

# Integrable probability on interlacing particles

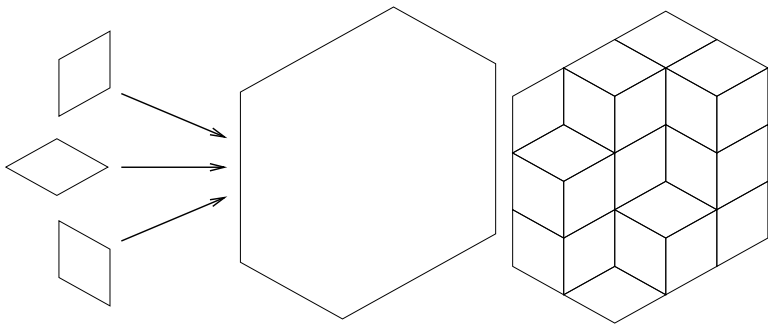
Vadim Gorin  
MIT, Cambridge and IITP, Moscow

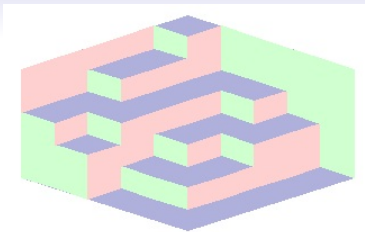
January, 2013

## Lozenge tilings: our basic model

Consider an equi-angular hexagon of side lengths  $a, b, c, a, b, c$  drawn on the regular triangular lattice.

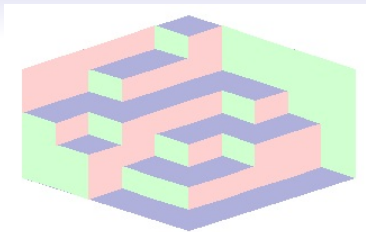
We are interested in tilings by rhombi with angles  $\pi/3$  and  $2\pi/3$  and side lengths 1. Each rhombus (“lozenge”) is a union of two elementary triangles.





## Overview

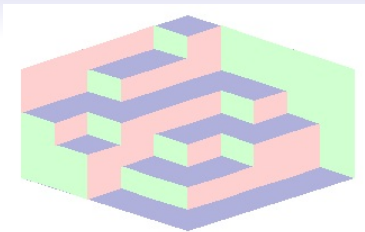
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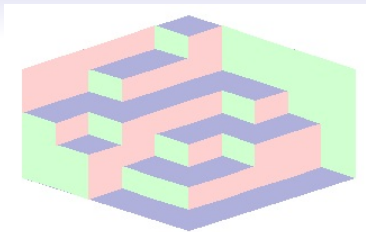


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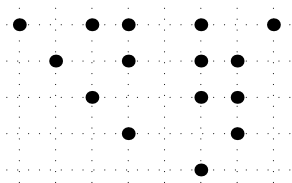
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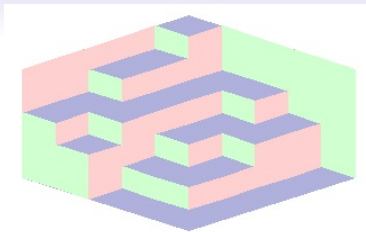
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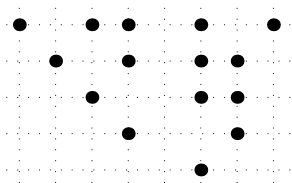
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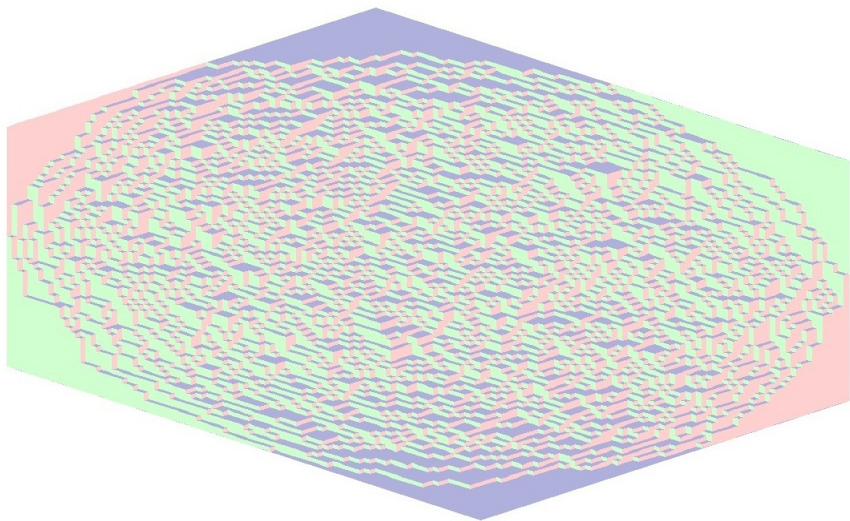


- Probability measures are “integrable” in the sense that many distributions and expectations of observables admit explicit formulas.

Consider *uniformly* random lozenge tiling of *large* hexagon. How does it look like?



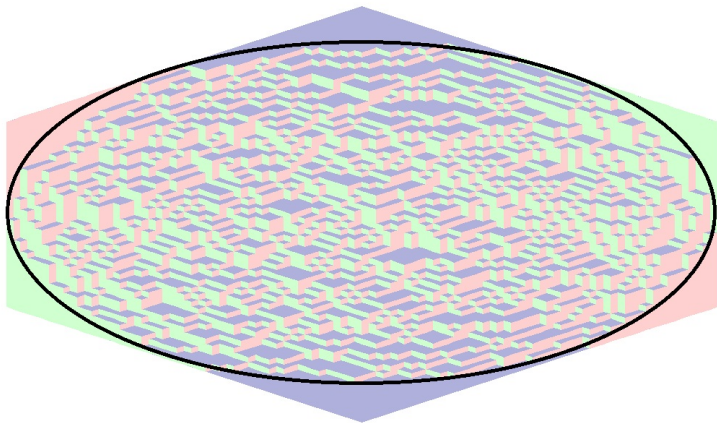
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3 types of lozenges — 3 colors 

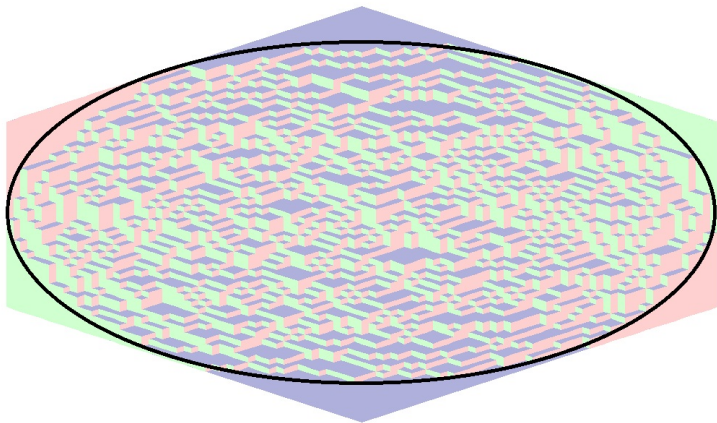
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How can this fact be proved?

## Lozenge tilings: frozen regions

**Method 1.** (Cohn–Larsen–Propp, Kenyon–Okounkov–Sheffield)  
“Limit shape” as the side lengths of the hexagon  $a, b, c \rightarrow \infty$  can be obtained as a solution of certain *variational problem*. Analyzing it via Euler–Lagrange equation one proves, in particular, the existence of frozen regions.

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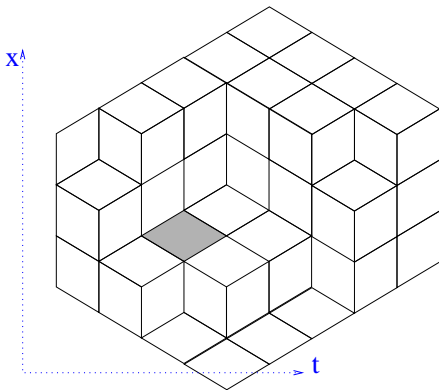
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(Kenyon–Okounkov) Tiling of a large *polygon* is asymptotically frozen outside inscribed algebraic curve of minimal degree.

**Disadvantage.** While solving the problem for the hexagon, this does not give the answer in explicit form for complicated domains.

## Lozenge tilings: frozen regions

**Method 2.** (“Integrable”) Consider *horizontal* lozenges of a tiling (they form an interlacing particle configuration).



$n$ th correlation function:

$$\rho_n((x_1, t_1) \dots (x_n, t_n)) = \text{Prob}\{\text{horiz. lozenges in given } n \text{ positions}\}$$



## Lozenge tilings: frozen regions

**Method 2.** (“Integrable”)

**Proposition.** Correlation functions for the uniform measure on tilings of a hexagon have determinantal form

$$\rho_n = \det_{i,j=1,\dots,n} \{K(t_i, x_i; t_j, x_j)\},$$

where *correlation kernel*  $K$  does not depend on  $n$ .

Thus, horizontal lozenges form a *determinantal point process*.

## Lozenge tilings: frozen regions

**Method 2A.** (Johansson–Nordenstam; Gorin) The correlation kernel  $K$  describing positions of horizontal lozenges in uniformly random lozenge tilings of a hexagon can be expressed through Hahn (hypergeometric) orthogonal polynomials.

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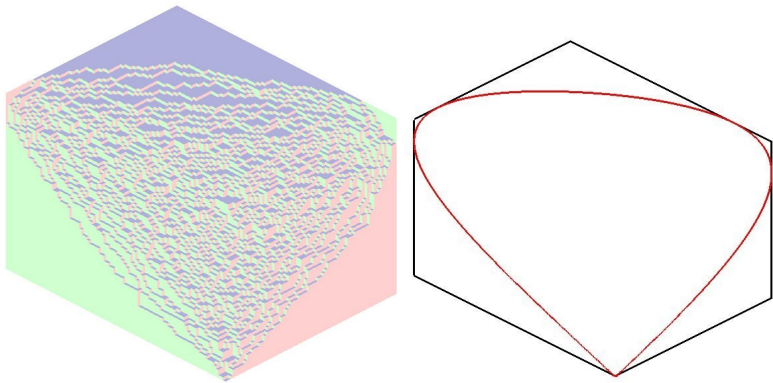
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**Advantage.** This method can be generalized to more complicated measures on tilings of hexagons. Still there are explicit formulas for the boundaries of frozen regions (Borodin–Gorin–Rains–2009).

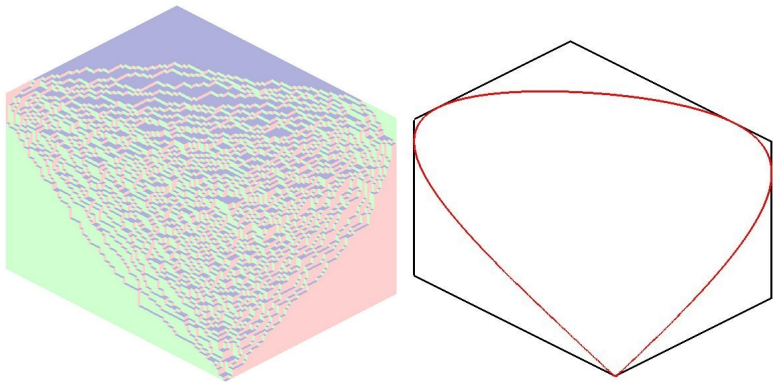
## Lozenge tilings: frozen regions

**Method 2A:** Non-uniform measures — one example.



## Lozenge tilings: frozen regions

**Method 2A:** Non-uniform measures — one example.



**Disadvantage.** Breaks down if we change the domain.

**General feature.** We can have rich models and answers, but methods are not resistant to small fluctuations.

## Lozenge tilings: frozen regions

**Method 2B.** (Petrov–2012) The correlation kernel  $K$  describing positions of horizontal lozenges in uniformly random lozenge tilings of a hexagon can be expressed as a *double contour integral* of elementary functions.

**Remark.** In the context of Schur processes related to random tilings, double contour representations for correlation kernels were found in (Okounkov–Reshetikhin-2001).

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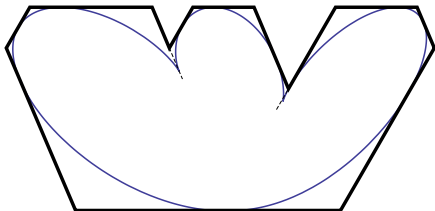
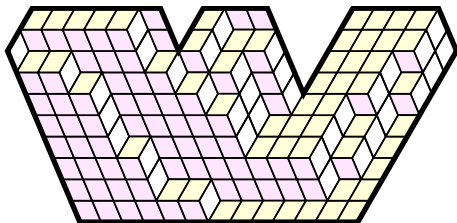
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**Advantage.** This method can be generalized to a class of polygonal domains with horizontal boundaries on two straight lines. Still there are explicit formulas (rational parameterization) for the boundaries of frozen regions (Petrov–2012).

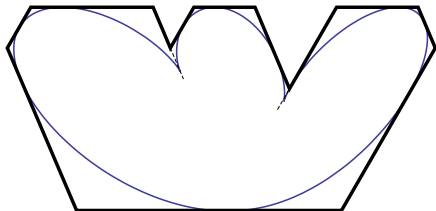
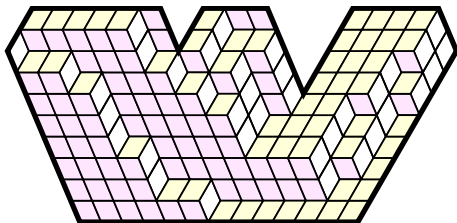
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**Disadvantage.** Breaks down if we change either class of domains or measure.

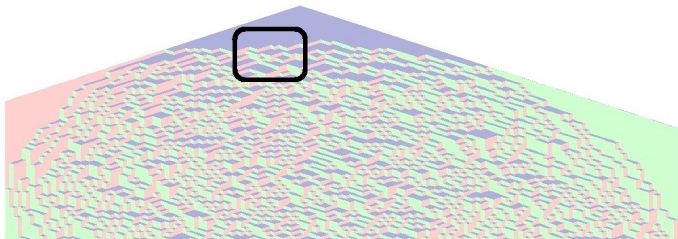
## Lozenge tilings: appearance of $N^{1/3}$

What are the fluctuations of the boundary of the frozen region of  $Na \times Nb \times Nc$  hexagon?

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(Baik-Kriecherbauer-McLaughlin-Miller; Petrov) For the uniformly random tilings the right scaling is  $N^{1/3}$  in normal direction and  $N^{2/3}$  in tangent direction to the frozen boundary.



After rescaling the limit fluctuations of the frozen boundary are governed by the Tracy-Widom distribution  $F_2$ .

The whole 2D picture is governed by the *extended Airy kernel*.

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**Baik-Kriecherbauer-McLaughlin-Miller:** a partial result via orthogonal polynomials.

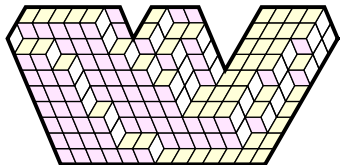
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**Baik-Kriecherbauer-McLaughlin-Miller:** a partial result via orthogonal polynomials.

**Petrov:** complete result via the steepest descent analysis of the double contour integrals.

The proof again extends to a class of polygonal domains with horizontal boundaries on two straight lines.



**Question.** Is this  $N^{1/3}$  the same as that for TASEP and, more generally, KPZ-class growth models?



## From lozenge tilings to TASEP

Relation to KPZ is obtained through the construction of a growth model for tilings.

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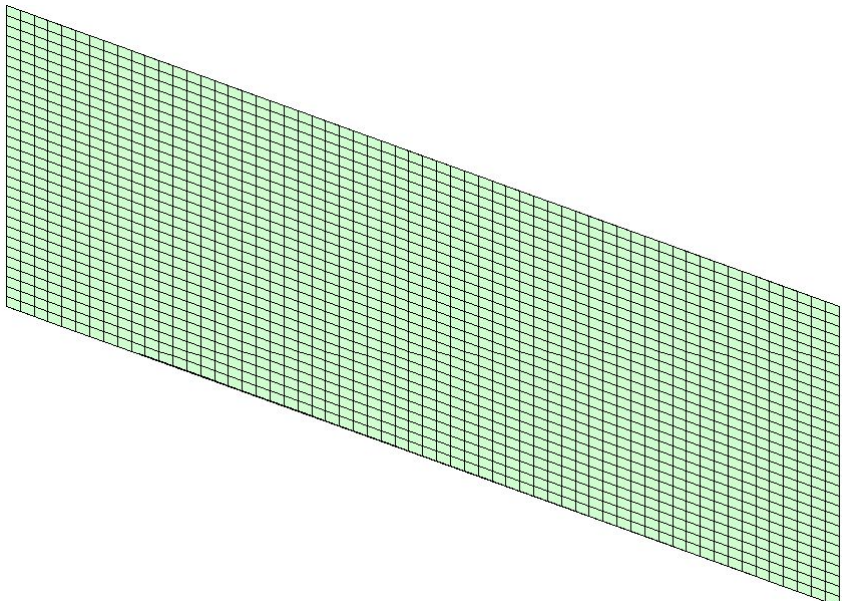
Relation to KPZ is obtained through the construction of a growth model for tilings.

**Theorem.** (Borodin-Gorin-2009) There exists a simple discrete time Markov chain which relates uniformly distributed tilings of hexagons of various sizes. Elementary step of this chain changes the size of hexagon from  $a \times b \times c$  to  $a \times (b - 1) \times (c + 1)$ . Algorithmically, one step involves generating some independent one-dimensional random variables.

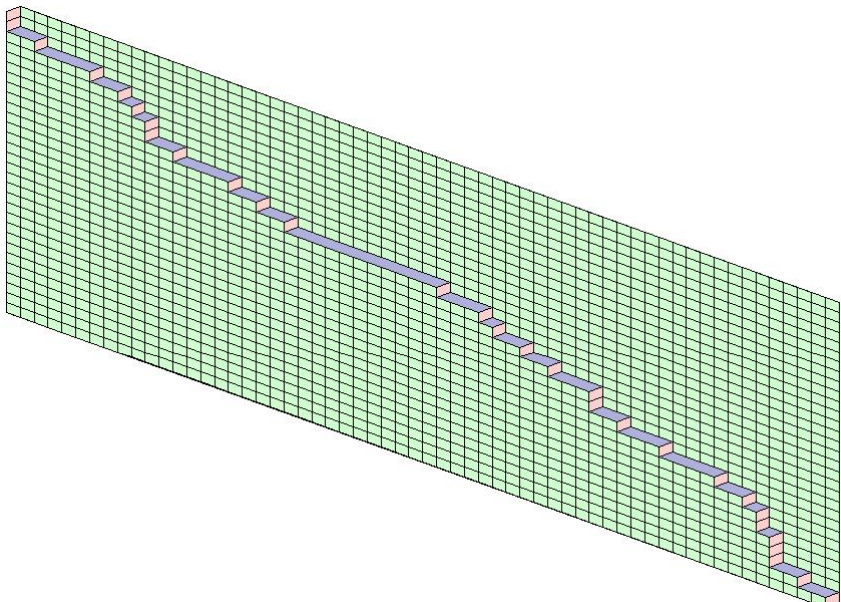
**Observation.** There is exactly one way to tile  $a \times b \times 0$  hexagon.

**Remark.** The construction can be generalized to more complicated *elliptic* weights on tilings of hexagon (Borodin-Gorin-Rains-2009; Betea-2011)

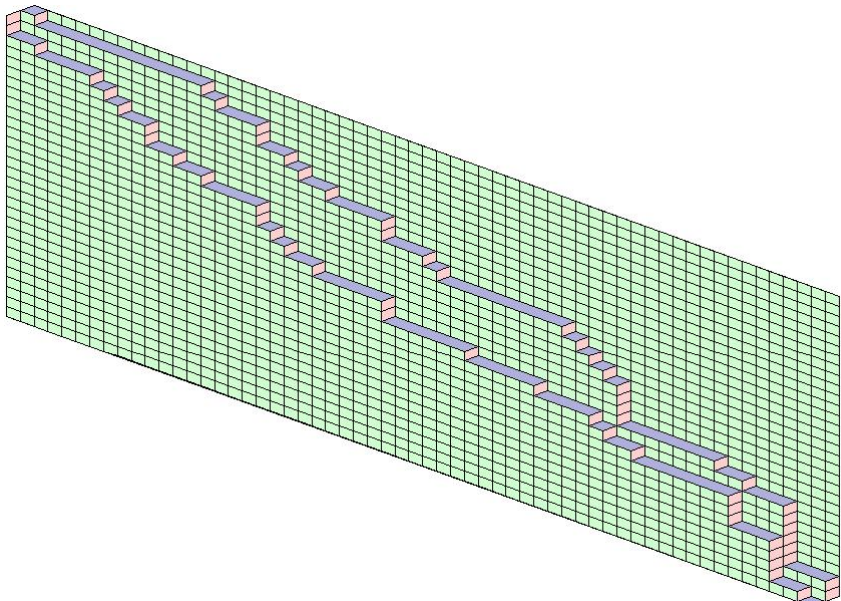
$30 \times 60 \times 0$



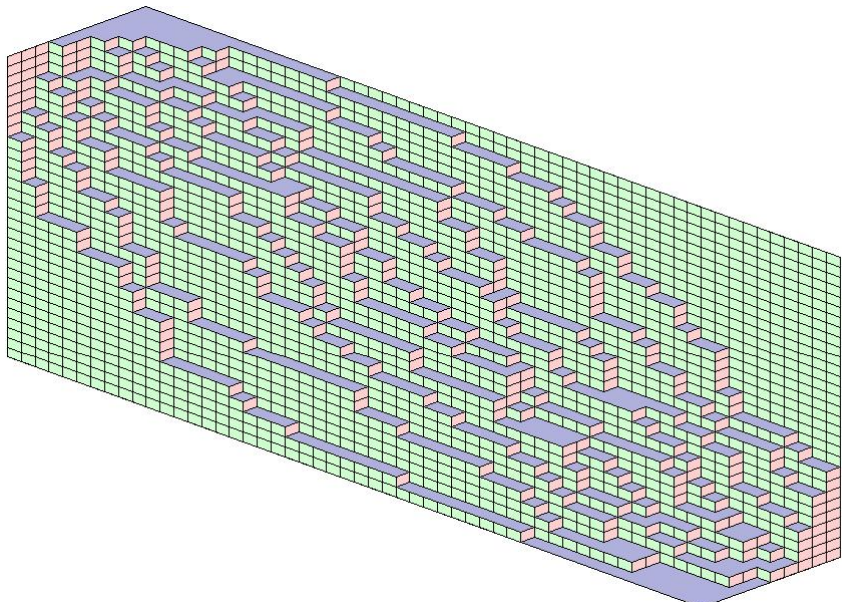
$30 \times 59 \times 1$



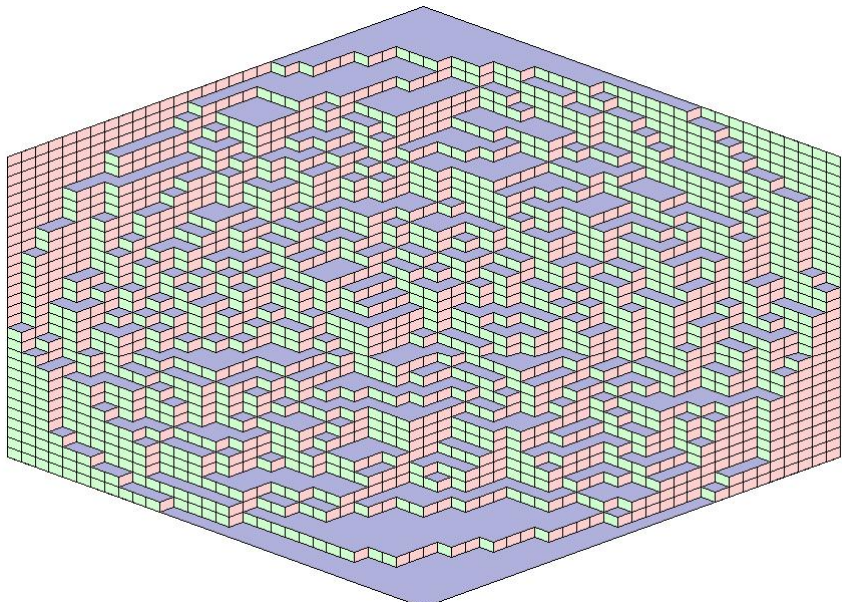
$$30 \times 58 \times 2$$



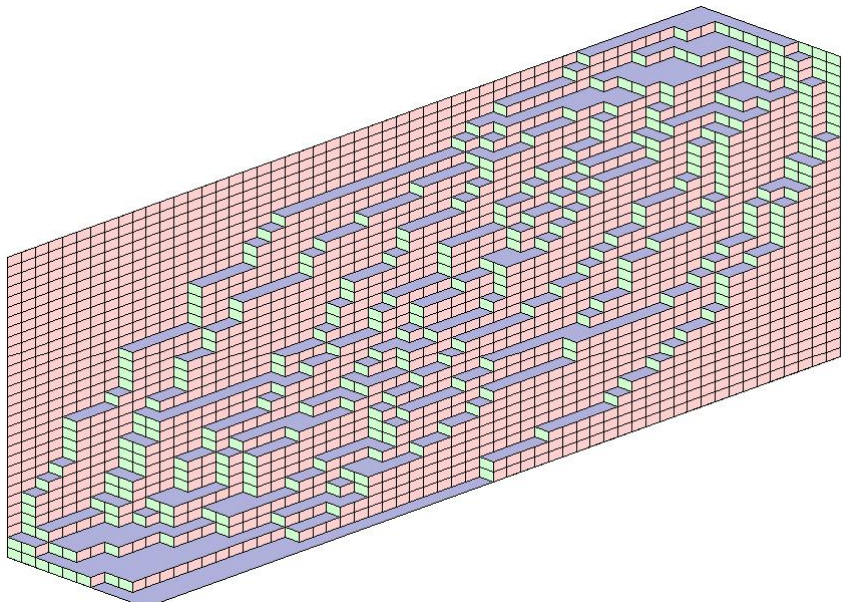
$30 \times 50 \times 10$



$30 \times 30 \times 30$

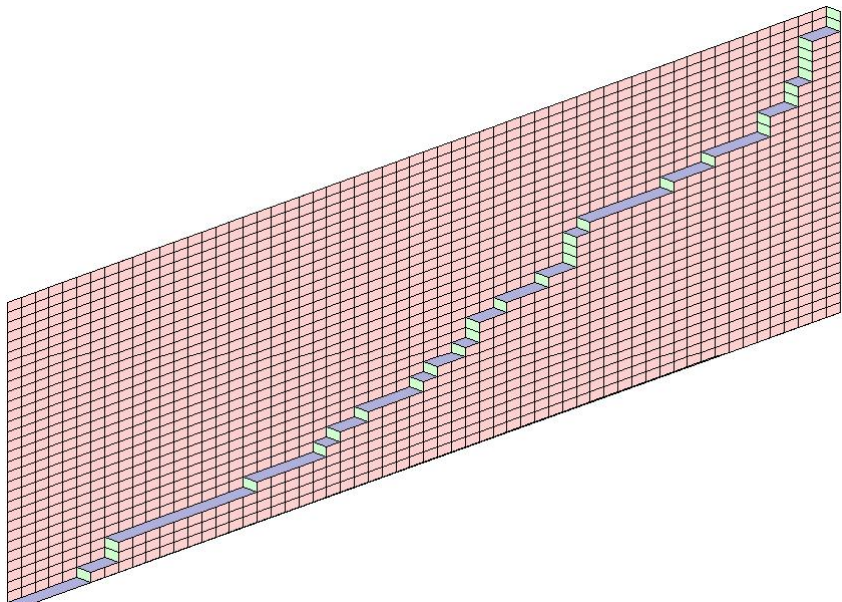


$30 \times 10 \times 50$

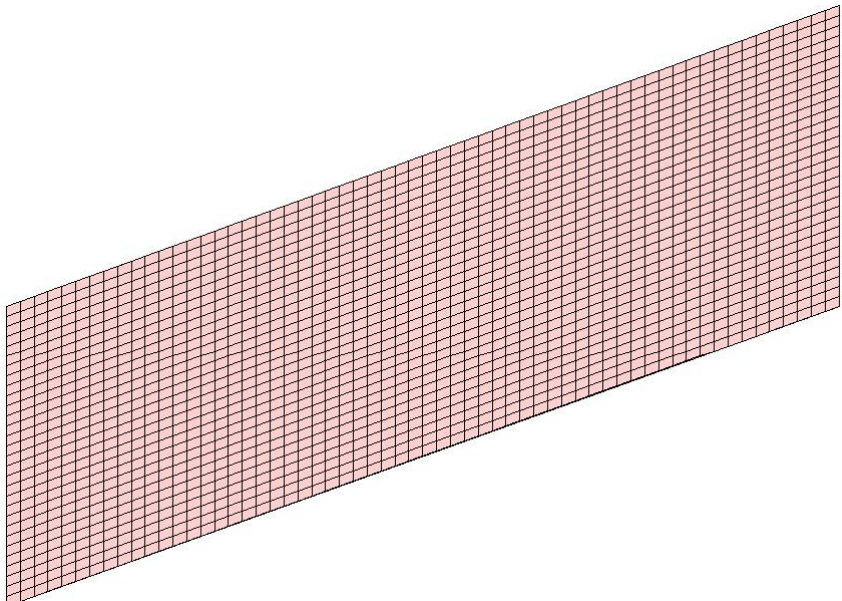




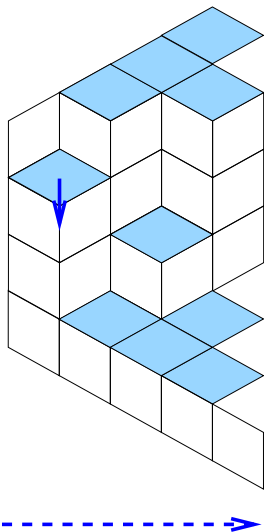
$$30 \times 1 \times 59$$



$30 \times 0 \times 60$

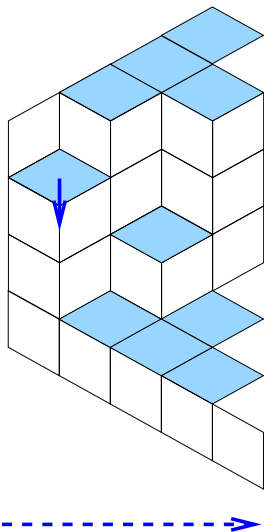


## From lozenge tilings to TASEP



**Description of Markov chain  
in terms of (interlacing)  
horizontal lozenges**

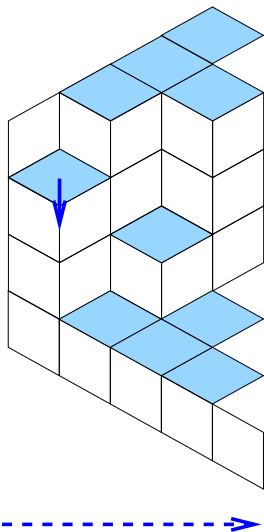
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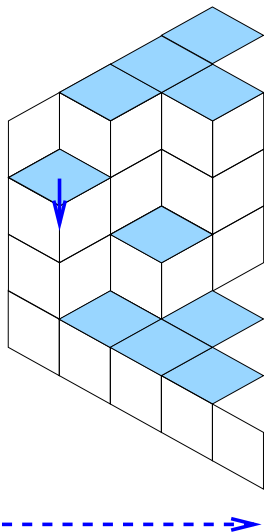
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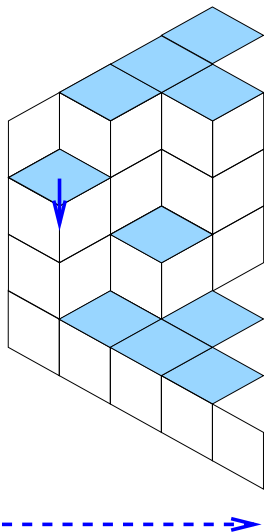
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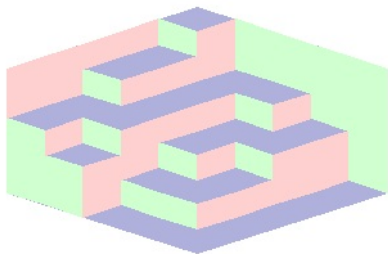
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Works for more complicated measures, but only for hexagon.

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This Markov chain is simplified in the limit transition  $a, b, c \rightarrow \infty$ ,  
 $ab/c \rightarrow t$  (time).

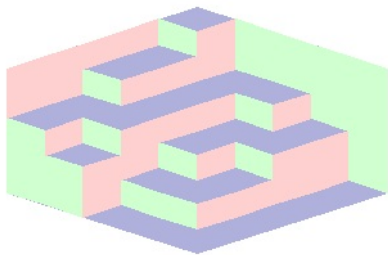


(leftmost horizontal lozenges)

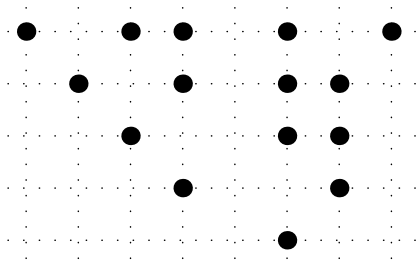


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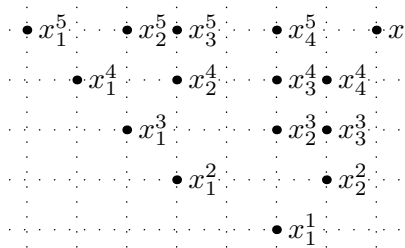


(leftmost horizontal lozenges)



(rotated picture)

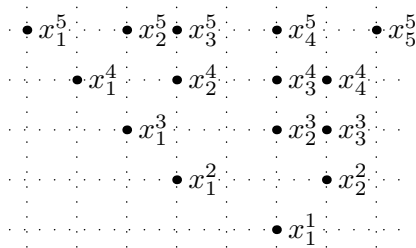
## From lozenge tilings to TASEP



In the limit we get a continuous time dynamics  $Y(t)$  on particle configurations with integer coordinates subject to interlacing conditions  $x_{i-1}^j < x_{i-1}^{j-1} \leq x_i^j$ .

Each particle has an exponential clock of rate 1. All clocks are independent. When the clock rings, the particle attempts to jump to the right.

## From lozenge tilings to TASEP



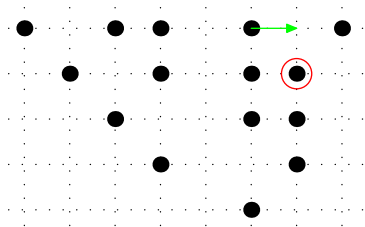
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The interlacing conditions are preserved by the rule “if higher, then lighter”.

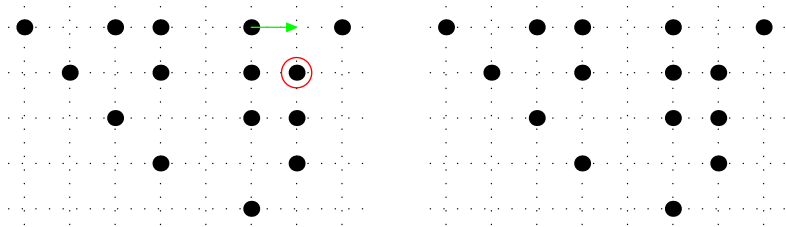
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Block:



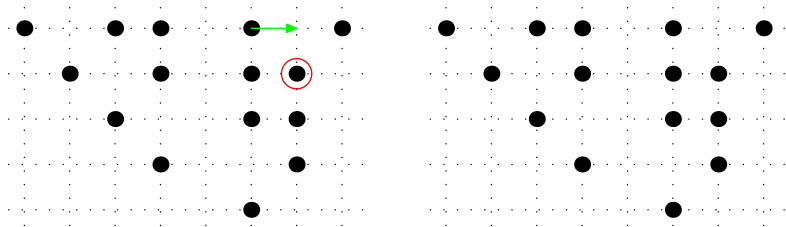
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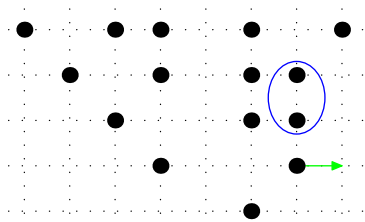


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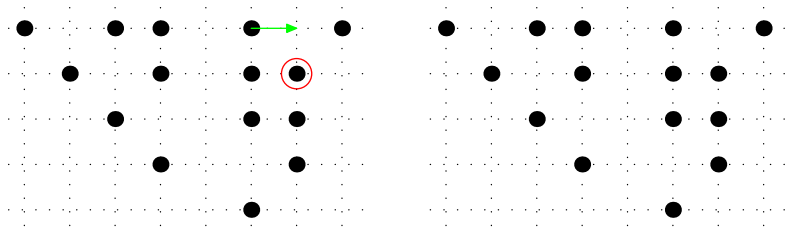


Push:

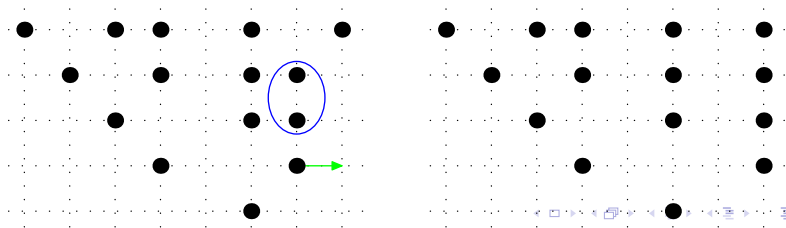


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Block:



Push:



## From lozenge tilings to TASEP

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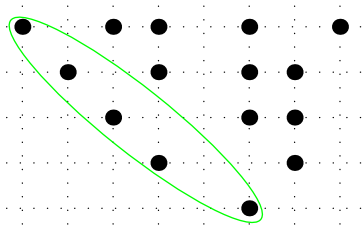
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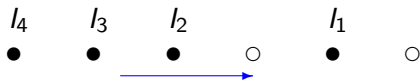
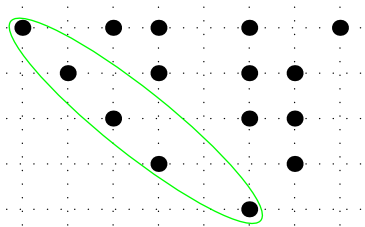
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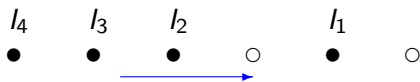
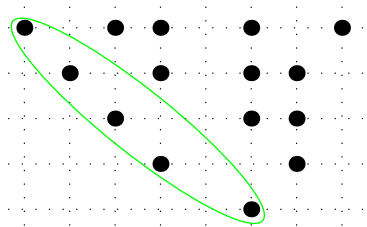


Totally Asymmetric Simple  
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## From lozenge tilings to TASEP

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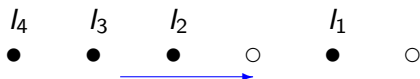
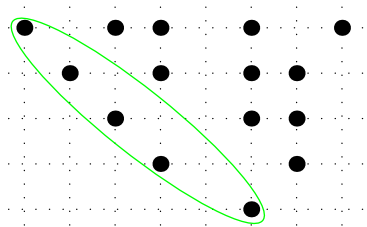
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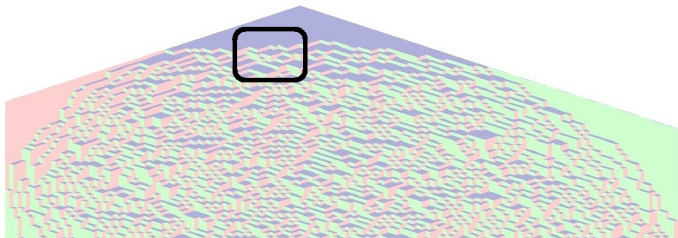
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Both TASEP and PushASEP are known to belong to the KPZ universality class of growth models.

## Lozenge tilings: appearance of $N^{1/3}$

What are the fluctuations of the boundary of the frozen region of  $Na \times Nb \times Nc$  hexagon?

(Baik-Kriecherbauer-McLaughlin-Miller; Petrov) For the uniformly random tilings the right scaling is  $N^{1/3}$  in normal direction and  $N^{2/3}$  in tangent direction to the frozen boundary.

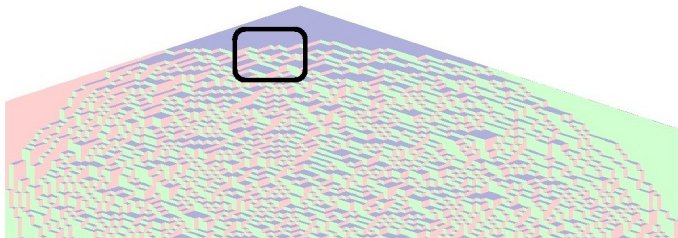


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**Question.** Is this  $N^{1/3}$  the same as that for eigenvalues of random matrices?

## $N^{1/3}$ in random matrices

Gaussian Unitary Ensemble of rank  $N$  is the distribution on the set of  $N \times N$  *Hermitian* matrices with density

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All eigenvalues are real. *GUE-eigenvalues* density is (Weyl, 20-30s)

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Wigner (1955): spacing between eigenvalues model the spacings between the lines in the “spectrum” of a heavy atom.

## $N^{1/3}$ in random matrices

$$\left( \begin{array}{c|c|c|c} a_{11} & a_{12} & a_{13} & a_{14} \\ \hline a_{21} & a_{22} & a_{23} & a_{24} \\ \hline a_{31} & a_{32} & a_{33} & a_{34} \\ \hline a_{41} & a_{42} & a_{43} & a_{44} \end{array} \right)$$

Let  $x_i^k$  be  $i$ th eigenvalue of top-left  $k \times k$  corner of GUE.

Interlacing condition:

$$x_{i-1}^j \leq x_{i-1}^{j-1} \leq x_i^j$$

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$$\begin{array}{cccc} x_1^4 & & x_2^4 & & x_3^4 & & x_4^4 \\ & x_1^3 & & x_2^3 & & x_3^3 & \\ & & x_1^2 & & x_2^2 & & \\ & & & x_1^1 & & & \end{array}$$

The joint distribution of  $x_i^j$  is known as GUE-minors process (although, correct name would be GUE-corners)

Given  $x_1^N, \dots, x_N^N$ , the distribution of  $x_i^j$ ,  $j < N$  is *uniform* on the polytope defined by interlacing conditions (Baryshnikov, 2001)

## $N^{1/3}$ in random matrices

**Theorem.** (Tracy–Widom–1994)

$N^{1/6}(x_N^N - 2N^{1/2})$  converges as  $N \rightarrow \infty$  towards the distribution  $F_2$ .

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Rescale the picture so that a typical spacing becomes of order 1 (as in lozenge tilings).

$$N^{-1/6} \cdot N^{1/2} = N^{1/3}!$$

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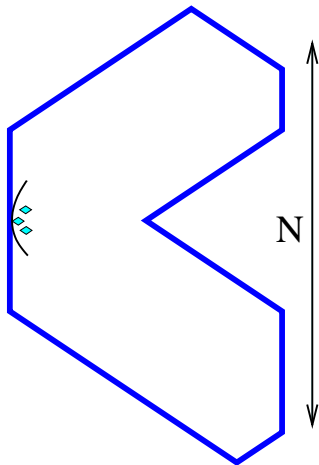
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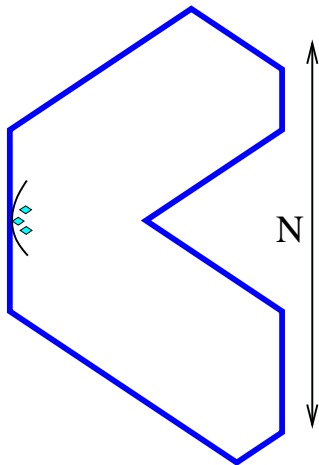
How can a lozenge tiling become a random matrix?

## From lozenge tilings to random matrices



**General conjecture.** For the domain of linear size  $N$  near the point where frozen boundary is tangent to the boundary of the domain, the fluctuations are of order  $\sqrt{N}$ . After rescaling, the distribution of position of one type of lozenges converges to *GUE*-corners process.

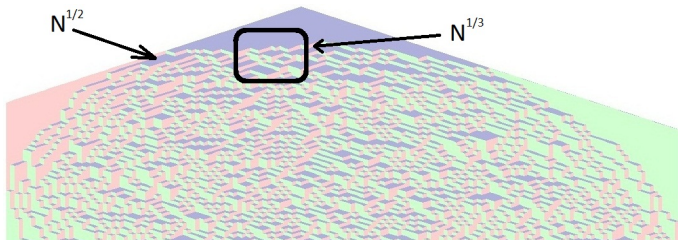
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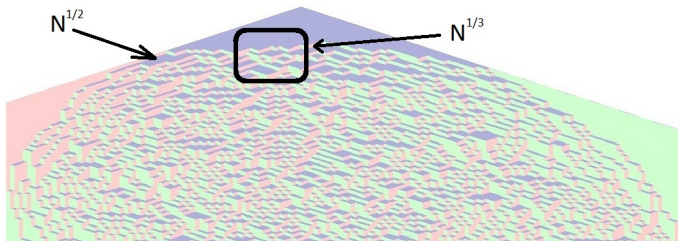
Stated by Okounkov and Reshetikhin in 2006. They also gave an *informal* argument explaining why this should be true.

## From lozenge tilings to random matrices



Away from the boundary the fluctuations are of order  $N^{1/3}$ , with limit governed by Tracy–Widom distribution  $F_2(s)$ .  
Near the boundary fluctuations of the frozen curve of lozenge tiling are of order  $N^{1/2}$  with Gaussian limit.  
GUE–corners process glues  $N^{1/2}$  and  $N^{1/3}$  classes together.

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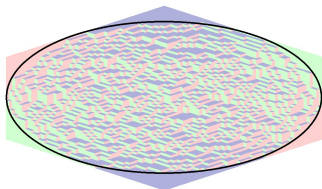
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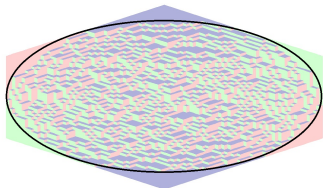
The conjecture itself is now proved in several cases.

## From lozenge tilings to random matrices



**Theorem.** (Johansson–Nordenstam, 2006; Nordenstam, 2009) For the hexagon the fluctuations near the point where inscribed ellipse touches the boundary are of order  $\sqrt{N}$  and after rescaling the point process formed by positions of one type of lozenges converges to *GUE*–corners process.

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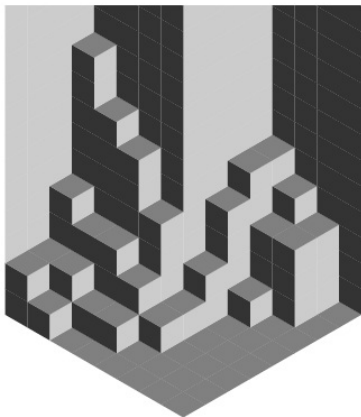


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**Method:** Computation based on Lindström–Gessel–Viennot formula for the number of non-intersecting paths + certain determinant evaluations. The asymptotics itself is a shadow of the convergence of Hahn (hypergeometric) orthogonal polynomials toward Hermite orthogonal polynomials.



## From lozenge tilings to random matrices

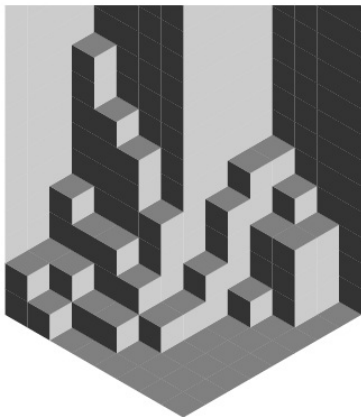


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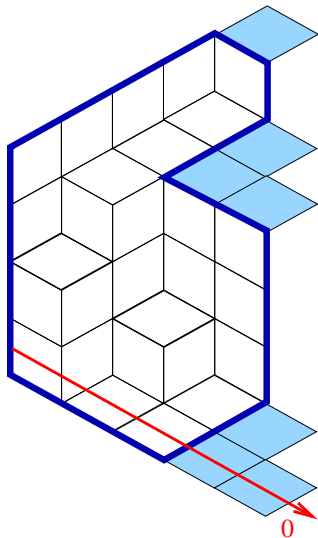
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**Method:** Determinantal point  
processes + formalism of Schur  
processes which leads to double  
contour integral representation  
of the correlation kernel +  
steepest descent analysis of  
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## From lozenge tilings to random matrices

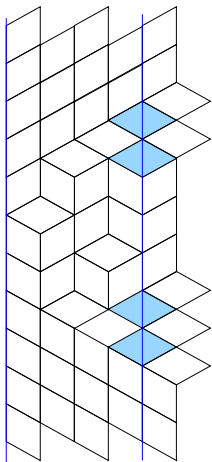
**Theorem.** (Gorin–Panova, 2013) GUE-convergence conjecture holds for the following class of domains.



Domain  $\Omega_{N, y(N)}$  is parameterized by its width  $N$  and positions  $y(N)_1 < y(N)_2 < \dots < y(N)_N$  of  $N$  horizontal lozenges at the right boundary. Tiling this domain is the same as tiling a certain polygon.

Here  $N = 5$  and  $y(5) = (0 < 1 < 5 < 6 < 8)$ .

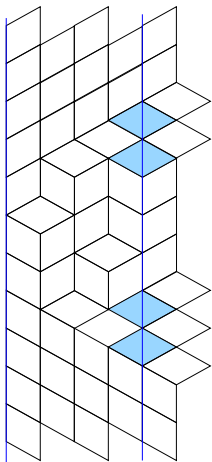
# From lozenge tilings to random matrices



## Method.

How to compute the distribution of  $k$  horizontal lozenges  $\lambda_1^k, \dots, \lambda_k^k$  on  $k$ th vertical line in uniformly random tiling of  $\Omega_{y(N), N}$ ?

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The distribution itself is complicated. But there are formulas for certain generating functions.

## From lozenge tilings to random matrices

Given  $\lambda_1 < \lambda_2 < \dots < \lambda_k$  the corresponding Schur function is a symmetric polynomial in  $x_1, \dots, x_k$  given by (note a bit non-standard notation)

$$s_\lambda(x_1, \dots, x_k) = \frac{\det [x_i^{\lambda_j}]_{i,j=1}^k}{\prod_{i < j} (x_j - x_i)}$$

**Proposition.** Let  $\lambda^{(k)} = \lambda_1^k, \dots, \lambda_k^k$  encode  $k$  horizontal lozenges on  $k$ th vertical line in uniformly random tiling of  $\Omega_{y(N), N}$ .

$$\mathbb{E} \left( \frac{s_{\lambda^{(k)}}(x_1, \dots, x_k)}{s_{\lambda^{(k)}}(1, \dots, 1)} \right) = \frac{s_{y(N)}(x_1, \dots, x_k, \underbrace{1, \dots, 1}_N)}{s_{y(N)}(\underbrace{1, \dots, 1}_N)}.$$

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Now the limit theorem boils down to the (still non-trivial!) study of asymptotics of Schur functions as the number of variables tends to infinity, which we do.

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**One of the reasons for the existence of rich structures:**

Interlacing particle configurations are *Gelfand–Tsetlin patterns* from representation theory.

(E.g. they parameterize bases of irreducible representations of unitary groups.)