

# Limit Theorems for the simplest parking process

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Ongoing research:

Cooperation between FAPESP and Universidad de Antioquia  
(Medellin)

In collaboration with

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- ▶ Cristian Coletti (UFABC, Brasil)

Many thanks to Eulalia and Giulio for the opportunity to share!

# Content

**The problem**

History and Motivation

Some new results

Idea of the proofs

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$\Rightarrow X_{\Lambda_n}$  is called the jamming limit of  $\Lambda_n$ .

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2. Let

$$N_n := \sum_{i \in \Lambda_n} X_{\Lambda_n}(i) \quad \text{and} \quad N_n^Y := \sum_{i \in \Lambda_n} Y(i)$$

What about the statistical properties of  $X_{\Lambda_n}$  and  $Y$ ?:

→ LLN, TCL, LIL... for  $N_n$  and  $N_n^Y$

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## Interesting model because:

- ▶ Peculiar type of dependence between the  $X_{\Lambda_n}(i)$ 's
- ▶ It is not defined through conditioning (specifications of statistical physics)
- ▶ Strongly non-Gibbsian (for those who know what it takes to be Gibbsian).
- ▶ Irreversibility of the dynamics.

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Note that

$$N_{n-1} = \frac{1}{2}Z_n$$

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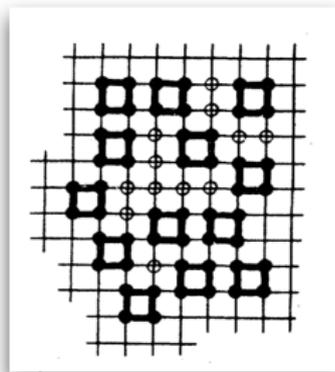
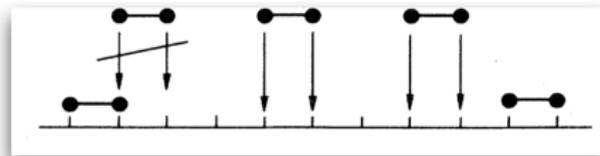
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Particles arrive at random locations, and each adsorbed particle occupies a region of the substrate which prevents the adsorption of any subsequently arriving particle in an overlapping surface region.

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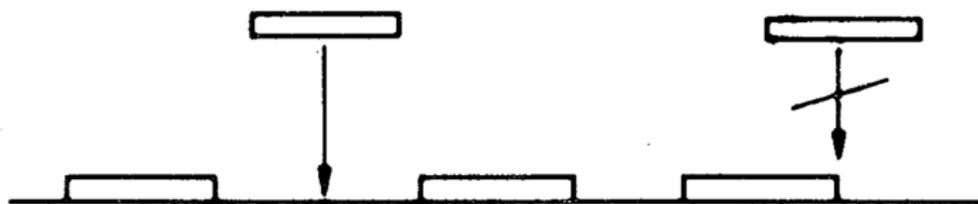
Particles arrive at random locations, and each adsorbed particle occupies a region of the substrate which prevents the adsorption of any subsequently arriving particle in an overlapping surface region.



**Figure:** 2-mers on the left, 2x2-mers on the right

## Parenthesis: Continuous counterparts

The Rényi car parking problem: Cars are parked uniformly at random in  $[0, x]$ ,  $x > 0$

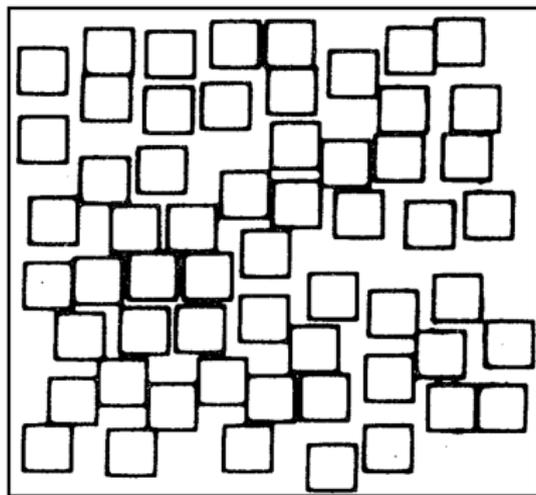


Rényi (1958) proved that

$$\frac{N[0, x]}{x} \rightarrow 0.7475979202... \text{ a.s.}$$

## Parenthesis: Continuous counterparts

Cars are parked uniformly at random in  $[0, x]^2$ ,  $x > 0$



(Brosilow *et al.*, 1991)  $\lim \frac{N([0, x]^2)}{x^2} \rightarrow 0,562009\dots$  a.s.

# Other nomenclature/applications/interpretation

- ▶ Fatmen seating problem
- ▶ Unfriendly seating problem
- ▶ Packing problem
- ▶ ...

Find applications in

- ▶ Polymer chemistry
- ▶ Independent sets (graph theory)
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See the paper by Evans (1993)

*“Random and Cooperative sequential adsorption”*

## Some literature (most in 1d)

- ▶ Page (1959), Freedman and Shepp (1962), Flajolet (1998), Pinsky (2014), ...

$$\frac{\mathbb{E}(N_n)}{n} = \frac{1}{2}(1 - e^{-2}) + \text{precise error term}$$
$$\frac{\text{Var}(N_n)}{n} = e^{-4} + \text{precise error term}$$

- ▶ Page (1959):  $\frac{N_n}{n} \xrightarrow{\mathbb{P}} \frac{1}{2}(1 - e^{-2})$
- ▶ Penrose (2002) (any dimension):  $\frac{N_n}{n} \xrightarrow{L^p} \rho_d$  and CLT.
- ▶ Ritchie (2006) (any dimension): Thermodynamic limit and  $\frac{N_n}{n} \xrightarrow{\text{a.s.}} \rho_d$
- ▶ Pinsky (2014) (very fat men): extended results of Page (1959).
- ▶ Gerin (2015): didn't know about Ritchie's paper it seems.
- ▶ Chern *et al* (2015): "Dinner table".
- ▶ And many others papers in Physics literature based on simulations.

## Much more related to our problem

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He obtains CLTs for general models, but through a very long and complicated path:

*“However, since we always obtain our systems **by taking the random input to come only from inside the target region, rather than restricting a stationary random field to the target region,** general CLTs such as that of Bolthausen (1982) are not directly applicable.”*

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- ▶ To use classical results from random field literature: needs a **stationary random fields** on  $\mathbb{Z}^d$

$$(Y(i))_{i \in \mathbb{Z}^d}, \quad Y(i) \in \{0, 1\}$$

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$$(Y(i))_{i \in \mathbb{Z}^d}, \quad Y(i) \in \{0, 1\}$$

- ▶ Satisfying the rules of RSA!

## Ritchie constructed such a random field

Thomas Ritchie:

- ▶ *Construction of the Thermodynamic Jamming Limit for the Parking Process and Other Exclusion Schemes on  $\mathbb{Z}^d$ .* (2006).

He proved:

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- ▶ *Construction of the Thermodynamic Jamming Limit for the Parking Process and Other Exclusion Schemes on  $\mathbb{Z}^d$ .* (2006).

He proved:

- ▶ Perfect simulation algorithm of  $Y$  on any  $\Lambda \subset \mathbb{Z}^d$ :

$$Y(i) = [f(U)](i), \quad \forall i \in \mathbb{Z}^d$$

where

$U = (U(i))_{i \in \mathbb{Z}^d}$  is i.i.d.  $U_i \sim \text{Unif}[0, 1]$   
 $f : U \rightarrow \{0, 1\}^{\mathbb{Z}^d}$  is translation equivariant.

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- ▶ Strong law of large numbers

$$\frac{1}{|\Lambda_n|} \sum_{i \in \Lambda_n} Y(i) \xrightarrow{n \rightarrow \infty} \rho_d, \text{ a.s.}$$

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- ▶ Strong law of large numbers

$$\frac{1}{|\Lambda_n|} \sum_{i \in \Lambda_n} Y(i) \xrightarrow{n \rightarrow \infty} \rho_d, \text{ a.s.}$$

- ▶ With a control of boundary effects he proved

$$\frac{1}{|\Lambda_n|} \sum_{i \in \Lambda_n} X_{\Lambda_n}(i) \xrightarrow{n \rightarrow \infty} \rho_d, \text{ a.s.}$$

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# About the random field $Y$ : asymptotic results

## Theorem

For any  $d \geq 1$ , the random field  $Y$  satisfies

$$(CLT) \quad \frac{N_n^Y - |\Lambda_n| \rho_d}{\sqrt{\sigma^2 |\Lambda_n|}} \xrightarrow[n \rightarrow \infty]{\mathcal{D}} N(0, 1)$$

$$(LIL) \quad \limsup_n \frac{N_n^Y - |\Lambda_n| \rho_d}{\sqrt{2\sigma^2 |\Lambda_n| \log \log |\Lambda_n|}} = 1 \quad \text{a.s.}$$

where

$$\sigma^2 = \sum_{i \in \mathbb{Z}^d} \text{Cov}(Y(\mathbf{0}), Y(i)) > 0. \quad (1)$$

# About the random field $Y$ : non-asymptotic result

## Theorem

For any  $\epsilon > 0, n, d \geq 1$

$$\mathbb{P} \left( \left| \mathbf{N}_n^Y - \rho |\Lambda_n| \right| > \epsilon \right) \leq e^{\frac{1}{e} - \frac{\epsilon^2}{4eB|\Lambda_n|}} \quad (2)$$

where  $B = B(d)$  is explicit.

# About the sequence $X_{\Lambda_n}, n \geq 1$

## Theorem

- ▶ For any  $n, d \geq 1$

$$\left| \mathbb{E}N_n - |\Lambda_n|\rho_d \right| \leq \frac{2d(2d-1)^n}{(n+1)!} + (2d)^2 \sum_{k=0}^{n-1} \frac{(2d-1)^k (2(n-k)+1)^{d-1}}{(k+1)!}.$$

- ▶ The LIL holds for the sequence  $X_{\Lambda_n}, n \geq 1$  in  $d = 1$ .

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Couldn't get rid of the boundary effects to get the LIL in  $d \geq 2$ ...

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# The base of our proofs: Ritchie's perfect simulation algorithm

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- ▶ **Solves the issue of Penrose**

It simulates from any region  $\Lambda \subset \mathbb{Z}^d$  a sample  $Y(\Lambda)$  which is a compatible projection of the whole random field  $(Y(i))_{i \in \mathbb{Z}^d}$ .

- ▶ **It gives all we want at once**

It easily yields good mixing properties allowing to use results from the literature.

- ▶ **It is very elegant!**

## First step: “the uniforms algorithm” (in $\mathbb{Z}^2$ )

Consider  $(U_i)_{i \in \mathbb{Z}^2}$  i.i.d.'s with  $U_0 \sim \text{Unif}[0, 1]$  and region  $\Lambda$

0,25	0,87	0,78	0,41	0,64	0,61	0,50
0,86	0,51	0,42	0,94	0,06	0,93	0,23
0,55	0,38	0,57	0,74	0,52	0,29	0,85
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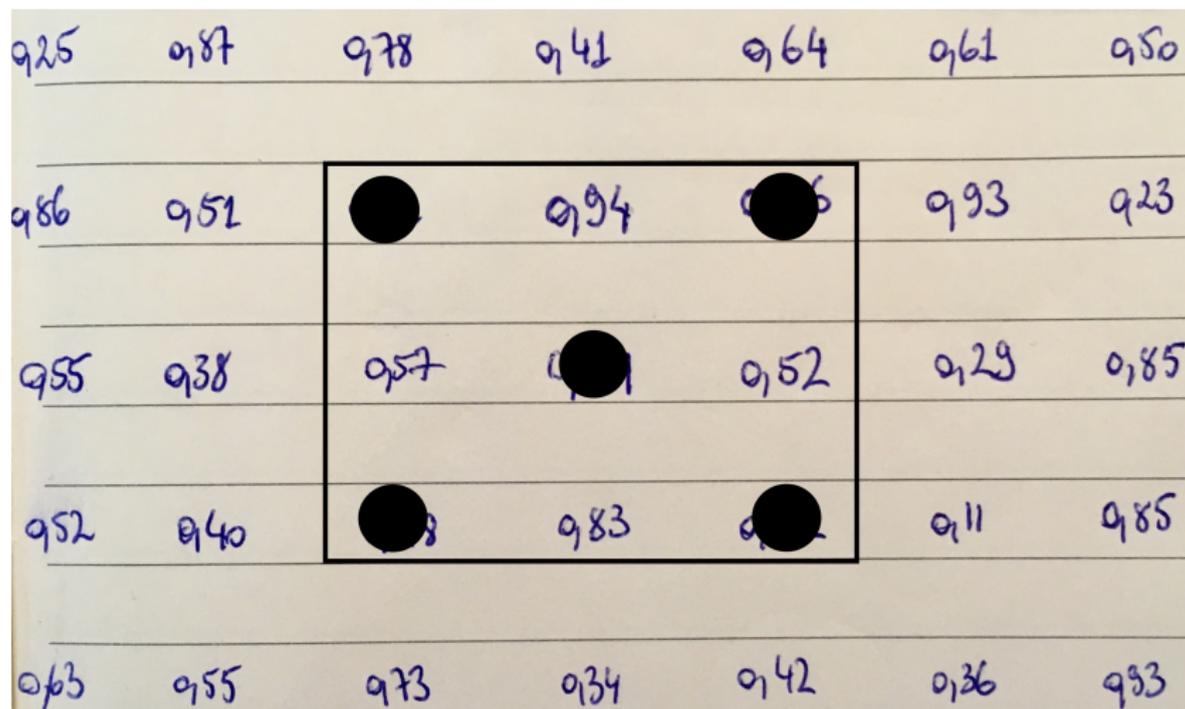
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## First step: “the uniforms algorithm” (in $\mathbb{Z}^2$ )

Observe that:

1. This has exactly the same distribution as the first (definition) algorithm for finite boxes.
2. It appears clearly now why we are not sampling from the thermodynamic limit: look at the 0,11!
3. This last observation is also the key to understanding how the PSA should work:

The decision of whether or not a particles is put at  $i \in \mathbb{Z}^2$  should not depend on the box, **but exclusively on the uniform random variables.**

# The perfect simulation algorithm

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- ▶ Define the “armour” of  $i \in \mathbb{Z}^2$  by

$$\mathcal{A}(\{i\}) := \bigcup_{y \in \mathbb{Z}^2: i \rightarrow y} \{\text{vertices on the path from } i \text{ to } y\}$$

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- ▶ **The PSA for  $i \in \mathbb{Z}^2$ :**

Just apply the “uniforms algorithm” in  $\mathcal{A}(\{i\})!!$

# Let us perfectly simulate $Y(\Lambda)$ for $|\Lambda| = 1$

Here is our  $\Lambda = \{i\}$ ...

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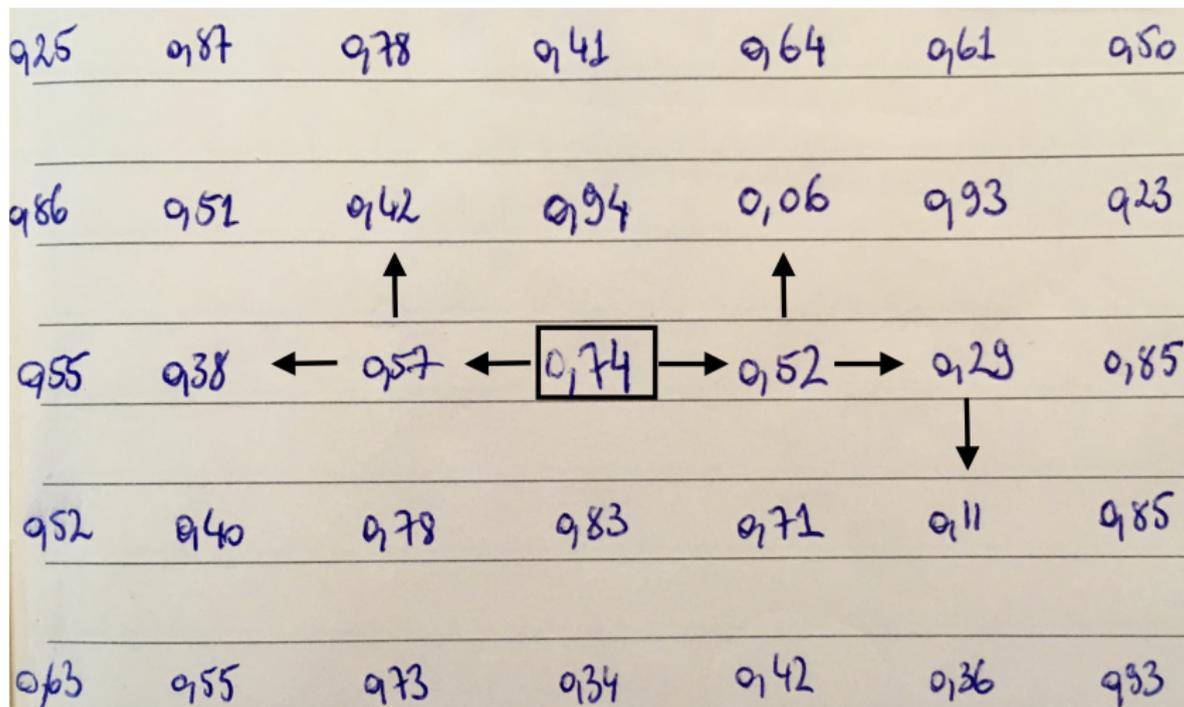
# Let us perfectly simulate $X_\Lambda$ for $|\Lambda| = 1$

... we construct its armour but going along “decreasing paths”...

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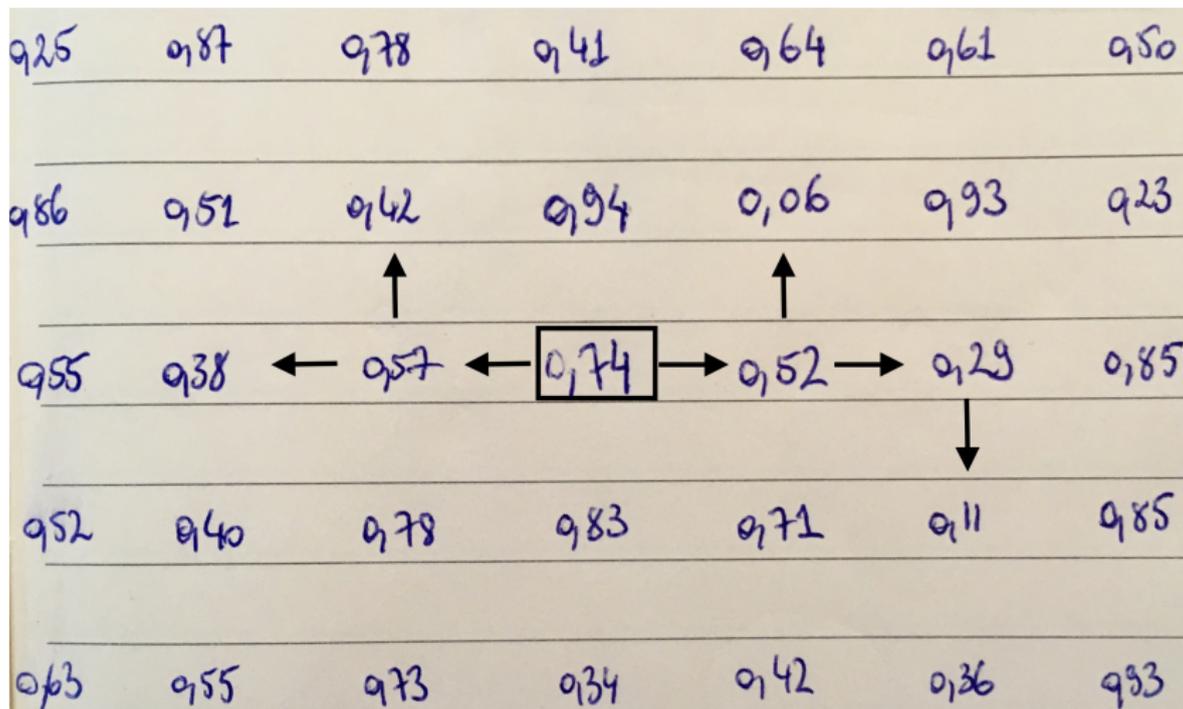
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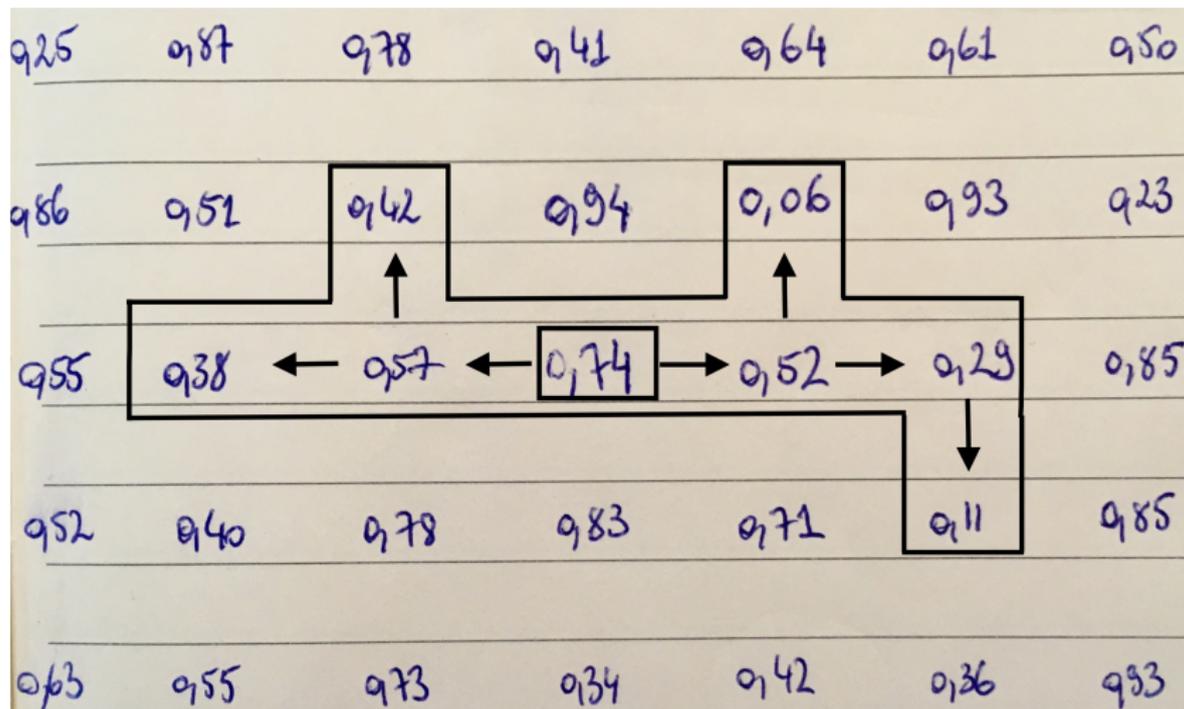
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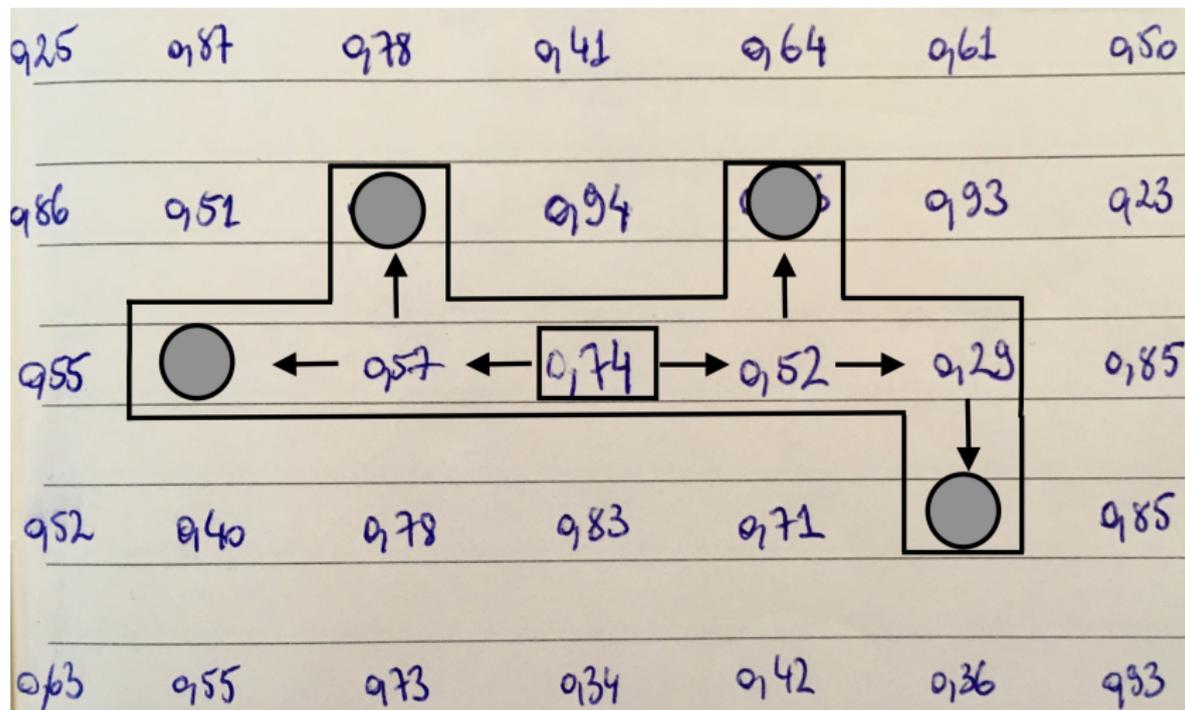
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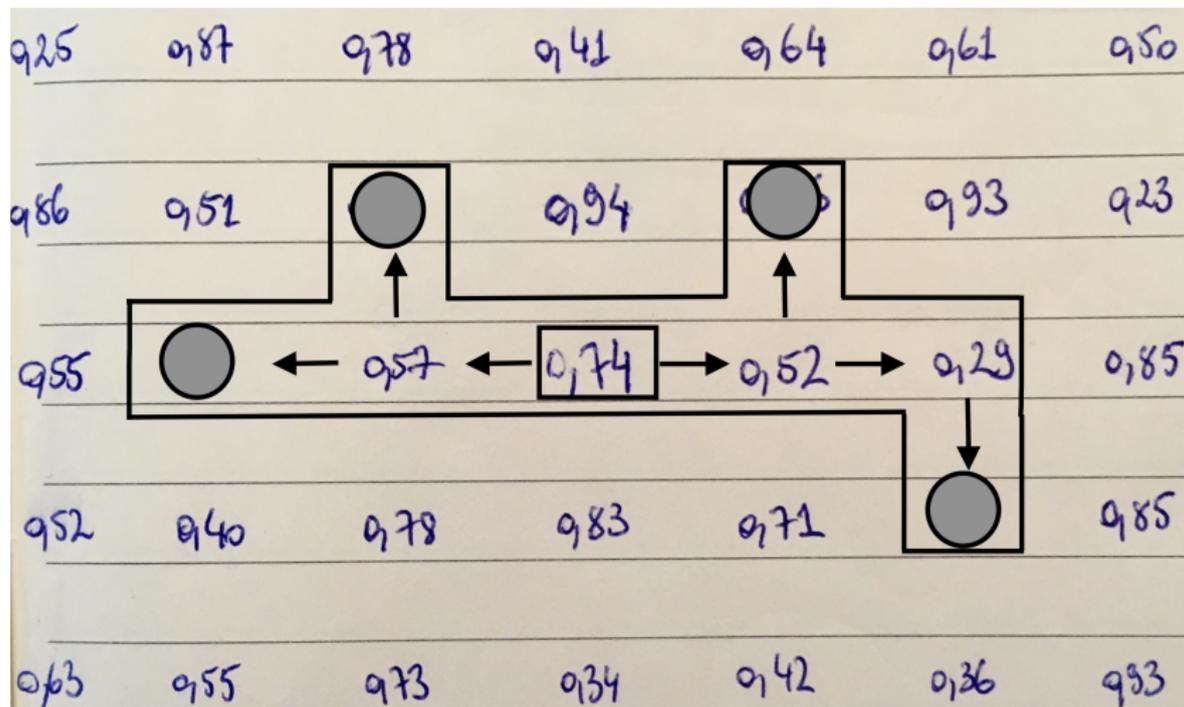
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... and we can now use the “uniform algorithm” inside  $\mathcal{A}(\{i\})$ ...



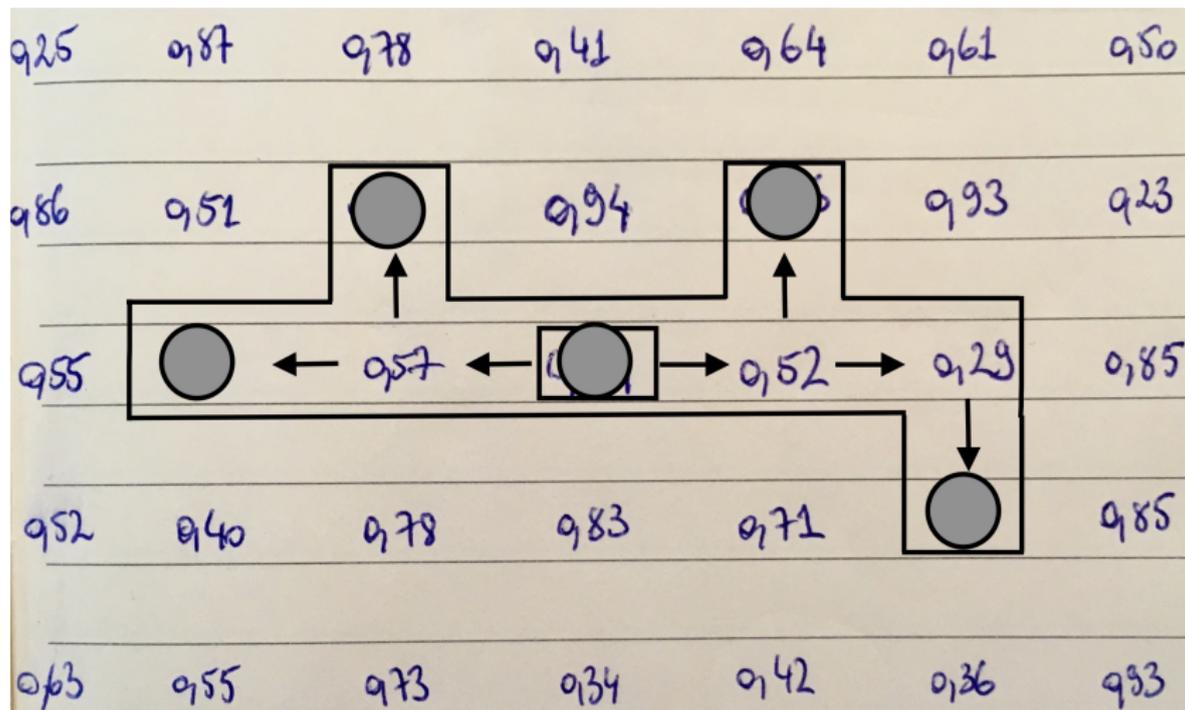
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... to conclude the algorithm:



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... to conclude the algorithm:  $Y(i) = 1$ .



## For the proofs concerning $\mathcal{Y}$

- ▶ We can make the PSA of any finite region  $\Lambda \subset \mathbb{Z}^d$  in finite time.
- ▶  $\mathcal{A}(\Lambda) = \cup_{i \in \Lambda} \mathcal{A}(\{i\})$
- ▶ Moreover  $|\mathcal{A}(\{i\})|$  has super-exponential tail.
- ▶ This gives very good  $\alpha$ -mixing
- ▶  $\Rightarrow$  Good mixing implies, for the random field:
  - ▶ SLLN,
  - ▶ CLT,
  - ▶ Berry-Esseen,
  - ▶ Concentration inequalities *etc...*

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- ▶ Thus  $|N_n - \bar{N}_n| > \sqrt{|\Lambda_n|}$  finitely many times.

**Thank you!**