Dynamique euclidienne : une approche symbolique

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Séminaire de combinatoire Philippe Flajolet

# Euclid algorithm...

and...

- continued fractions
- dynamical analysis, costs
- symbolic dynamics : the Sturmian case
- higher-dimensional generalizations



Analysis of algorithms

An algorithm

Euclid algorithm

According to Knuth

'the granddaddy of all algorithms, because it is the oldest nontrivial algorithm that has survived to the present day'

J. Shallit-Origins of the Analysis of the Euclidean Algorithm-Historia Mathematica (1994)

## **Euclidean dynamics**

An algorithm Euclid algorithm

together with a dynamical system

Gauss map

 $T\colon [0,1]\to [0,1],\ x\mapsto \{1/x\}$ 

## Euclid algorithm

We start with two nonnegative integers  $u_0$  and  $u_1$ 

$$u_0 = u_1 \left[ \frac{u_0}{u_1} \right] + u_2$$
$$u_1 = u_2 \left[ \frac{u_1}{u_2} \right] + u_3$$
$$\vdots$$
$$u_{m-1} = u_m \left[ \frac{u_{m-1}}{u_m} \right] + u_{m+1}$$
$$u_{m+1} = \gcd(u_0, u_1)$$
$$u_{m+2} = 0$$

### Euclid algorithm and continued fractions

We start with two coprime integers  $u_0$  and  $u_1$ 

$$u_{0} = u_{1}a_{1} + u_{2}$$

$$\vdots$$

$$u_{m-1} = u_{m}a_{m} + u_{m+1}$$

$$u_{m} = u_{m+1}a_{m+1} + 0$$

$$u_{m+1} = 1 = \gcd(u_{0}, u_{1})$$

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Euclid's algorithm yields the digits for the continued fraction expansion of  $\frac{u_1}{u_0}$ 

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$$u_{m+1} = 1 = gcd(u_{0}, u_{1})$$



#### Continued fractions and dynamical systems

Consider the Gauss map

$$T: [0, 1] \to [0, 1], \ x \mapsto \{1/x\}$$
$$x_1 = T(x) = \{1/x\} = \frac{1}{x} - \left[\frac{1}{x}\right] = \frac{1}{x} - a_1$$
$$x = \frac{1}{a_1 + x_1} \qquad a_n = \left[\frac{1}{T^{n-1}x}\right]$$
$$x = \frac{1}{a_1 + \frac{1}{a_2 + \frac{1}{a_3 + \cdots}}}$$

#### Continued fractions and dynamical systems Consider the Gauss map



 $T: [0,1] \to [0,1], \ x \mapsto \{1/x\}$ 

We are given a dynamical system

 $T \colon X \to X$ 

Discrete stands for discrete time

We consider orbits/trajectories of points of X under the action of the map T

 $\{T^n x \mid n \in \mathbb{N}\}$ 

How well are the orbits distributed? According to which measure?

#### Continued fractions and ergodicity Ergodicity has to do with the long term statistical behaviour of orbits

Continued fractions and ergodicity Ergodicity has to do with the long term statistical behaviour of orbits

The Gauss map is ergodic with respect to the Gauss measure

$$\mu(B) = \frac{1}{\log 2} \int_{B} \frac{1}{1+x} dx$$
$$\mu(B) = \mu(T^{-1}B) \quad T\text{-invariance}$$
$$T^{-1}B = B \implies \mu(B) = 0 \text{ or } 1 \text{ ergodicity}$$

$$\frac{1}{n}\sum_{j=0}^{n-1}f(T^{j}x)=\int fd\mu \text{ ergodic theorem}$$

The mean behaviour along an orbit= the mean value of *f* with respect to  $\mu$ 

#### Measure-theoretic results

Gauss measure

$$\mu(A) = \frac{1}{\log 2} \int_A \frac{dx}{1+x}$$

• Convergents

For a.e. *x*, 
$$\lim \frac{\log q_n}{n} = \frac{\pi^2}{12 \log 2}$$

• Densities of partial quotients

For a.e. x and  $a \ge 1$ 

$$\lim_{N \to \infty} \frac{1}{N} \{ k \le N; \ a_k = a \} = \frac{1}{\log 2} \log \frac{(a+1)^2}{a(a+2)}$$

### Rational vs. irrational parameters

Euclid algorithm  $\rightsquigarrow$  gcd  $\rightsquigarrow$  rational parameters Continued fractions  $\rightsquigarrow$  irrational parameters

## Rational vs. irrational parameters

Euclid algorithm  $\rightsquigarrow$  gcd  $\rightsquigarrow$  rational parameters Continued fractions  $\rightsquigarrow$  irrational parameters

- When computing a gcd, we work with integer/rational parameters
- This set has zero measure
- Ergodic methods produce results that hold only almost everywhere

Is it relevant to compare generic orbits and orbits for integer parameters?

# Dynamical analysis of

# Euclid algorithm



Number of steps  $\ell(u, v)$ 

 $\ell(u, v)$  : number of steps in Euclid algorithm 0 < v < u

Worst case

 $\ell(u, v) = O(\log v)$  ( $\leq 5 \log_{10} v$ , Lamé 1844) Reynaud 1821 [ $\ell(u, v) < v/2$ ], see Shallit's survey

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#### Worst case

 $\ell(u, v) = O(\log v)$  ( $\leq 5 \log_{10} v$ , Lamé 1844)

• Mean case 
$$0 < v < u \le N$$
  $\gcd(u, v) = 1$  $\mathbb{E}_N(\ell) \sim rac{12\log 2}{\pi^2} \cdot \log N + \eta$ 

[see Knuth, Vol. 2]

## Number of steps $\ell(u, v)$

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#### Worst case

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• Mean case  $0 < v < u \le N$  gcd(u, v) = 1

$$\frac{12\log 2}{\pi^2} \cdot \log \textit{N} + \eta + \textit{O}(\textit{N}^{-\gamma})$$

asymptotically normal distribution

[Heilbronn'69, Dixon'70, Porter'75, Hensley'94, Baladi-Vallée'05...]

#### Distributional dynamical analysis $gcd(u_0, u_1) = 1$ $N \ge u_0 > u_1 > \cdots$ $u_{k-1} = a_k u_k + u_{k+1}$

Cost of moderate growth  $c(a) = O(\log a)$ 

- Number of steps in Euclid algorithm  $c \equiv 1$
- Number of occurrences of a quotient  $c = 1_a$
- Binary length of a quotient  $c(a) = \log_2(a)$

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Theorem [Baladi-Vallée'05]

$$\mathbb{E}_{N}[\text{Cost}] = \frac{12\log 2}{\pi^{2}} \cdot \hat{\mu}(\text{Cost}) \cdot \log N + O(1)$$

The distribution is asymptotically Gaussian (CLT)

Discrete framework-Euclid algorithm

# **Ergodic theorem**

Theorem [Baladi-Vallée'05]

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## **Ergodic theorem**

Theorem [Baladi-Vallée'05]

$$\mathbb{E}_{N}[\text{Cost}] = \frac{12\log 2}{\pi^{2}} \cdot \hat{\mu}(\text{Cost}) \cdot \log N + O(1)$$

$$\mathbb{E}_{N}[c] = \frac{\text{dimension}}{\text{entropy}} \cdot \hat{\mu}(c) \cdot \log N + O(1)$$
$$\hat{\mu}(c) = \int_{0}^{1} c([1/x]) \frac{1}{\log 2} \frac{1}{1+x} dx$$

#### Continuous framework-truncated trajectories

#### Cost of truncated trajectories

Cost of moderate growth

$$c(a_i) = O(\log a_i)$$
 for  $a_i$  partial quotient



# Cost of truncated trajectories

Cost of moderate growth

 $c(a_i) = O(\log a_i)$  for  $a_i$  partial quotient

Cost of a truncated trajectory

$$C_n(x) = \sum_{i=1}^n c(a_i(x)) \qquad a_i = \left[\frac{1}{T^{i-1}(x)}\right]$$

According to the ergodic theorem, for a.e.  $x \in [0, 1]$ 

 $C_n(x)/n o \hat{\mu}(x)$ 

$$\hat{\mu}(C) = \int_0^1 c\left(\left[\frac{1}{x}\right]\right) \cdot \frac{1}{\log 2} \frac{1}{1+x} \cdot dx$$
$$\mathbb{E}_N[C] = \frac{2}{\pi^2/(6\log 2)} \cdot \hat{\mu}(C) \cdot \log N$$

#### Dynamical analysis of algorithms [Vallée] It belongs to the area of

• Analysis of algorithms [Knuth'63]

probabilistic, combinatorial, and analytic methods

Analytic combinatorics [Flajolet-Sedgewick]



generating functions and complex analysis, analytic functions, analysis of the singularities

# Dynamical analysis of algorithms [Vallée]

It mixes tools from

• dynamical systems (transfer operators, density transformers, Ruelle-Perron-Frobenius operators)

• analytic combinatorics (generating functions of Dirichlet type)

the singularities of (Dirichlet) generating functions are expressed in terms of transfer operators

# Euclidean dynamics [Vallée]

One starts with a discrete algorithm

• This algorithm is extended into a continuous one in terms of a dynamical system

Orbits/trajectories = executions

• Main parameters of the algorithm are studied in the continuous framework

rational trajectories ↔ generic trajectories

• One comes back to the discrete algorithm

#### A transfer from continuous to discrete

'The probabilistic behaviour of gcd algorithms is quite similar to the behaviour of their continuous counterparts'

#### Rational vs. irrational parameters

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Is it relevant to compare generic orbits and orbits for integer parameters?

## Rational vs. irrational parameters

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Is it relevant to compare generic orbits and orbits for integer parameters?

Average-case analysis vs. a.e. results

Fact Orbits of rational points tend to behave like generic orbits

And their probabilistic bevaviour can be captured thanks to the methods of dynamical analysis of algorithms

# Gauss map



# symbolic dynamics



We are given a dynamical system

 $T \colon X \to X$ 

We are given a dynamical system

$$T: X \to X$$

We consider orbits/trajectories of points of X under the action of the map T

 $\{T^n x \mid n \in \mathbb{N}\}$ 

We are given a dynamical system

$$T\colon X\to X$$

We partition X in to a finite number of subsets  $X = \bigcup_{i=1}^{d} X_i$ We code the trajectory of a point x with respect to  $(X_i)$ 

$$\{T^n x \mid n \in \mathbb{N}\} \rightsquigarrow (u_n)_{n \in \mathbb{N}} \in \{1, 2, \cdots, d\}^{\mathbb{N}}$$

We are given a dynamical system

 $T: X \to X$ 

We code the trajectory of a point x with respect to  $(X_i)$ 

$$\{T^n x \mid n \in \mathbb{N}\} \rightsquigarrow (u_n)_{n \in \mathbb{N}} \in \{1, 2, \cdots, d\}^{\mathbb{N}}$$

The map acting on  $\{1, 2, \cdots, d\}^{\mathbb{N}}$  is the shift *S* 

$$S((u_n)_n) = (u_{n+1})_n$$

 $(X, T) \rightsquigarrow (Y, S)$  with  $Y \subset \{1, 2, \cdots, d\}^{\mathbb{N}}$ 

From geometric dynamical systems to symbolic dynamical systems and backwards
# Arithmetic dynamics

Arithmetic dynamics [Sidorov-Vershik] arithmetic codings of dynamical systems that preserve their arithmetic structure

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Example Let  $R_{\alpha} \colon \mathbb{R}/\mathbb{Z} \to \mathbb{R}/\mathbb{Z}, x \mapsto x + \alpha \mod 1$ One codes trajectories according to the finite partition

{
$$I_0 = [0, 1 - \alpha[, I_1 = [1 - \alpha, 1[]$$
}



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This yields a measure-theoretic isomorphism

$$(\pmb{R}_lpha,\mathbb{R}/\mathbb{Z})\sim(\pmb{X}_lpha,\pmb{S})$$

where *S* is the shift and  $X_{\alpha} \subset \{0, 1\}^{\mathbb{N}}$ 

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$$I_0 = [0, 1 - \alpha[, I_1 = [1 - \alpha, 1[]$$
}



[Lothaire, Algebraic combinatorics on words, N. Pytheas Fogg, Substitutions in dynamics, arithmetics and combinatorics

CANT Combinatorics, Automata and Number theory]

### 0110110101101101

### 0110110101101101

### 11 and 00 cannot occur simultaneously



### 0110110101101101

One considers the substitutions

$$\sigma_0: \mathbf{0} \mapsto \mathbf{0}, \ \sigma_0: \mathbf{1} \mapsto \mathbf{10}$$
$$\sigma_1: \mathbf{0} \mapsto \mathbf{01}, \ \sigma_1: \mathbf{1} \mapsto \mathbf{1}$$

One has

01 1 01 1 01 01 1 01 1 01 =  $\sigma_1(0101001010)$ 0 10 10 0 10 10 =  $\sigma_0(011011)$ 01 1 01 1 =  $\sigma_1(0101)$ 01 01 =  $\sigma_1(00)$ 

### 0110110101101101

One considers the substitutions

$$\sigma_0: \mathbf{0} \mapsto \mathbf{0}, \ \sigma_0: \mathbf{1} \mapsto \mathbf{10}$$
$$\sigma_1: \mathbf{0} \mapsto \mathbf{01}, \ \sigma_1: \mathbf{1} \mapsto \mathbf{1}$$

The Sturmian words of slope  $\alpha$  are provided by an infinite composition of substitutions

$$\lim_{n \to +\infty} \sigma_0^{a_1} \sigma_1^{a_2} \cdots \sigma_{2n}^{a_{2n}} \sigma_{2n+1}^{a_{2n+1}}(\mathbf{0})$$

where the  $a_i$  are produced by the continued fraction expansion of  $\alpha$ 

### 0110110101101101



## Euclid algorithm and discrete segments

$$\begin{array}{rcl}
11 &=& 2 \cdot 4 + 3 \\
4 &=& 1 \cdot 3 + 1 \\
3 &=& 3 \cdot 1 + 0
\end{array}$$



$$(11,4) \stackrel{\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{2}}{\longleftarrow} (3,4) \stackrel{\begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}}{\longleftarrow} (3,1) \stackrel{\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{3}}{\longleftarrow} (0,1)$$
$$\overset{a \mapsto a}{\underbrace{b \mapsto aab}} \stackrel{b \mapsto b}{\underbrace{b \mapsto b}} \stackrel{ab}{\underbrace{b \mapsto aaab}} \stackrel{a \mapsto a}{\underbrace{b \mapsto aaab}} w_{3} = b$$

### Euclid algorithm and discrete segments



$$(11,4) \underbrace{\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{2}}_{b \mapsto aab} (3,4) \underbrace{\begin{pmatrix} 1 & 0 \\ 1 & 1 \end{pmatrix}}_{b \mapsto b} (3,1) \underbrace{\begin{pmatrix} 1 & 1 \\ 0 & 1 \end{pmatrix}^{3}}_{b \mapsto aab} (0,1)$$
$$\mathbf{w} = \mathbf{w}_{0} \underbrace{\longleftarrow}_{b \mapsto aab} \mathbf{w}_{1} \underbrace{\longleftarrow}_{b \mapsto b} \mathbf{w}_{2} \underbrace{\longleftarrow}_{b \mapsto aaab} \mathbf{w}_{3} = b$$

## Higher-dimensional framework

- How to discretize a line in the space?
- How to compute the gcd of three or more numbers?
- How to compare gcd/cf algorithms?
- Integer parameters vs. rational parameters
- Can we generalize the Sturmian framework to translations on T<sup>d</sup>?

### The Tribonacci fractal

The Tribonacci substitution  $\sigma$ : 1  $\mapsto$  12, 2  $\mapsto$  13, 3  $\mapsto$  1

 $\sigma^{\infty}(1) = 121312112\cdots$ 

One represents  $\sigma^{\infty}(1)$  as a broken line

$$1\mapsto \vec{e}_1,\; 2\mapsto \vec{e}_2,\; 3\mapsto \vec{e}_3,\;$$

that we will be projected according to the eigenspaces of  $M_{\sigma} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix}$ 

# Rauzy fractal and dynamics

One first defines an exchange of pieces acting on the Rauzy fractal



### Rauzy fractal and dynamics

One first defines an exchange of pieces acting on the Rauzy fractal.

This due to the fact that the subtiles are disjoint in

measure



This exchange of pieces factorizes into a translation of  $\mathbb{T}^2$ This due to the fact that the Rauzy fractal tiles periodically the plane



### $\sigma: \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$ $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$

### σ: 1 ↦ 12, 2 ↦ 1312, 3 ↦ 112 $σ^{∞}(1) = 12131212112...$



$$σ: 1 ↦ 12, 2 ↦ 1312, 3 ↦ 112$$
  
 $σ^{∞}(1) = 12131212112...$ 

#### Trajectories are coded according to the partition



### $\sigma: \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$ $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$

Trajectory : 2



$$\sigma: \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$$
  
 $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$ 

Trajectory : 21



$$\sigma \colon \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$$
$$\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$$

Trajectory: 213



$$\sigma: 1 \mapsto 12, 2 \mapsto 1312, 3 \mapsto 112$$
  
 $\sigma^{\infty}(1) = 12131212112...$ 

Trajectory : 2131



$$\sigma \colon \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$$
$$\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$$

Trajectory : 21312



### $\sigma: \mathbf{1} \mapsto \mathbf{12}, \mathbf{2} \mapsto \mathbf{1312}, \mathbf{3} \mapsto \mathbf{112}$ $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$

Trajectory 2 : 213121



### $\sigma: \mathbf{1} \mapsto \mathbf{12}, \mathbf{2} \mapsto \mathbf{1312}, \mathbf{3} \mapsto \mathbf{112}$ $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131212112...}$

#### Trajectory : 2131212



$$\sigma: \mathbf{1} \mapsto \mathbf{12}, \ \mathbf{2} \mapsto \mathbf{1312}, \ \mathbf{3} \mapsto \mathbf{112}$$
  
 $\sigma^{\infty}(\mathbf{1}) = \mathbf{12131211212...}$ 

#### Density and even equidistribution of orbits



### **Tribonacci rotation** $\sigma$ : 1 $\mapsto$ 12, 2 $\mapsto$ 13, 3 $\mapsto$ 1

### **Theorem** [Rauzy, Chekhovaya-Hubert-Messaoudi]

•  $(X_{\sigma}, S)$  is measure-theoretically isomorphic with a two-dimensional translation and is equal to the codings of the orbits under the action of the

$$R_{\beta}:\mathbb{T}^2
ightarrow\mathbb{T}^2,\;x\mapsto x+(1/eta,1/eta^2)$$

with respect to the pieces of the Rauzy fractal

### **Tribonacci rotation** $\sigma$ : 1 $\mapsto$ 12, 2 $\mapsto$ 13, 3 $\mapsto$ 1 **Theorem** [Rauzy, Chekhovaya-Hubert-Messaoudi]

 (X<sub>σ</sub>, S) is measure-theoretically isomorphic with a two-dimensional translation and is equal to the codings of the orbits under the action of the translation

$${\it R}_eta:\mathbb{T}^2 o\mathbb{T}^2,\; x\mapsto x+(1/eta,1/eta^2)$$

with respect to the pieces of the Rauzy fractal

 The points of the broken line corresponding to σ<sup>n</sup>(1), n ∈ N, produce the sequence of best approximations for the vector (<sup>1</sup>/<sub>β</sub>, <sup>1</sup>/<sub>β<sup>2</sup></sub>) for a given norm associated with

the incidence matrix  $M_{\sigma}$ 

# S-adic Rauzy fractals

We want to find

'good' symbolic codings for *d*-dimensional translations

$$R_{(\alpha_1,\cdots,\alpha_d)} \colon \mathbb{T}^d \to \mathbb{T}^d$$

• 'good' partitions of the torus  $\mathbb{T}^d$ 

Take a multidimensional continued fraction algorithm and transform it into substitutions

[B.-Steiner-Thuswaldner, B.-Jolivet-Siegel, Arnoux-B.-Labbé]

# Comparing Euclid/cf algorithms

 Number of steps and costs functions for algorithms defined on rational entries

worst-case, mean behavior, average-case analysis

- Convergence properties
- Ergodic properties

ergodic invariant measure, natural extension

Arithmetic properties

cubic numbers and periodic expansions, Diophantine approximation

### Multidimensional Euclid's algorithms

 Jacobi-Perron We subtract the first one to the two other ones with 0 ≤ u<sub>1</sub>, u<sub>2</sub> ≤ u<sub>3</sub>

$$(u_1, u_2, u_3) \mapsto (u_2 - [\frac{u_2}{u_1}]u_1, u_3 - [\frac{u_3}{u_1}]u_1, u_1)$$

• Brun We subtract the second largest entry and we reorder. If  $u_1 \le u_2 \le u_3$ 

$$(u_1, u_2, u_3) \mapsto (u_1, u_2, u_3 - u_2)$$

Poincaré We subtract the previous entry and we reorder

$$(u_1, u_2, u_3) \mapsto (u_1, u_2 - u_1, u_3 - u_2)$$

- Selmer We subtract the smallest to the largest and we reorder  $(u_1, u_2, u_3) \mapsto (u_1, u_2, u_3 - u_1)$
- Fully subtractive We subtract the smallest one to the other ones and we reorder

$$(u_1, u_2, u_3) \mapsto (u_1, u_2 - u_1, u_3 - u_1)$$

Consider parameters  $(u_1, \cdots, u_d)$  with  $0 \le u_1, \cdots, u_d \le N$ 

**Thm** Expectation of the number of steps =  $\frac{\text{dimension}}{\text{Entropy}} \times \log N$ 

Dimension

• d= Number of parameters

### Number of steps

Consider parameters  $(u_1, \cdots, u_d)$  with  $0 \le u_1, \cdots, u_d \le N$ 

**Thm** Expectation of the number of steps =  $\frac{\text{dimension}}{\text{Entropy}} \times \log N$ 

• Euclid algorithm

$$\frac{2}{\pi^2/(6\log 2)}\log N$$

[Heilbronn'69,Dixon'70,Hensley'94,Baladi-Vallée'03,Lhote-Vallée'08,...]

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Jacobi-Perron

[Fischer-Schweiger'75]

Brun

[B.-Lhote-Vallée, work in progress]
#### Number of steps

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**Thm** Expectation of the number of steps =  $\frac{\text{dimension}}{\text{Entropy}} \times \log N$ 

• Formal power series with coefficients in a finite field and polynomials with degree less than *m* 

$$\frac{2}{2\frac{q}{q-1}}m=\frac{q-1}{q}m$$

[Knopfmacher-Knopfmacher'88, Friesen-Hensley'96, Lhote-Vallée'06'08, B.-Nakada-Natsui-Vallée'12]

### Formal power series

Let q be a power of a prime number p

We have the correspondence

• 
$$\mathbb{Z} \sim \mathbb{F}_q[X]$$
  
•  $\mathbb{Q} \sim \mathbb{F}_q(X)$   
•  $\mathbb{R} \sim \mathbb{F}_q((X^{-1}))$   
 $f = a_n X^n + a_{n-1} X^{n-1} + \dots + a_0 + a_{-1} X^{-1} + \dots$ 

Laurent formal power series

## Formal power series

Let 
$$f \in \mathbb{F}_q((X^{-1}))$$
  $f \neq 0$   
 $f = a_n X^n + a_{n-1} X^{n-1} + \cdots \qquad a_n \neq 0$ 

Degree deg *f* = *n* Distance |*f*| = *q*<sup>deg *f*</sup>

Ultrametric space

 $|f+g| \leq \max(|f|, |gl)$ 

No carry propagation !

#### **Continued fractions**

One can expand series f into continued fractions

$$f = a_0(X) + \frac{1}{a_1(X) + \frac{1}{a_2(X) + \frac$$

The digits  $a_i(X)$  are polynomials of positive degree

$$a_k \geq 1 \rightsquigarrow \deg a_k(X) \geq 1$$

- Unique expansion even if *f* does not belong to  $\mathbb{F}_q(X)$
- Finite expansion iff  $f \in \mathbb{F}_q(X)$
- But there exist explicit examples of algebraic series with bounded partial quotients [Baum-Sweet]
- Roth's theorem does not hold for algebraic series (see e.g. [Lasjaunias-de Mathan])

[B.-Nakada, Expositiones Mathematicae]

# Why is everything simpler?

Ultrametric space !

- Digits are equidistributed : the Haar measure is invariant
- Hence, understanding the 'polynomial case' can help the understanding of the 'integer case'

## And now..

- Numeration dynamics  $T_{\beta}$ :  $x \mapsto \{\beta x\}$
- Discrete lines and planes
- Invariant measures



