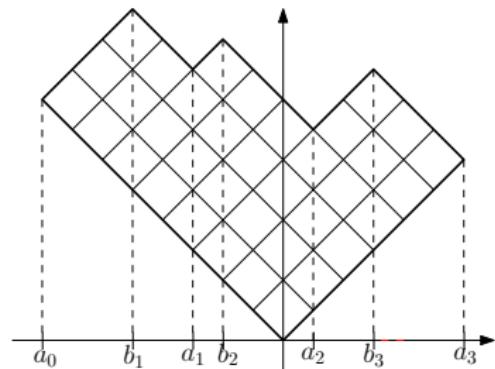
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Question: How to compute  $H^\infty$ ?

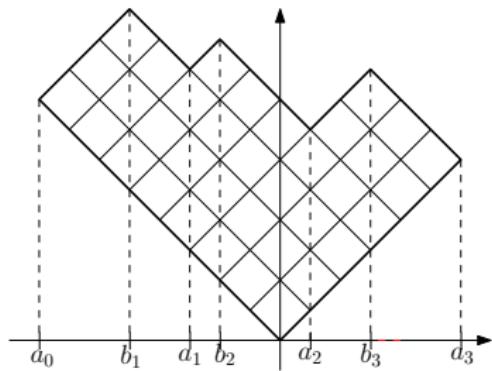
# The critical equation

We encode  $\lambda^0$  by its interlacing coordinates  $a_0 < b_1 < a_1 < \dots < b_m < a_m$ :



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**Definition: the critical equation**

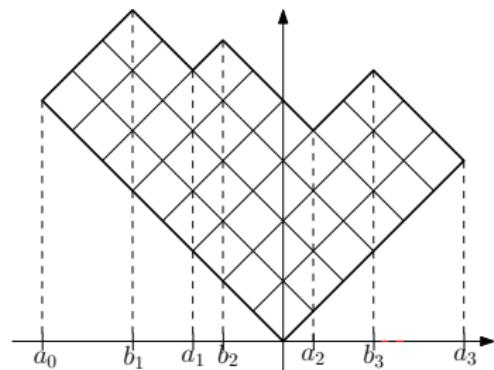
For parameters  $(x, t)$ , we consider the polynomial equation

$$\begin{aligned} U \prod_{i=1}^m (x - \eta b_i + U) \\ = (1-t) \prod_{i=0}^m (x - \eta a_i + U), \end{aligned}$$

where  $\eta = \sqrt{|\lambda^0|}$ .

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## Lemma

*The critical equation has at least  $m-1$  real roots.*

We denote  $U_c(x, t)$  its **complex root with positive imaginary part**, if it exists (in this case, we say that  $(x, t)$  is in the “liquid region”).

## Formula for the limiting height function

Theorem (Borga, Boutillier, Féray, Méliot, '23)

$$H^\infty(x, t) = \frac{1}{\pi} \int_0^t \frac{\operatorname{Im} U_c(x, s)}{1-s} ds.$$

Convention:  $\operatorname{Im} U_c(x, s) = 0$  if the critical equation has only real root ("frozen region").

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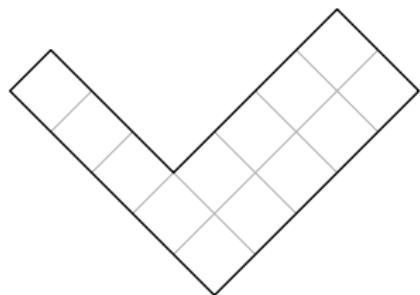
Example: square shape tableaux (Romik–Pittel, '07),  $a_0 = -1, b_1 = 0, a_1 = 1$

The critical equation  $U(x+U) = (1-t)(x+1+U)(x-1+U)$  is a second degree polynomial equation, and we get

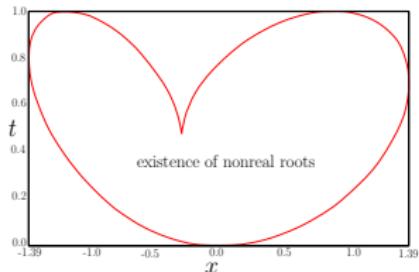
$$H_\diamond^\infty(x, t) = \frac{1}{\pi} \int_0^t \frac{\sqrt{4s - 4s^2 - x^2}}{2s - 2s^2} ds,$$

with the convention that  $\sqrt{y} = 0$  if  $y \leq 0$ .

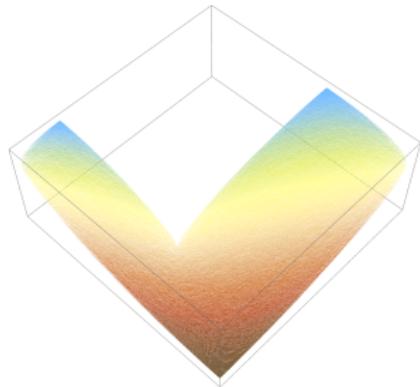
# The heart example



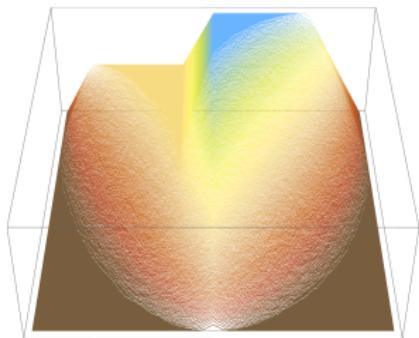
the Young diagram  $\lambda^0$



boundary of the liquid region

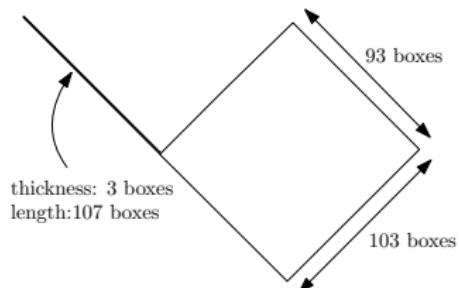


a realization of  $T_N$

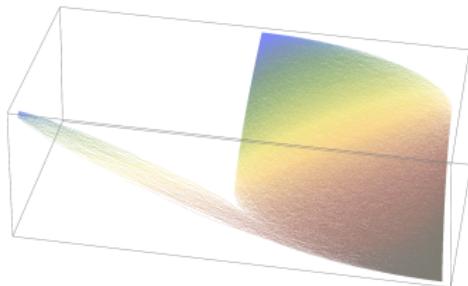


its height function  $H_{T_N}$

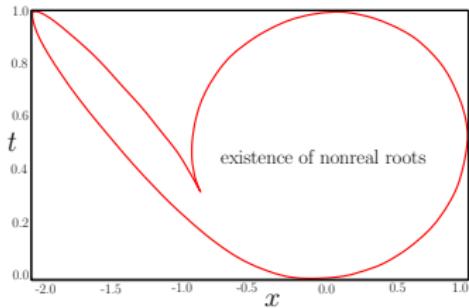
# The pipe example



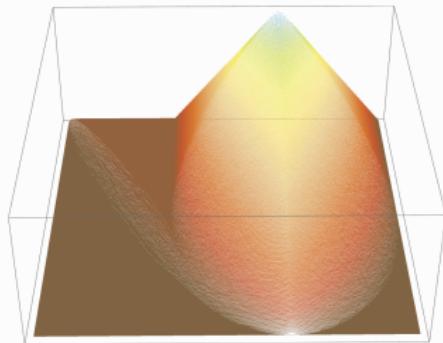
the Young diagram  $\lambda^0$



a realization of  $T_N$

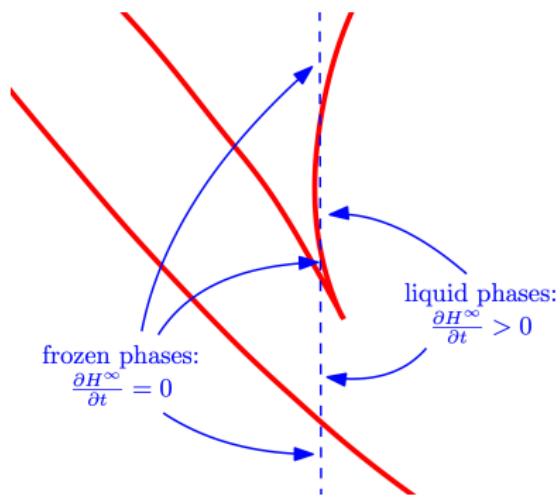


boundary of the liquid region



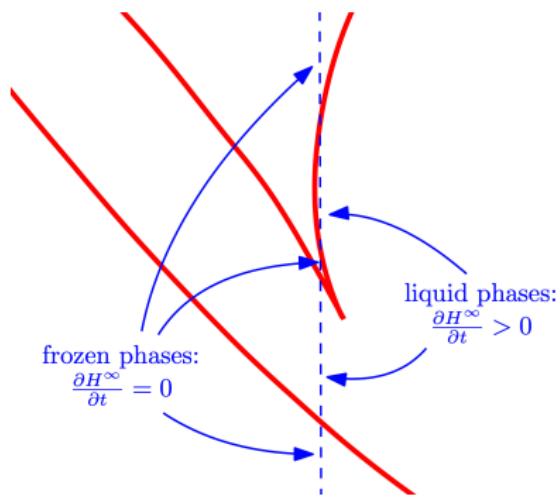
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# Why is there a discontinuity in the pipe example?

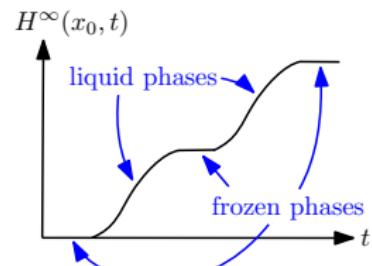


Zoom on the boundary of the liquid region (blue line  $x = x_0 \approx -0.9$ )

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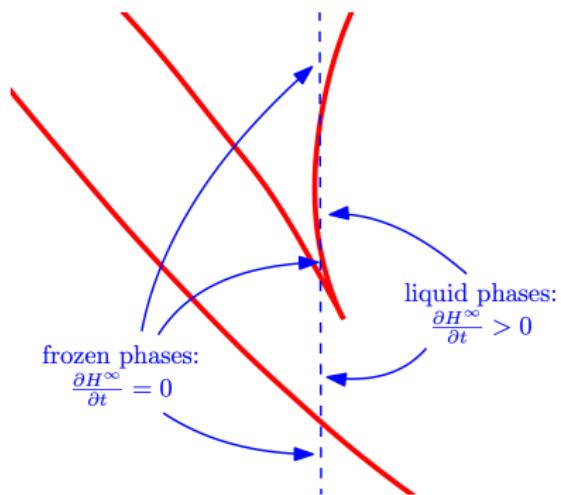


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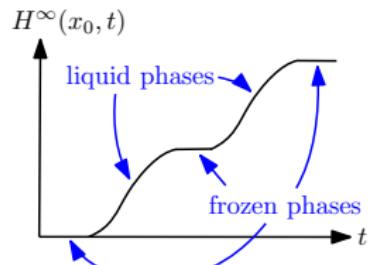


Schematic representation of the function  $t \mapsto H^\infty(x_0, t)$

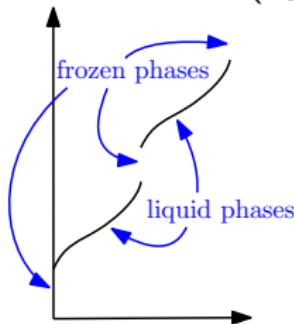
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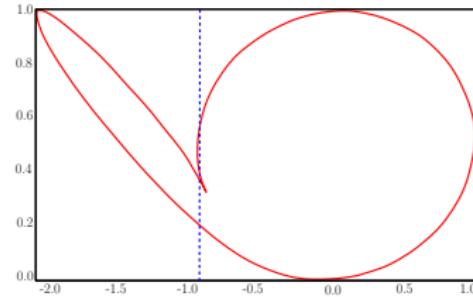
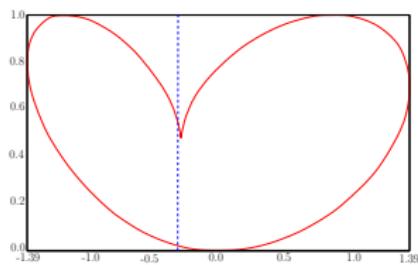
Schematic representation of the function  $t \mapsto H^\infty(x_0, t)$



Schematic representation of the function  $y \mapsto T^\infty(x_0, y)$

## When is there a discontinuity?

There is a discontinuity as soon as **the tangent at one of cusp is not vertical** (both curves leaving a cusp have the same tangent; think at  $x^2 = y^3$ ).



(In general, there are  $m-1$  cusps, where  $m$  is the number of distinct parts in  $\lambda_0$ .)

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With some computation, we get

**Theorem (Borga, Boutillier, F., Méliot, '23)**

*The limiting surface  $T_{\lambda^0}^\infty$  is continuous if and only if the interlacing coordinates  $a_0 < b_1 < a_1 < \dots < b_m < a_m$  of  $\lambda^0$  satisfy*

$$\sum_{\substack{i=0 \\ i \neq i_0}}^m \frac{1}{a_{i_0} - a_i} = \sum_{i=1}^m \frac{1}{a_{i_0} - b_i}, \quad \text{for all } i_0 = 1, \dots, m-1.$$

In particular, for  $m > 1$ , the limit surfaces are generically discontinuous!

# Proof strategy 1 – determinantal point processes

Notation:

$E$ : locally compact Polish space

$\mu$ : reference measure on  $E$

$K$ : measurable function  $E^2 \rightarrow \mathbb{C}$ .

$X$ : simple point process on  $E$

Definition (determinantal point process)

$X$  is a determinantal point process on  $E$  with kernel  $K$  if it has a joint intensity with respect to  $\mu$  given by

$$\rho_n(x_1, \dots, x_n) = \det[K(x_i, x_j)]_{1 \leq i, j \leq n},$$

for every  $n \geq 1$  and distinct  $x_1, \dots, x_n \in E$ .

Used a lot in integrable probability theory/statistical physics since 90's, but also in random matrix theory, statistics, ...

## Proof strategy 2 – tableaux and bead configurations

### Definition (Poissonized tableaux)

A Poissonized tableau of shape  $\lambda$  is an upward increasing filling of  $\lambda$  with real numbers in  $[0,1]$ .

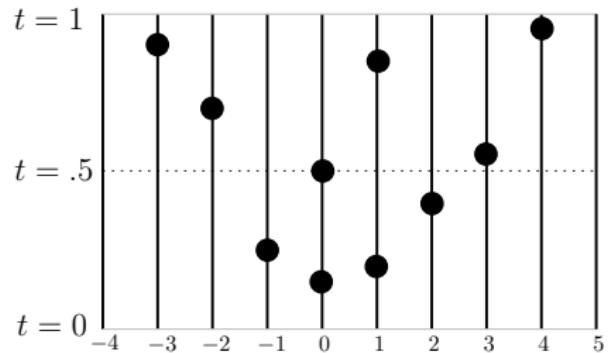
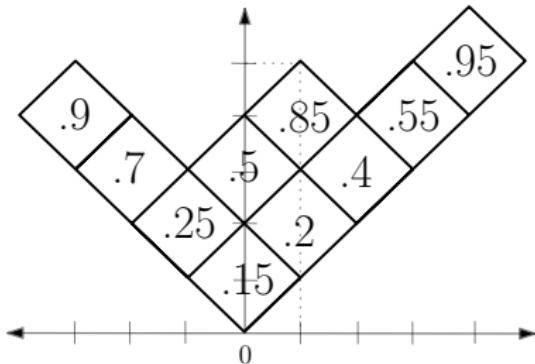
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With a Poissonized tableau  $T$ , we associate a bead configuration

$$M_T := \{(x, T(x, y)), (x, y) \in \lambda\} \subseteq \mathbb{Z} \times [0, 1].$$



Note:  $H_T(x, t)$  is the number of beads in  $\{x\} \times [0, t]$ .

## Proof strategy 3 – Gorin–Rahman theorem

Theorem (Gorin, Rahman, '19)

Let  $T$  be a uniform random Poissonized tableau of fixed shape  $\lambda$ . Then its associated bead process  $M_T$  is a determinantal point process on  $\mathbb{Z} \times [0,1]$  with correlation kernel

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$$K_\lambda((x_1, t_1), (x_2, t_2)) = -\frac{1}{(2i\pi)^2} \cdot \oint_{\gamma_z} \oint_{\gamma_w} \frac{F_\lambda(z)}{F_\lambda(w)} \frac{\Gamma(w - x_1 + 1)}{\Gamma(z - x_2 + 1)} \frac{(1 - t_2)^{z - x_2} (1 - t_1)^{-w + x_1 - 1}}{z - w} dw dz,$$

where  $F_\lambda(u) = \Gamma(u + 1) \prod_{i=1}^{\infty} \frac{u+i}{u-\lambda_i+i}$  and the double contour integral runs over counterclockwise paths  $\gamma_w$  and  $\gamma_z$  such that

- $\gamma_w$  is inside (resp. outside)  $\gamma_z$  if  $t_1 \geq t_2$  (resp.  $t_1 < t_2$ );
- $\gamma_w$  and  $\gamma_z$  contain all the integers in  $[-\ell(\lambda), x_1 - 1]$  and in  $[x_2, \lambda_1 - 1]$  respectively;
- the ratio  $\frac{1}{z-w}$  remains uniformly bounded.

## Proof strategy 4 – Rewriting the kernel

Consequence of Gorin–Rahman’s formula:

$$\mathbb{E}[H_T(x, t)] = \int_0^t K((x, s), (x, s)) ds$$

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Via Stirling approximation and standard calculus, we get

$$\frac{1}{\sqrt{N}} K((\lfloor x\sqrt{N} \rfloor, s), (\lfloor x\sqrt{N} \rfloor, s)) \approx -\frac{1}{(2i\pi)^2}.$$

$$\oint_{\gamma_Z} \oint_{\gamma_W} e^{\sqrt{N}(S(W) - S(Z))} \frac{h(W, Z)}{W - Z} dW dZ,$$

where

$$S(U) = g(U) - U \log(1 - t_0) - \sum_{i=0}^m g(x_0 - \eta a_i + U) + \sum_{i=1}^m g(x_0 - \eta b_i + U)$$

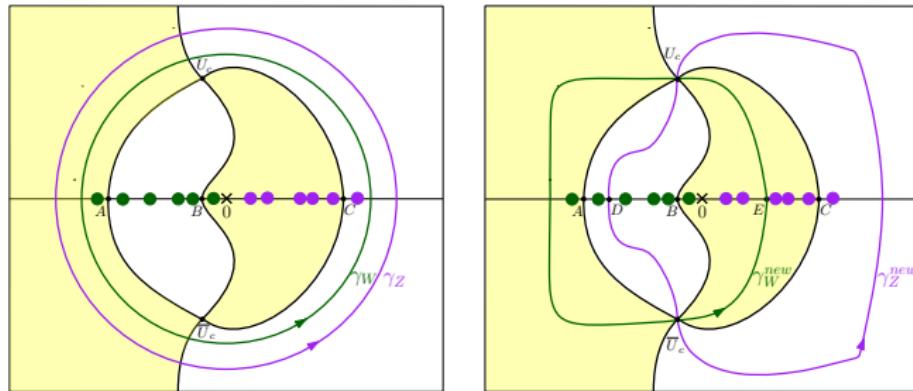
with  $g(U) = U \log(U)$  and some function  $h$ .

## Proof strategy 5 – Steepest descent analysis

Reminder: we are interested in

$$\oint_{\gamma_Z} \oint_{\gamma_W} e^{\sqrt{N}(S(W)-S(Z))} \frac{h(W, Z)}{W - Z} dW dZ.$$

Idea: deform  $\gamma_Z$  and  $\gamma_W$  such that  $\operatorname{Re}(S(W)) < \operatorname{Re}(S(Z))$  on the new contours.



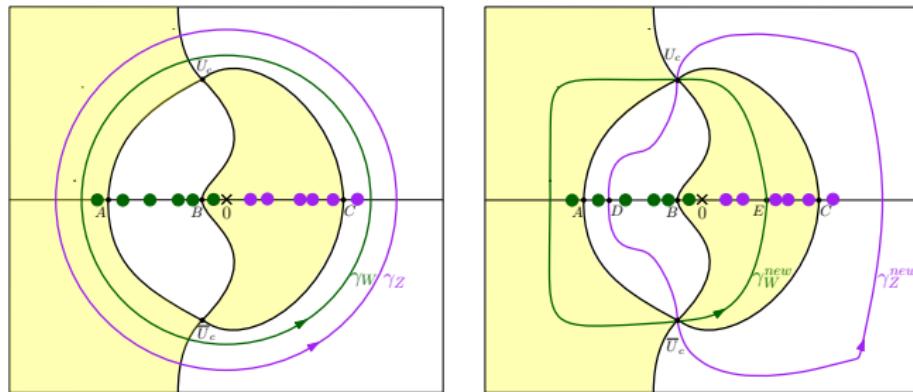
Schematic representation of the integration contours before and after transformation: in the white (resp. yellow) regions, we have  $\operatorname{Re}(S(Z)) > \operatorname{Re}(S(U_c))$  (resp.  $\operatorname{Re}(S(Z)) < \operatorname{Re}(S(U_c))$ ).

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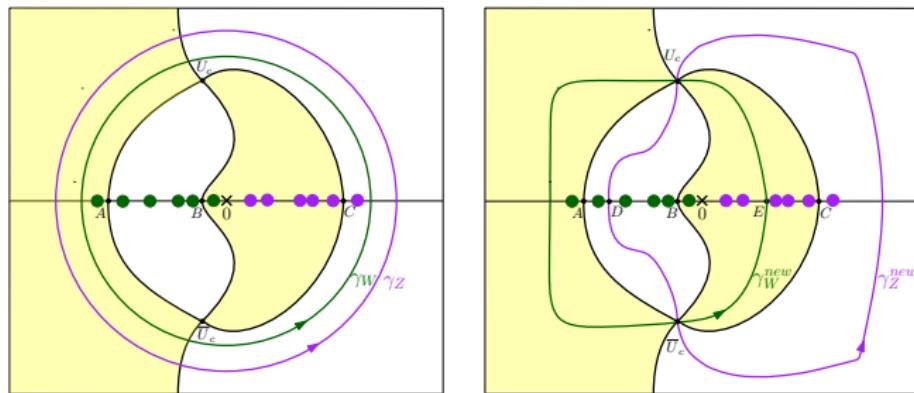
The point  $U_c$  on the above picture should satisfy  $S'(U_c) = 0$ , which is exactly the critical equation! (So the above picture is valid in the liquid region only.)

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After change of contour, the integral tends to 0. The dominant term asymptotically is the residue term for the pole  $W - Z$ , which is an integral from  $\overline{U_c}$  to  $U_c$ .

# Thanks for listening!

## Commercials:

1. LOUCCOUUM Research School (Large Objects Under Combinatorial Constraints and Outside Uniform Model) at CIRM, June 8-12, 2026.

## Mini-courses:

- [Jean-François Marckert](#) (Bordeaux): Stochastic geometry with combinatorial glasses;
- [Sumit Mukherjee](#) (Columbia): Permutons in Statistics;
- [Fiona Skerman](#) (Upsalla): Learning on random graphs,

and long talks by [Eva-Maria Heinzl](#) and [Lucas Teyssier](#).