

# Projections of fractal percolations II

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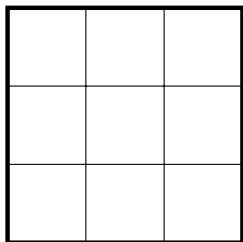
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- 1 History
- 2 The projections
- 3 Percolation phenomenon
- 4 New results
- 5 Non-homogeneous Fractal percolation sets
- 6 Homogeneous percolation of small dimension
- 7 The sum of three linear random Cantor sets

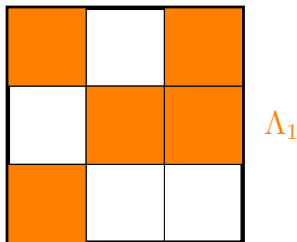
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We partition the unit square into  $M^2$  congruent sub squares each of them are independently retained with probability  $p$  and discarded with probability  $1 - p$ . In the squares retained after the previous step we repeat the same process at infinitum.



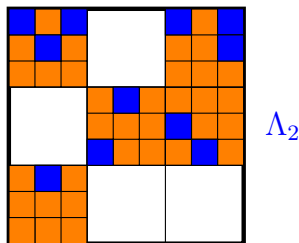
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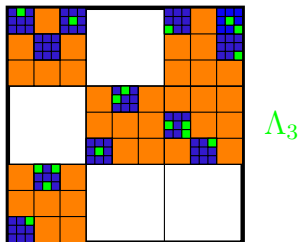
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Let  $\Lambda_n$  be the union of the level  $n$  retained squares. Then the statistically self-similar set of interest is:

$$\Lambda := \bigcap_{n=1}^{\infty} \Lambda_n.$$

It was proved by Falconer and independently Mauldin, Willims that conditioned on non-extinction:

$$\dim_{\mathbb{H}} \Lambda = \dim_{\mathbb{B}} \Lambda = \frac{\log(M^2 \cdot p)}{\log M} \text{ a.s.}$$

The expected number of descendants of every square is:  $M^2 \cdot p$ . Therefore, if  $M^2 \cdot p < 1$  then  $\Lambda = \emptyset$  a.s.

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So, we have almost surely:

- If  $p \leq 1/M^2$  then  $\Lambda = \emptyset$ .
- If  $1/M^2 < p < 1/M$  then  $\dim_{\mathbb{H}}(\Lambda) < 1$  (but  $\Lambda \neq \emptyset$  with positive probability).
- If  $p > \frac{1}{M}$  then either
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# Marstrand Theorem

## Theorem 1 (Marstrand)

Let  $B \subset \mathbb{R}^2$  be a Borel set.

- ① If  $\dim_{\mathbb{H}}(B) \leq 1$  then for  $\mathcal{L}eb$ -a.e.  $\theta$ , we have

$$\dim_{\mathbb{H}}(\text{proj}_{\theta}(B)) = \dim_{\mathbb{H}}(B)$$

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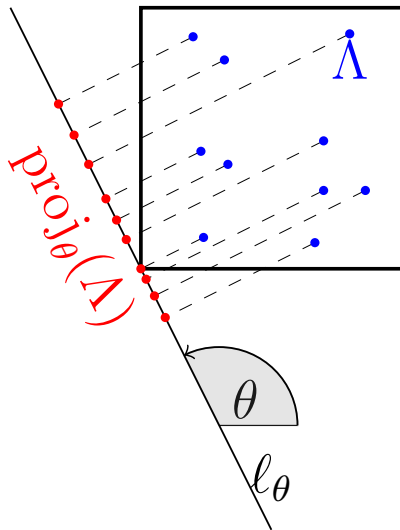
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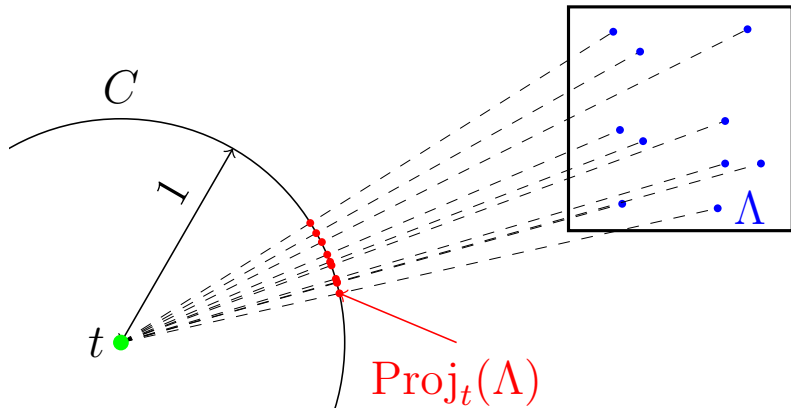
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# Orthogonal projection to $\ell_\theta$

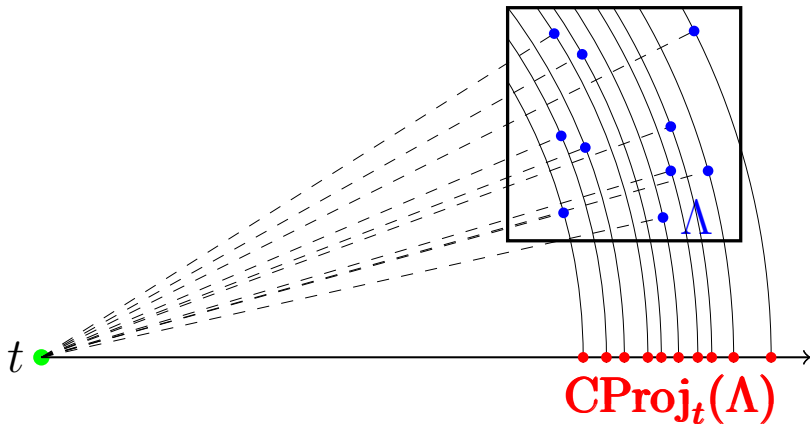


# Radial and co-radial projections with center $t$



Let  $\text{CProj}_t(\Lambda) := \{\text{dist}(t, x) : x \in \Lambda\}$  ( $\text{CProj}_t(\Lambda)$  is the set of the length of dashed lines above).

# The co-radial projection





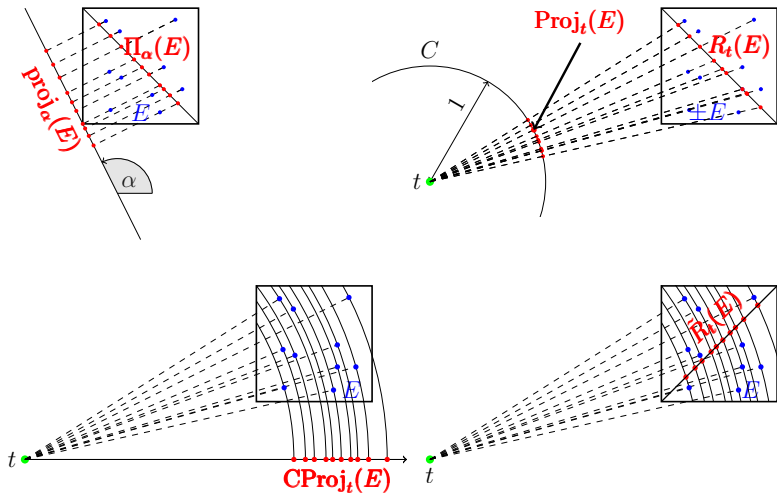


Figure: The orthogonal  $\text{proj}_\alpha$ , radial  $\text{Proj}_t$ , co-radial  $\text{CProj}_t$  projections and the auxiliary projections  $\Pi_\alpha$ ,  $R_t$ , and  $\tilde{R}_t$ .

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# $\Lambda$ percolates

Let  $\Lambda(\omega)$  be a realization of this random Cantor set. We say that  $\Lambda(\omega)$  **percolates** if there is a connected component of  $\Lambda(\omega)$  which connects the left and the right walls of the square  $[0, 1]^2$ .

Let us write  $E_{|\leftrightarrow|}$  for the event that the random self-similar set  $\Lambda$  **percolates**.

# Theorem [J.T. Chayes, L. Chayes, R. Durrett]

Let  $TD$  be the event that  $\Lambda$  is totally disconnected. That is all connected components are singletons. Let

$$\rho_c := \inf \{ \rho : \mathbb{P}_\rho (E_{|\leftrightarrow|}) > 0 \}$$

Then  $0 < \rho_c < 1$  and

$$\rho_c = \sup \{ \rho : \mathbb{P}_\rho (TD) = 1 \}.$$

If  $\rho < \rho_c < 1$  then all connected components of  $\Lambda$  are singletons. If  $\rho > \rho_c$  then  $\Lambda$  percolates with positive probability.

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# Theorem [R., S.] (When $p > \frac{1}{M}$ )

We assume that

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Then the following statements hold almost surely conditioned on  $\Lambda \neq \emptyset$ :

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$\forall t \in \mathbb{R}^2$ ,  $\text{Proj}_t(\Lambda)$  and  $\text{CProj}_t(\Lambda)$  contain an interval .

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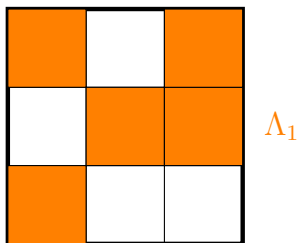
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$p_{0,2}$	$p_{1,2}$	$p_{2,2}$
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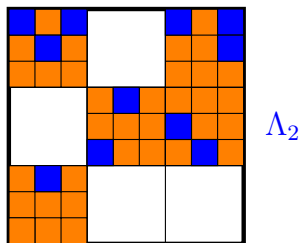
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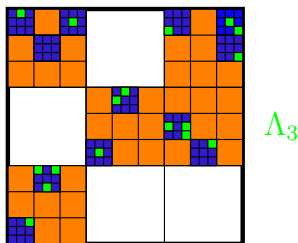
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# Theorem Rams, S.

Assume that

$$\textcircled{1} \quad \forall k: \sum_{i=0}^{M-1} p_{i,k} > 1 \text{ and } \sum_{j=0}^{M-1} p_{k,j} > 1 \text{ and}$$

$$\textcircled{2} \quad \forall \alpha \in \left(0, \frac{\pi}{2}\right) \cup \left(\frac{\pi}{2}, \pi\right), \alpha \text{ is good.}$$

Then the following statements hold almost surely conditioned on  $\Lambda \neq \emptyset$ :

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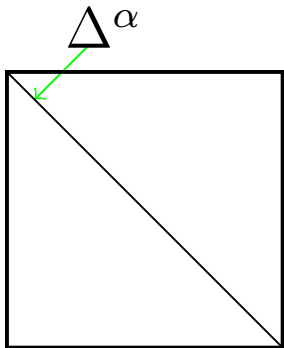
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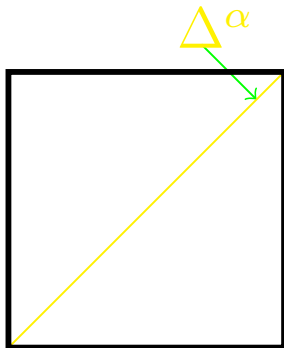
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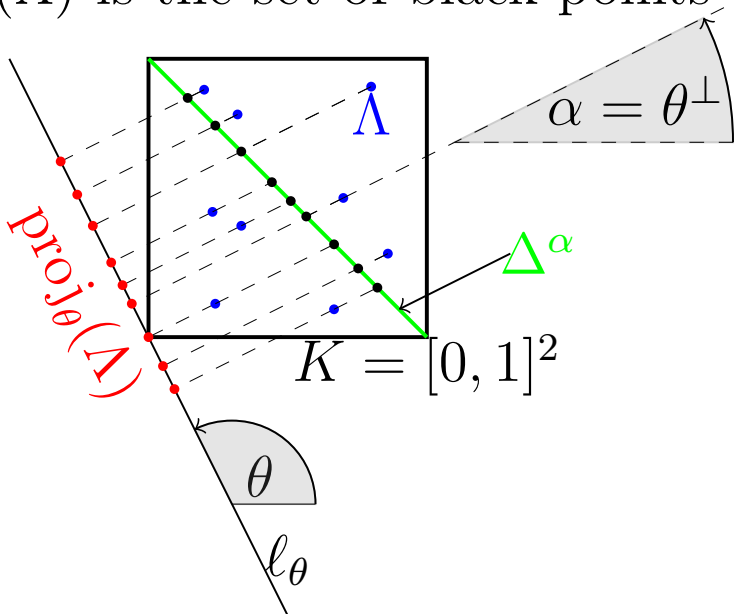
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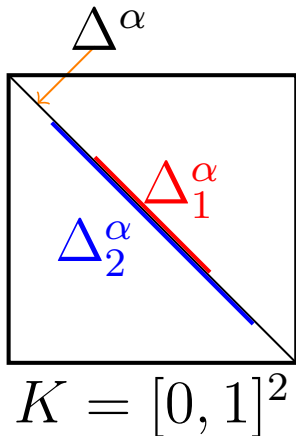
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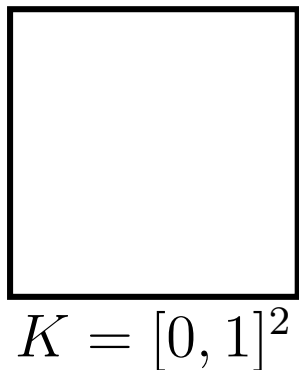
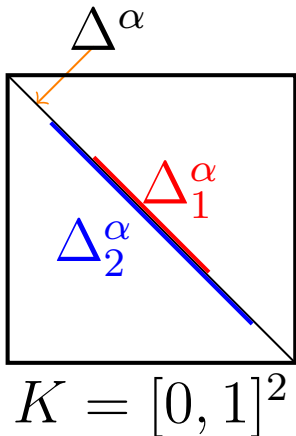
$\Pi_\alpha(\Lambda)$  is the set of black points



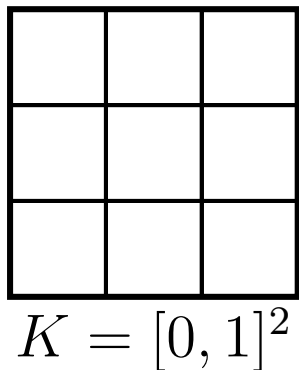
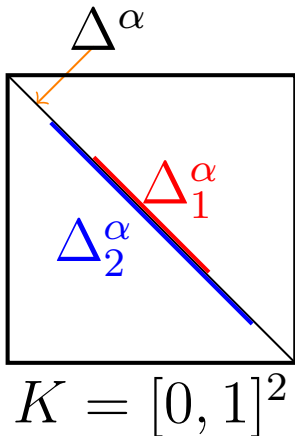
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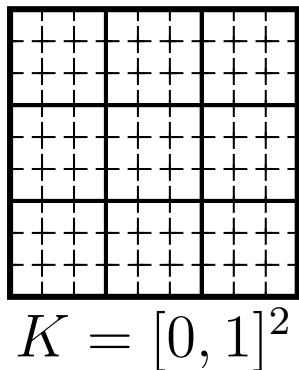
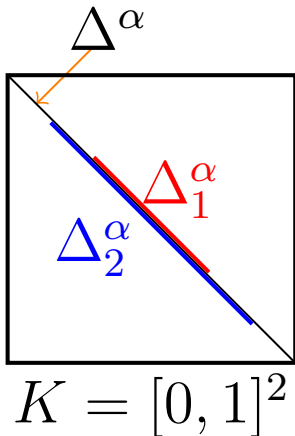
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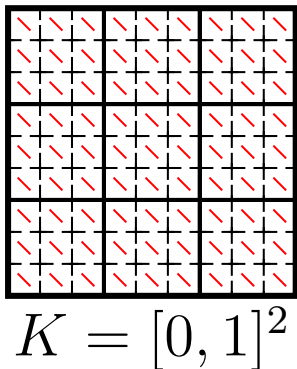
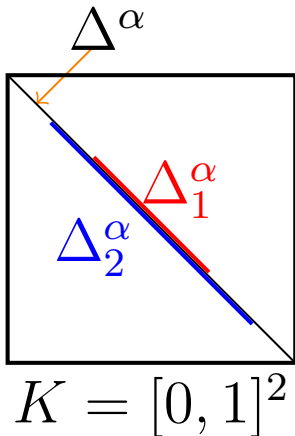
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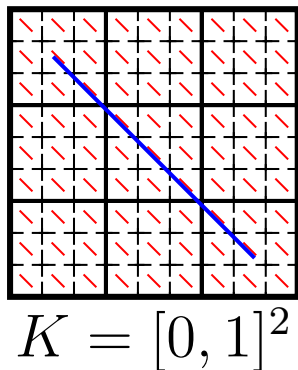
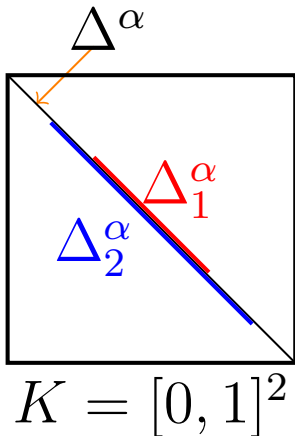
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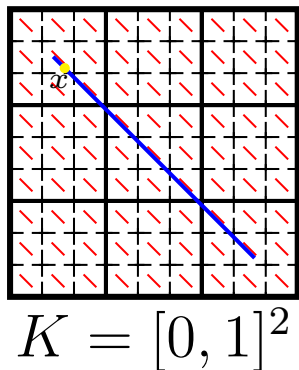
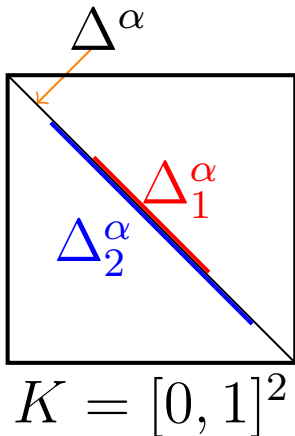
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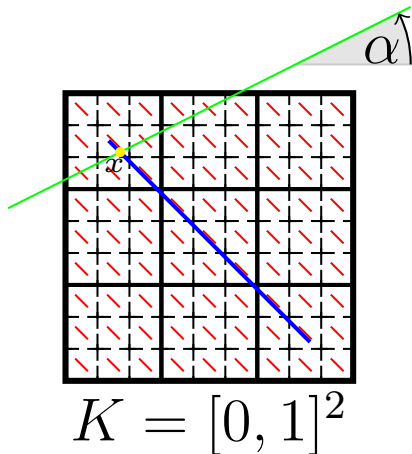
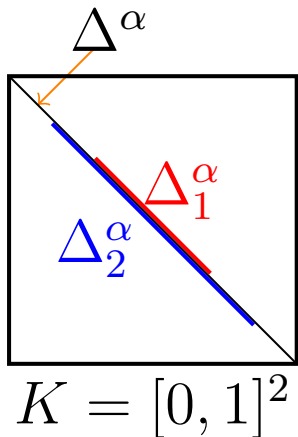


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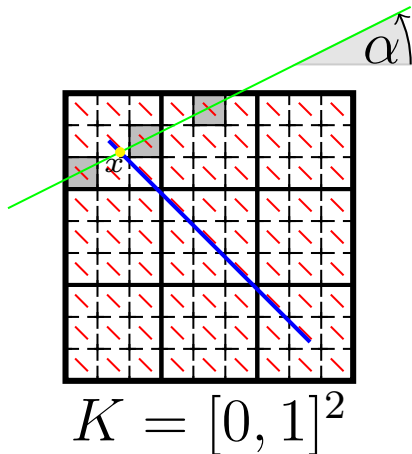
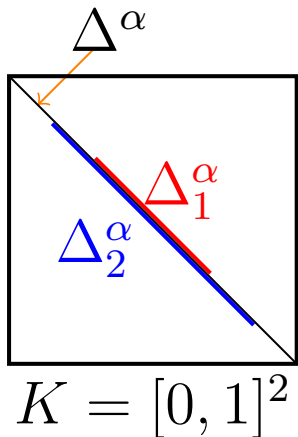




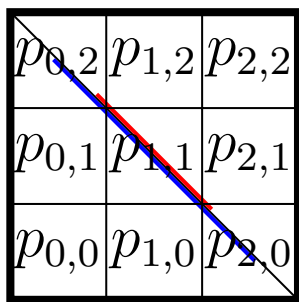
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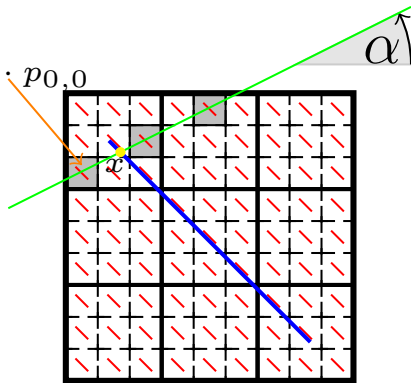


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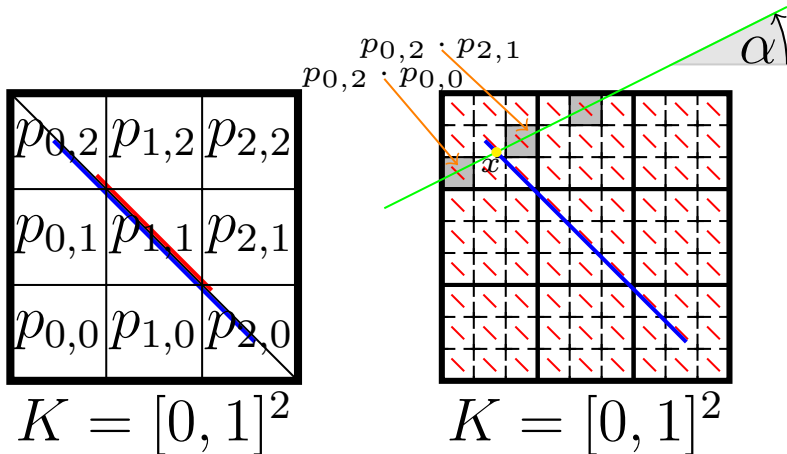
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$p_{0,2} \cdot p_{0,0}$



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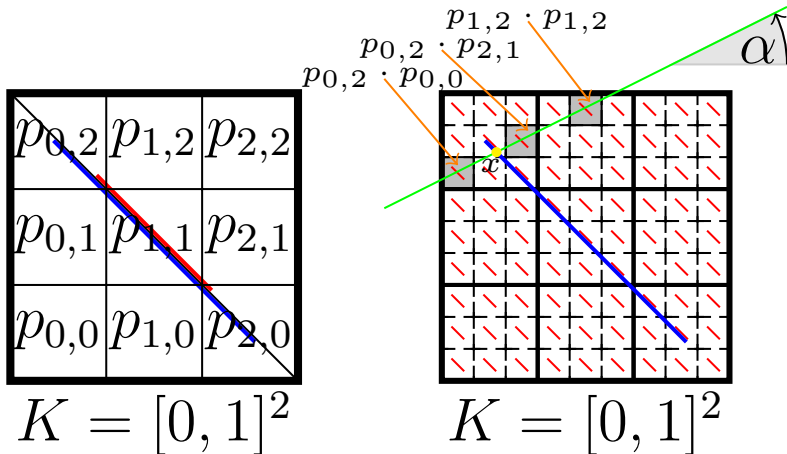
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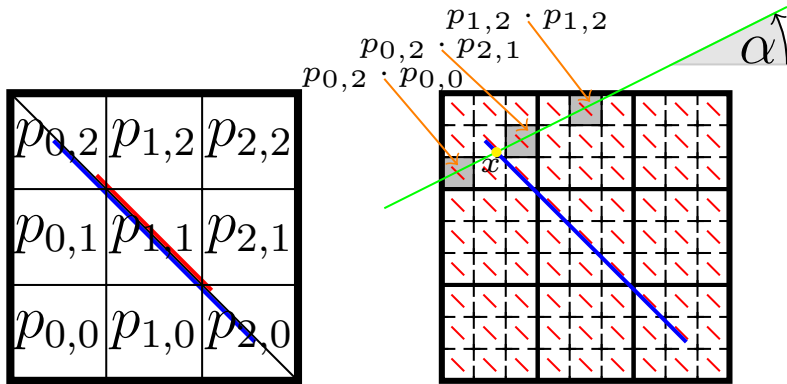
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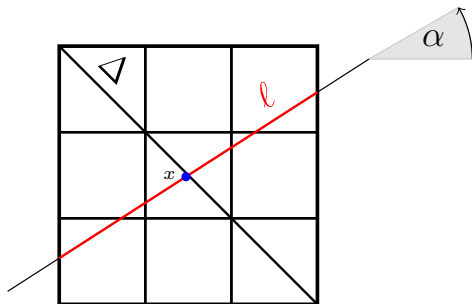


$$p_{0,2} \cdot p_{0,0} + p_{0,2} \cdot p_{2,1} + p_{1,2} \cdot p_{1,2} > 2$$

# Remarks

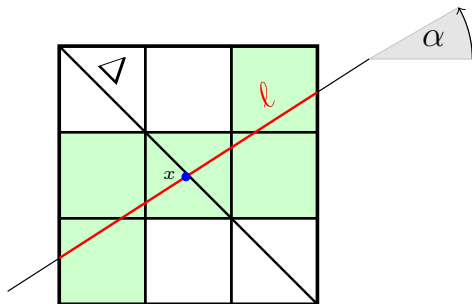
The **gray sum** is equal to the expected number of level  $r_\alpha$  **red diagonals** whose  $\Pi_\alpha$ -projection covers  $x$ .

# How to find our if $\alpha$ is a good angle?

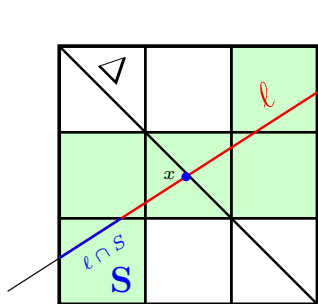




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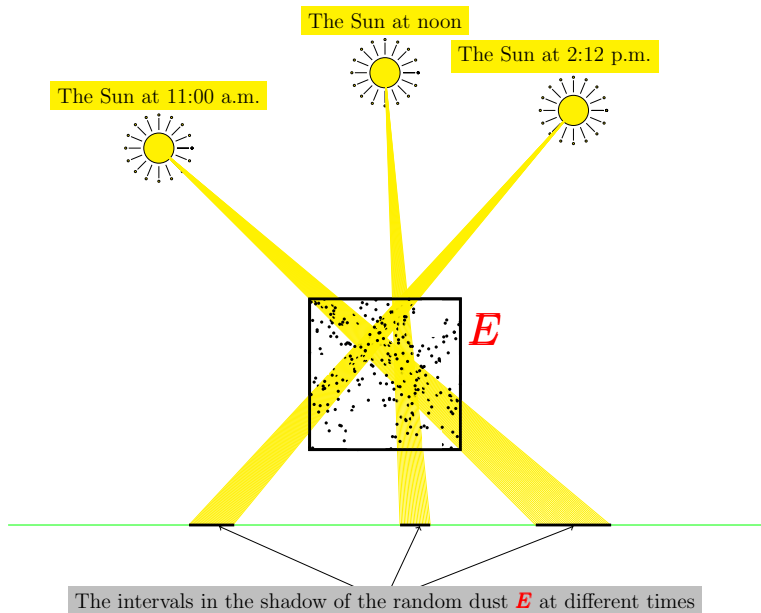
# How to find out if $\alpha$ is a good angle?



If  $\exists \varepsilon > 0$  s.t.  $\forall x \in \Delta$ :

$$\sum_{S, S \cap \ell \neq \emptyset} p_S \cdot \frac{1}{M} \cdot |\ell \cap S| \geq (1 + \varepsilon) \cdot |\ell|$$

then  $\alpha$  is a good angle.



# What happens in dimension higher than 2

## Theorem 2 (Vagó and S.)

*The same happens in dimension higher than 2 as on the plane.*

The method of the proofs is the same in higher dimension. However, there are some technical difficulties that appear in higher dimension which are not present when we work on the plane.

- 1 History
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### Theorem 3

Let  $\ell \subset \mathbb{R}^2$  be a straight line and let  $\Lambda_\ell$  be the orthogonal projection of  $\Lambda$  to  $\ell$ .

Then for almost all realizations of  $\Lambda$  (conditioned on  $\Lambda \neq \emptyset$ ) and for all straight lines  $\ell$  we have:

$$\dim_{\mathbb{H}}(\Lambda_\ell) = \dim_{\mathbb{H}}(\Lambda). \quad (1)$$

Actually much more is true:

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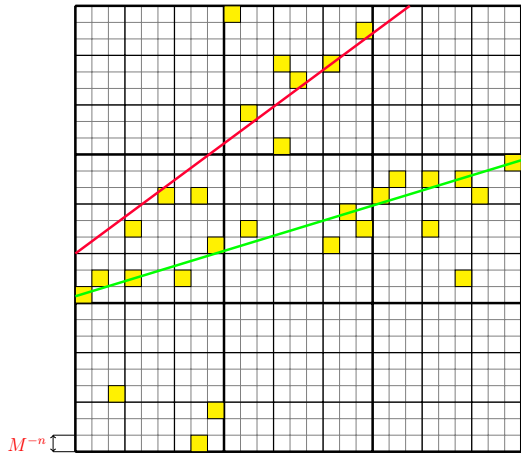
Lines intersect  $\leq c \cdot n$  squares of level  $n$

## Theorem 4 (Rams, S.)

If  $\frac{1}{M^2} < p \leq \frac{1}{M}$  then for almost all realizations of  $\Lambda$  (conditioned on  $\Lambda \neq \emptyset$ ) and for **all** straight lines  $\ell$  : there exists a constant  $C$  such that **the number of level  $n$  squares having nonempty intersection with  $\Lambda$  is at most  $c \cdot n$ .**

On the other hand, almost surely for  $n$  big enough, we can find **some** line of  $45^\circ$  angle which intersects  $\text{const} \cdot n$  level  $n$  squares.

First I draw the theorem and then I state it more precisely.



## Recall: 2

$\frac{1}{M^2} < p \leq \frac{1}{M} \Rightarrow$  Then every line  $\ell$  intersects at most  $\text{const} \cdot n$  level  $n$  squares.

# Previous theorem stated more precisely I

Recall that  $\Lambda_n$  is the union of **retained** level- $n$  squares. Let  $\Delta$  be the decreasing diagonal of the unit square  $K$  (the diagonal connecting points  $(0, 1)$  and  $(1, 0)$ ).

## Definition 5 (Slices of $\Lambda$ )

Consider the family of all lines with argument between  $0$  and  $\pi/2$  having non-empty intersection with  $\text{int}(\Delta)$ . The unit square  $K$  cuts out a line segment from each of these lines. Let  $\mathcal{L}$  be the set of all line segments obtained in this way. The sets of the form  $\Lambda \cap \ell, \ell \in \mathcal{L}$  are the **slices of  $\Lambda$** .

Let  $L_n(\ell) := |\Lambda_n \cap \ell|, \ell \in \mathcal{L}$ .

# Previous theorem stated more precisely II

Clearly,  $\mathfrak{L}$  can be presented as a countable union of families of lines segments  $\mathfrak{L}^\theta$  whose angles  $\text{Arg}(\ell)$  are  $\theta$ -separated from both 0 and  $\pi/2$ :

$$\mathfrak{L}^\theta := \left\{ \ell \in \mathfrak{L} : \min \left\{ \text{Arg}(\ell), \frac{\pi}{2} - \text{Arg}(\ell) \right\} > \theta \right\}, 0 < \theta < \frac{\pi}{4}.$$

# Previous theorem stated more precisely II

## Corollary 6

*For almost all realizations of  $E$  we have*

$$\forall \theta \in \left(0, \frac{\pi}{4}\right), \exists N, \forall n \geq N, \forall \ell \in \mathcal{L}^\theta; \# \mathcal{E}_n(\ell) \leq \text{const} \cdot n, \quad (2)$$

*where  $\mathcal{E}_n(\ell)$  is the number of selected level  $n$  squares that intersects  $\Lambda$ .*

# Large deviation estimate for $L_n(\ell)$ I

## Theorem 7 (Hoeffding)

Let  $X_1, \dots, X_m$  be independent bounded random variables with  $a_i \leq X_i \leq b_i$ , ( $i = 1, \dots, m$ ). Then for any  $t > 0$ :

$$\mathbb{P}(X_1 + \dots + X_m - \mathbb{E}[X_1 + \dots + X_m] \geq t) \leq \exp\left(\frac{-2t^2}{\sum_{i=1}^m (b_i - a_i)^2}\right).$$

# Large deviation estimate for $L_n(\ell)$ II

We apply this to prove:

## Lemma 8

*For every  $u > 1$  there is a constant  $r = r(u) > 0$  such that for every  $n \geq 1$ ,  $\ell \in \mathfrak{L}$  and  $0 < R < |\ell|$ ,*

$$\mathbb{P}(L_n(\ell) > pL_{n-1}(\ell) \cdot u \mid L_{n-1}(\ell) \geq R) < \exp(-rM^{(n-1)}R) \quad (3)$$

## Recall: 3

$$L_n(\ell) := |\Lambda_n \cap \ell|, \quad \ell \in \mathfrak{L}.$$

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# Summary

- 1 If  $0 < p \leq 1/M^2$  then  $\Lambda$  dies out in finitely many steps almost surely.
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## Definition 9

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## Theorem 10 (Rams, S.)

If  $p > \frac{1}{M}$  ( $\dim_{\mathbb{H}} \Lambda > 1$ ) then for every strictly monotonic smooth function  $f$ ,  $f(\Lambda)$  contains an interval, almost surely conditioned on non-extinction.

### Examples:

- $\{x + y : (x, y) \in \Lambda\} \supset \text{interval}.$
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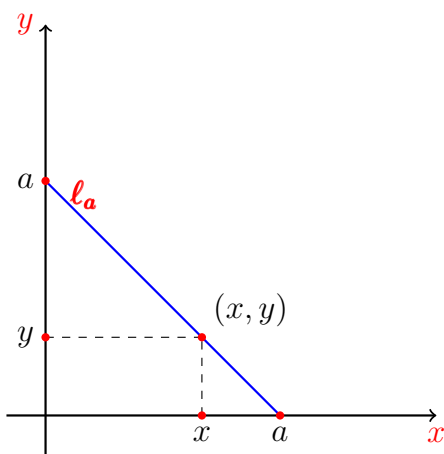
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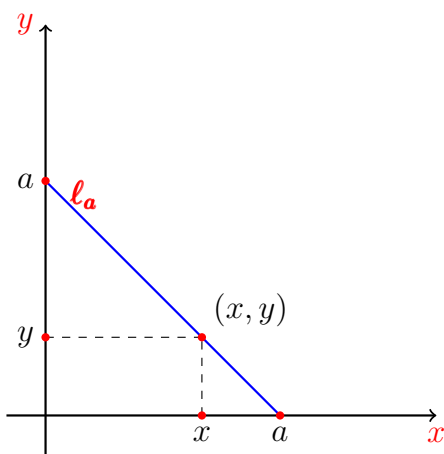
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$$\Lambda_1 + \Lambda_2 := \{x + y : x \in \Lambda_1, y \in \Lambda_2\}$$

The geometric interpretation of the arithmetic sum is:

$$\Lambda_1 + \Lambda_2 := \{a : \ell_a \cap \Lambda_1 \times \Lambda_2 \neq \emptyset\}.$$

So,  $\Lambda_1 + \Lambda_2$  is the  $45^\circ$  projection of  $\Lambda_1 \times \Lambda_2$ .



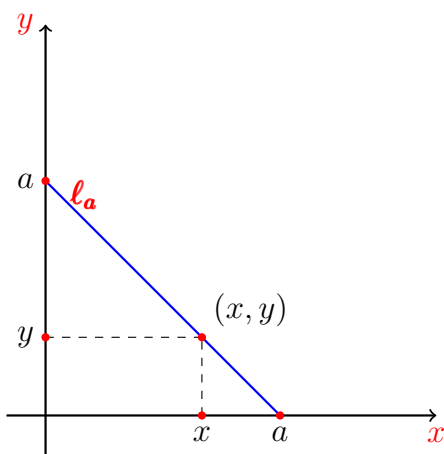
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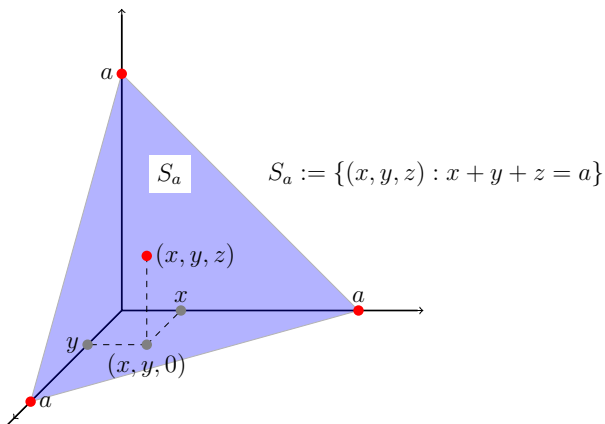
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$$a = x + y + z \iff (x, y, z) \in S_a$$

$$\Lambda_1 + \Lambda_2 + \Lambda_3 = \{a : S_a \cap \Lambda_1 \times \Lambda_2 \times \Lambda_3 \neq \emptyset\}.$$

## Recall: 4

If  $\frac{1}{M^2} < p \leq \frac{1}{M}$  then for almost all realizations of  $\Lambda$  (conditioned on  $\Lambda \neq \emptyset$ ) and for all straight lines  $\ell$  : there exists a constant  $C$  such that **the number of level  $n$  squares having nonempty intersection with  $\Lambda$  is at most  $c \cdot n$ .**

The same theorem holds if we substitute the two-dimensional Mandelbrot percolation Cantor set with the product of two independent one dimensional Cantor sets having the same  $M$  and probabilities  $p_1, p_2$  such that  $p = p_1 \cdot p_2$ .

Let  $\Lambda_1, \Lambda_2, \Lambda_3$  be one dimensional Mandelbrot percolation fractals constructed with the same  $M$  but with may be different probabilities  $p_1, p_2, p_3$ . Let  $\Lambda$  be the three dimensional Mandelbrot percolation with the same  $M$  and

$$p := p_1 p_2 p_3$$

The random Cantor sets

$$\Lambda_1 \times \Lambda_2 \times \Lambda_3 \text{ and } \Lambda$$

share many common features:

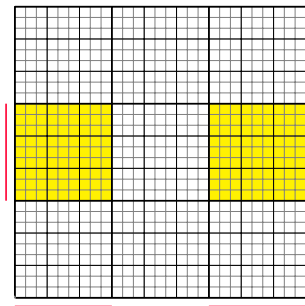
$$\dim \Lambda_1 \times \Lambda_2 \times \Lambda_3 = \dim \Lambda = \frac{\log M^3 p}{\log M}.$$

conditioned on non-extinction.

# Dependency in the product set

$$\Lambda_{123} := \Lambda_1 \times \Lambda_2 \times \Lambda_3, \quad \Lambda_{12} := \Lambda_1 \times \Lambda_2.$$

In  $\Lambda_{123}$  and in  $\Lambda_{12}$  there is NO independence between the successors of two cubes having one side common.

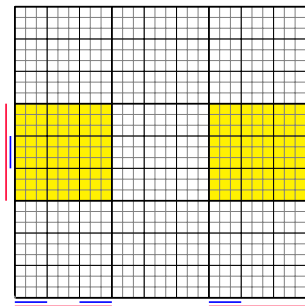




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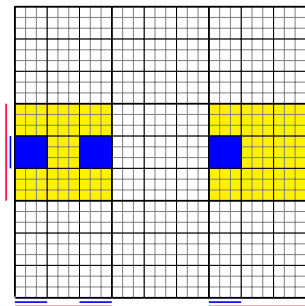
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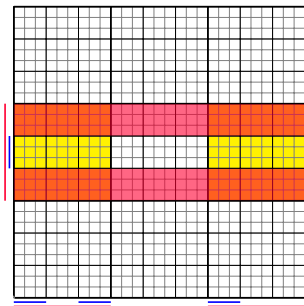
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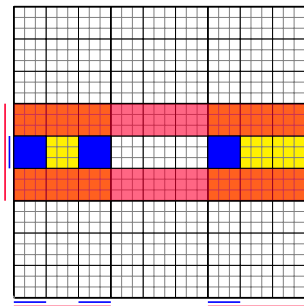
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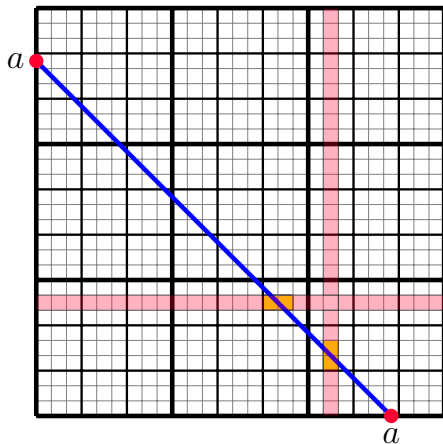
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$\Lambda$  and  $\Lambda_{12}$  are a little bit different from the point of  $45^\circ$  projection

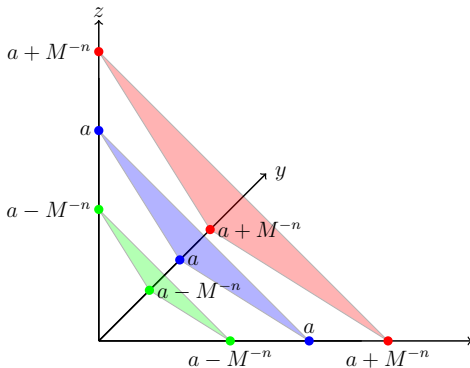


From now we focus on  $\Lambda_{123}$ :

Let  $\mathcal{E}^n$  be the set of selected level  $n$  cubes in  $\Lambda_{1,2,3}^n$ .  
 Since  $\dim_{\mathbb{B}} \Lambda_{123} > 1$  so for a  $\tau > 0$ :

$$\#\mathcal{E}^n \approx M^n \cdot M^{\tau \cdot n}.$$

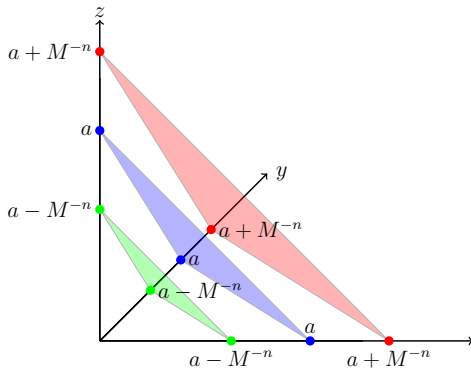
The colored planes:  $3M^n$  planes that are orthogonal to  $(1, 1, 1)$  and the consecutive ones are separated by  $M^{-n}$ . By pigeon hole principle one of the planes intersects  $\text{const} \cdot M^{\tau n}$  selected level  $n$  cubes. Assume that this is the **blue plane**.



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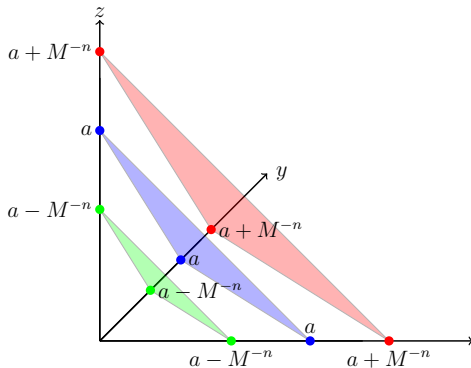
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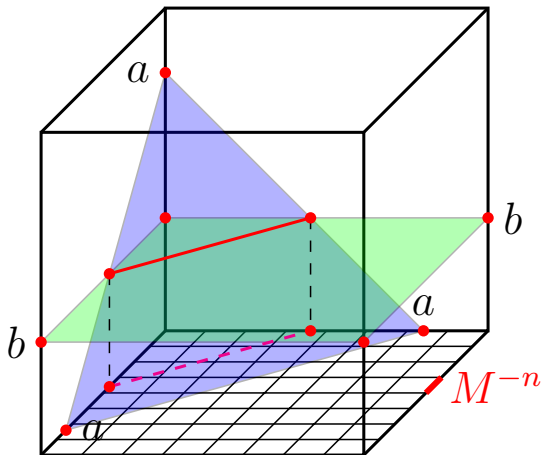
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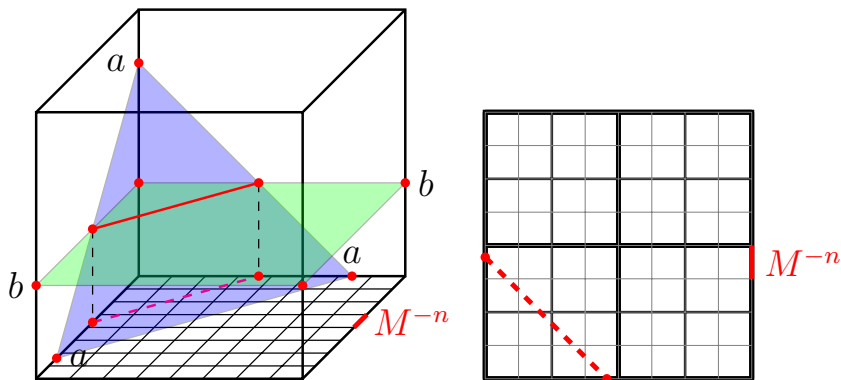




Among the  $M^{\tau^n}$  cubes which intersect the blue plane the ones sharing one common side are NOT independent.  
 For example those who intersect the red line are NOT independent.



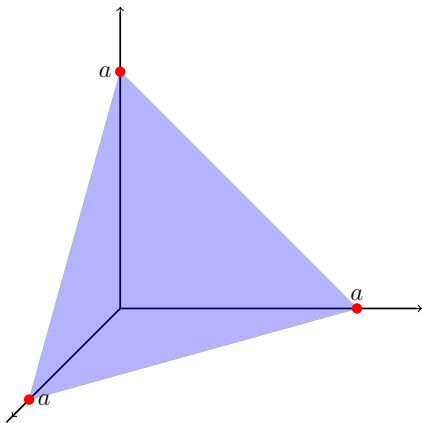
$\dim_{\text{H}} \Lambda_{123} > 1$  but  $\dim_{\text{H}} \Lambda_{12}, \dim_{\text{H}} \Lambda_{23}, \dim_{\text{H}} \Lambda_{31} < 1$ .



The point is that on the red dashed line there could be potentially  $M^n$  selected level  $n$  squares but in reality there will be only  $c \cdot n$  selected squares.

An easy combinatorial Lemma shows that for a  $t > 0$  constant there are  $M^{nt}$  selected level  $n$  squares that have

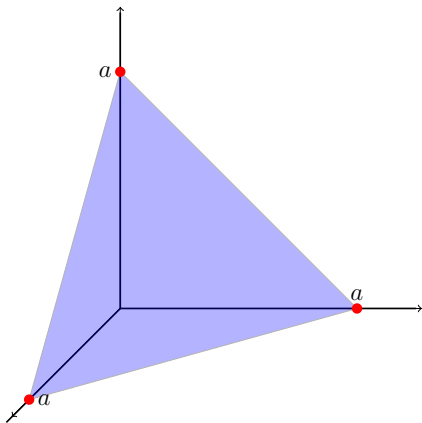
- no common sides (so what ever happens in these cubes in the future is independent)
- such that they all intersect the blue plane.



Then we use Large deviation theory similarly to Falconer Grimett to get intervals in the projection.

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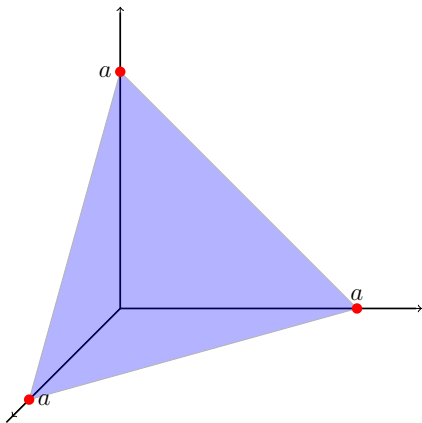
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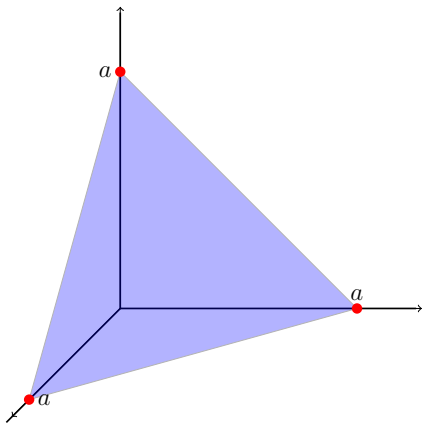
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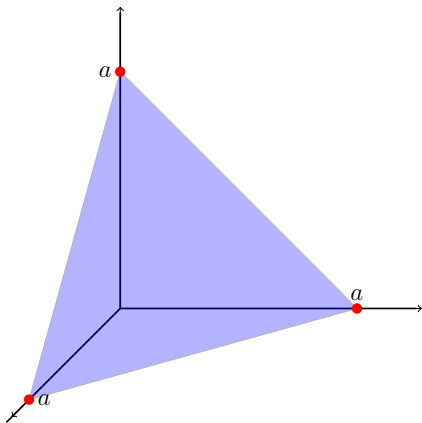
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