

# ANALYSIS of EUCLIDEAN ALGORITHMS

*An Arithmetical Instance of Dynamical Analysis*

*Dynamical Analysis:=*

*Analysis of Algorithms + Dynamical Systems*

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## Dynamical Analysis –main principles.

**Input.-** A discrete algorithm.

**Step 1.-** Extend the discrete algorithm into a continuous process, i.e. a dynamical system.  $(X, V)$   $X$  compact,  $V : X \rightarrow X$ , where the discrete alg. gives rise to particular trajectories.

**Step 2.-** Study this (continuous) Dynamical system, via its generic trajectories. A main tool: the transfer operator.

**Step 3.-** Coming back to the algorithm: Use the transfer operator as a generating operator, and prove that the particular trajectories due to the algorithm behave as the generic trajectories.

**Output.-** Probabilistic analysis of the Algorithm.

## **Plan of the talk.**

I – Four types, six instances of Euclidean algorithms

II – The average-case analysis: The results.

III – The dynamical systems underlying the algorithms.

IV – The method: Dynamical Analysis

**I – Four types, six instances of Euclidean algorithms**

## The Euclid Algorithm: the grand father of all the algorithms.

On the input  $(u, v)$ , it computes the **gcd** of  $u$  and  $v$ , together with the **Continued Fraction Expansion** of  $u/v$ .  $u_0 := v$ ;  $u_1 := u$ ;  $u_0 \geq u_1$

$$\left\{ \begin{array}{l} u_0 = m_1 u_1 + u_2 \quad 0 \leq u_2 < u_1 \\ u_1 = m_2 u_2 + u_3 \quad 0 \leq u_3 < u_2 \\ \dots = \dots + \dots \\ u_{p-2} = m_{p-1} u_{p-1} + u_p \quad 0 \leq u_p < u_{p-1} \\ u_{p-1} = m_p u_p + 0 \quad u_{p+1} = 0 \end{array} \right.$$

$u_p$  is the **gcd** of  $u$  and  $v$ , the  $m_i$ 's are the **digits**.  $p$  is the **depth**.

CFE of  $\frac{u}{v}$ :

$$\frac{u}{v} = \frac{1}{m_1 + \frac{1}{m_2 + \frac{1}{\dots + \frac{1}{m_p}}}},$$

## The extended Euclid Algorithm

also returns the Bezout pair  $(r, s)$  for which  $d = rv + su$ .

It computes the sequence  $s_i$  defined by

$$s_0 = 0, \quad s_1 = 1, \quad s_i = s_{i-2} - s_{i-1} \cdot m_{i-1}, \quad 0 \leq i < p.$$

The last element  $s_p$  is the Bezout coefficient  $s$ .

Used for **computing modular inverses**: crucial in cryptography.

## A Euclidean algorithm:=

any algorithm which performs a **sequence of divisions**  $v = mu + r$ .

### Various possible types of Euclidean divisions

– **MSB divisions** [directed by the **Most Significant Bits**]  
shorten integers on the **left**,

and provide a remainder  $r$  smaller than  $u$ ,

(w.r.t the **usual** absolute value), i.e. with more zeroes on the **left**.

– **LSB divisions** [directed by the **Least Significant Bits**]  
shorten integers on the **right**,

and provide a remainder  $r$  smaller than  $u$

(w.r.t the **dyadic** absolute value), i.e. with more zeroes on the **right**.

– **Mixed divisions**

shorten integers both on the **right** and on the **left**,

with new zeroes both on the **right** and on the **left**.

## Instances of MSB Algorithms.

- Variants according to the position of remainder  $r$ ,

**By Default:**  $v = mu + r$  with  $0 \leq r < u$

**By Excess:**  $v = mu - r$  with  $0 \leq r < u$

**Centered:**  $v = mu + \epsilon r$  with  $\epsilon = \pm 1, 0 \leq r \leq u/2$

- **Subtractive** Algorithm :

A **division** with quotient  $m$  can be replaced by  $m$  **subtractions**

While  $v > u$  do  $v := v - u$



## An instance of a Mixed Algorithm.

The Subtractive Algorithm,

where the zeroes on the right are removed from the remainder defines the Binary Algorithm.

### Subtractive Gcd Algorithm.

**Input.**  $u, v; v \geq u$

While  $(u \neq v)$  do

    While  $v > u$  do

$v := v - u$

    Exchange  $u$  and  $v$ .

**Output.**  $u$  (or  $v$ ).

### Binary Gcd Algorithm.

**Input.**  $u, v$  odd;  $v \geq u$

While  $(u \neq v)$  do

    While  $v > u$  do

$k := \nu_2(v - u);$

$v := \frac{v - u}{2^k};$

    Exchange  $u$  and  $v$ .

**Output.**  $u$  (or  $v$ ).

The 2-adic valuation  $\nu_2$  counts the number of zeroes on the right

### An instance of a LSB Algorithm.

On a pair  $(u, v)$  with  $v$  odd and  $u$  even,

with  $\nu_2(u) = k$ , of the form  $u := 2^k u'$

the LSB division produces

– a quotient  $a$  odd, and  $a \in [-(2^k - 1), 2^k - 1]$ ,

– and a remainder  $r$  with  $\nu_2(r) > k$ , of the form  $r := 2^k r'$ ,

and writes  $v = a \cdot u' + 2^k \cdot r'$ .

The pair  $(r', u')$  satisfies

$$\nu_2(r') > \nu_2(u') = 0 \text{ and } \gcd(u, v) = \gcd(r', u').$$

It will be the new pair for the next step.

An execution of the  
LSB Algorithm on  
(72001, 2011176)

$i$	$u_i$ [base 2]	$a_i$	$k_i$
1	111101011000000101000	-3	3
2	11001001101101010000	1	1
3	110000110001010000000	1	3
4	10011000111100000000	-1	1
5	111010010101000000000	-1	1
6	110000010010000000000	1	1
7	100010001100000000000	-1	1
8	100000101100000000000	1	1
9	1100000000000000000	1	2
10	100000100000000000000	-1	1
11	100010000000000000000	1	1
12	110000000000000000000	-5	3
13	100000000000000000000	3	2

Comparison for five algorithms on the input (2011176, 72001)

Evolution of the remainders

Standard	Centered	By-Excess	Binary	LSB
67149	4852	4852	44849	51637
4852	779	779	1697	12485
4073	178	601	1697	2447
779	67	423	125	3733
178	23	245	125	1545
67	2	67	9	547
44	—	23	9	523
23	—	2	5	3
19	—	—	—	65
4	—	—	—	17
3	—	—	—	3

I – Four types, Six instances of Euclidean algorithms

**II – The average-case analysis: The results.**

III – The dynamical systems underlying the algorithms.

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## A general framework.

Each division–step of each algorithm uses a “digit”  $d = (m, \epsilon, a, b)$ , changes the **old pair**  $(u, v)$  into the **new pair**  $(r', u')$  as

$$u = 2^a \cdot u', \quad v = m \cdot u' + \epsilon \cdot 2^b \cdot r'.$$

On integer pairs, it uses the **matrix** transformation  $M_{[d]}$

$$\begin{pmatrix} u \\ v \end{pmatrix} = M_{[d]} \begin{pmatrix} r' \\ u' \end{pmatrix}, \quad \text{with} \quad M_{[d]} := \begin{pmatrix} 0 & 2^a \\ \epsilon 2^b & m \end{pmatrix}$$

and, on rationals (the **old**  $x = u/v$  and the **new**  $y = r'/u'$ ), it uses the **LFT**  $h_{[d]}$ ,

$$x = h_{[d]}(y) \quad \text{with} \quad h_{[d]}(y) = \frac{2^a}{m + \epsilon 2^b y}.$$

Then  $|\det h_{[d]}| = 2^{a+b}$  involves the total number  $a + b$  of **binary shifts**.

## A generic execution.

On the input pair  $(u, v) = (u_1, u_0)$ , it is of the form

$$\left\{ \begin{array}{l} u_1 := 2^{-a_1} u_1, \quad u_0 = m_1 u_1 + \epsilon_1 2^{b_1} u_2, \\ u_2 := 2^{-a_2} u_2, \quad u_1 = m_2 u_2 + \epsilon_2 2^{b_2} u_3, \\ \dots, \quad \dots \\ u_i := 2^{-a_i} u_i, \quad u_{i-1} = m_i u_i + \epsilon_i 2^{b_i} u_{i+1} \\ \dots, \quad \dots \end{array} \right\},$$

and uses the sequence of digits  $d_i := (m_i, \epsilon_i, a_i, b_i)$ .

It stops at the  $p$ -th iteration with  $u_{p+1} = \eta \cdot u_p$  [ $\eta = 0$  or  $\eta = 1$ ].

Then  $\gcd(u, v) = u_p$ .

## Cost of an execution.

Given a positive **step-cost**  $c$  defined on the set  $\mathcal{D}$  of digits, consider the **total cost**  $C$  defined on the input  $(u, v)$  in an **additive** way as

$$C(u, v) := \sum_{i=1}^p c(d_i), \quad d_i := (m_i, \epsilon_i, a_i, b_i)$$

The step-cost  $c$  is of **moderate growth**, when  $c(d) = O(\log m)$

**Main costs of moderate growth.**

- if  $c \equiv 1$ , then  $C = P$  is the **number of iterations**
- if  $c$  is the characteristic function  $\mathbf{1}_{d_0}$  of a given digit  $d_0$ , then  $C$  is the **number of occurrences of  $d_0$**  in the CF.
- if  $c(d) = a + b$ , then  $C$  is the **total number of binary shifts**.
- if  $c(d) = \ell(m)$ , the binary length of digit  $m$ , then  $C$  is the **encoding length** of the CF.



## An important (non additive) cost.

The bit-complexity

$$B(u, v) := \sum_{i=1}^p \ell(m_i) \cdot \ell(u_i)$$

involves **both digits**  $d_i$  (via a cost of moderate length) and size of **remainders**  $u_i$ ... The most **precise** cost

## An Important Question.

**Compare** the behaviour of these various Euclidean algorithms with respect to different costs, and particularly **the bit-complexity**.

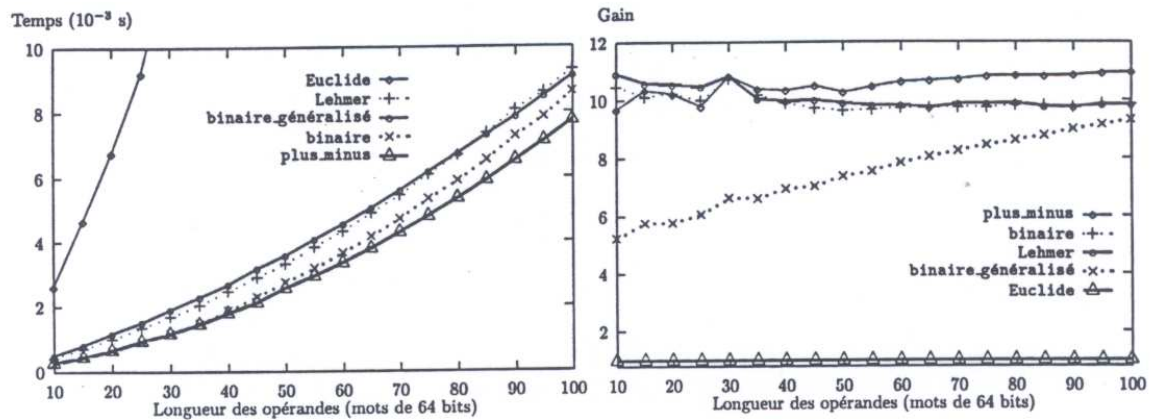


FIG. 9.20 - Temps et gain par rapport à Euclide (station DEC).

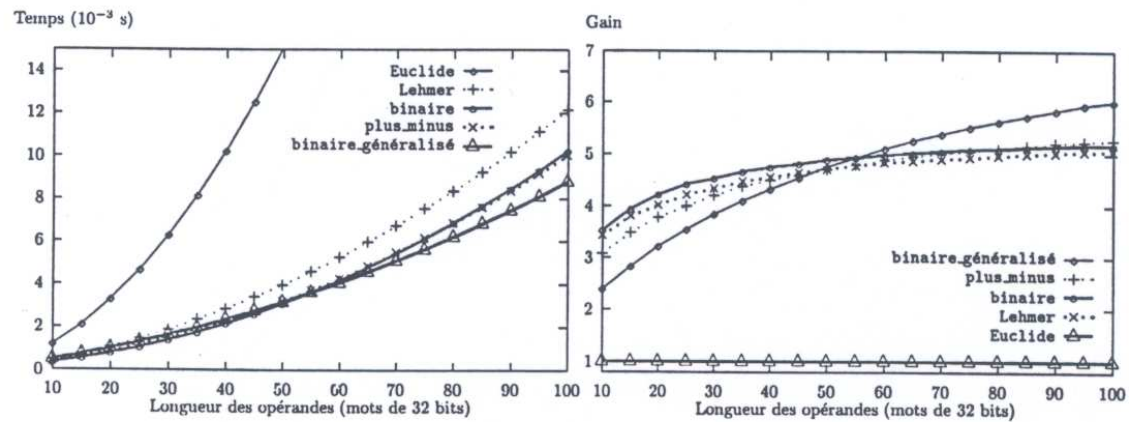


FIG. 9.21 - Temps et gain par rapport à Euclide (station SUN).

## The analysis of the Euclidean Algorithms.

The length of an input  $(u, v)$  is  $|(u, v)| \sim (u^2 + v^2)^{1/2}$ .

Its size is  $L(u, v) \sim (1/2) \lg(u^2 + v^2)$

If the set of all possible inputs  $(u, v)$  of the algorithm is  $\Omega$ ,  
the algorithm is studied on  $\Omega_n := \{(u, v) \in \Omega; L(u, v) = n\}$   
for  $n \rightarrow \infty$ .

**Previous results**, mostly in the **average**-case,  
**only** for parameter  $P$ , and **specific** to **particular** algorithms...  
well-described in Knuth's book (Tome II)

Heilbronn, Dixon, Rieger (70): **Standard** and **Centered** Alg.

Yao and Knuth (75): **Subtractive** Alg.

Brent (78): **Binary** Alg (**partly heuristic**),

Hensley (94) : A **distributional** study for the **Standard** Alg.

Stehlé and Zimmermann (05) : **LSB** Alg (**experiments**)

Then **Dynamical Analysis** [Caen group, 1995  $\rightarrow?$ ] provides

- a **complete classification** into two classes,
  - the **Fast Class** = {Standard, Centered, Binary, LSB},
  - the **Slow Class** = {By-Excess, Subtractive}.
- an **average-case** analysis of a broad **class of costs**,
  - all the **additive costs**,
  - and also **the bit-complexity**.
- a **distributional** analysis of a subclass of the Fast Class,
  - the **Good Class** = {Standard, Centered}.

**Asymptotic gaussian laws** hold for:

- $P$ , and **additive** costs of **moderate** growth,
- the size of the remainders  $\log u_i$  when  $i \sim \delta P$ ,
- the **bit-complexity** of the extended Alg.

**Here, focus on average-case results.**

- For the **Fast Class** = {Standard, Centered, Binary, LSB} ,
- the mean values of costs  $C, B$  are **linear** wrt  $n$ ,
- the mean bit-complexity is **quadratic**.

$$\mathbb{E}_n[P] \sim \frac{2 \log 2}{h(\mathcal{S})} n, \quad \mathbb{E}_n[C] \sim \frac{2 \log 2}{h(\mathcal{S})} \mu[c] n, \quad \mathbb{E}_n[B] \sim \frac{\log 2}{h(\mathcal{S})} \mu[\ell] n^2.$$

$h(\mathcal{S})$  is the entropy of the system,  $\mu[c]$  the mean value of step-cost  $c$ .

- Moreover, these costs are **concentrated**:  $\mathbb{E}_n[C^k] \sim E_n[C]^k$

- For the **Slow Class** = {By-Excess, Subtractive},
- the mean values of costs  $P, C$  are **quadratic**,
- the mean bit-complexity of  $B$  is **cubic**,
- the moments of order  $k \geq 2$  are **exponential**:  $\mathbb{E}_n[C^k] = \Theta(2^{n(k-1)})$ .

The main constant  $h(\mathcal{S})$  is the **entropy** of the Dynamical System.

A **well-defined** mathematical object, **computable**.

– Related to classical constants for the **first two algs**

$$h(\mathcal{S}) = \frac{\pi^2}{6 \log 2} \sim 2.37 \text{ [Standard]}, \quad h(\mathcal{S}) = \frac{\pi^2}{6 \log \phi} \sim 3.41 \text{ [Centered]}.$$

– For the **LSB alg**,  $h(\mathcal{S}) = 4 - 2\gamma \sim 3.91$  involves the **Lyapounov exponent**  $\gamma$  of the set of random matrices, where

$$N_{a,k} = \frac{1}{2^k} \begin{pmatrix} 0 & 2^k \\ 2^k & a \end{pmatrix} \text{ with } k \geq 1, a \text{ odd, } |a| < 2^k \text{ is taken with prob. } 2^{-2k},$$

– For the **Binary alg**,  $h(\mathcal{S}) = \pi^2 f(1) \sim 3.6$  involves the value  $f(1)$  of the unique density which satisfies the functional equation

$$f(x) = \sum_{k \geq 1} \sum_{\substack{a \text{ odd} \\ 1 \leq a < 2^k}} \left( \frac{1}{2^k x + a} \right)^2 f \left( \frac{1}{2^k x + a} \right)$$

## Precise comparisons between the four Fast Algorithms

Algs	Nb of iterations	Bit-complexity
Standard	$0.584 n$	$1.242 n^2$
Centered	$0.406 n$	$1.126 n^2$
(Ind.) Binary	$0.381 n$	$0.720 n^2$
LSB	$0.511 n$	$1.115 n^2$

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## The Euclidean dynamical System (I).

The trace of the execution of the Euclid Algorithm on  $(u_1, u_0)$  is:

$$(u_1, u_0) \rightarrow (u_2, u_1) \rightarrow (u_3, u_2) \rightarrow \dots \rightarrow (u_{p-1}, u_p) \rightarrow (u_{p+1}, u_p) = (0, u_p)$$

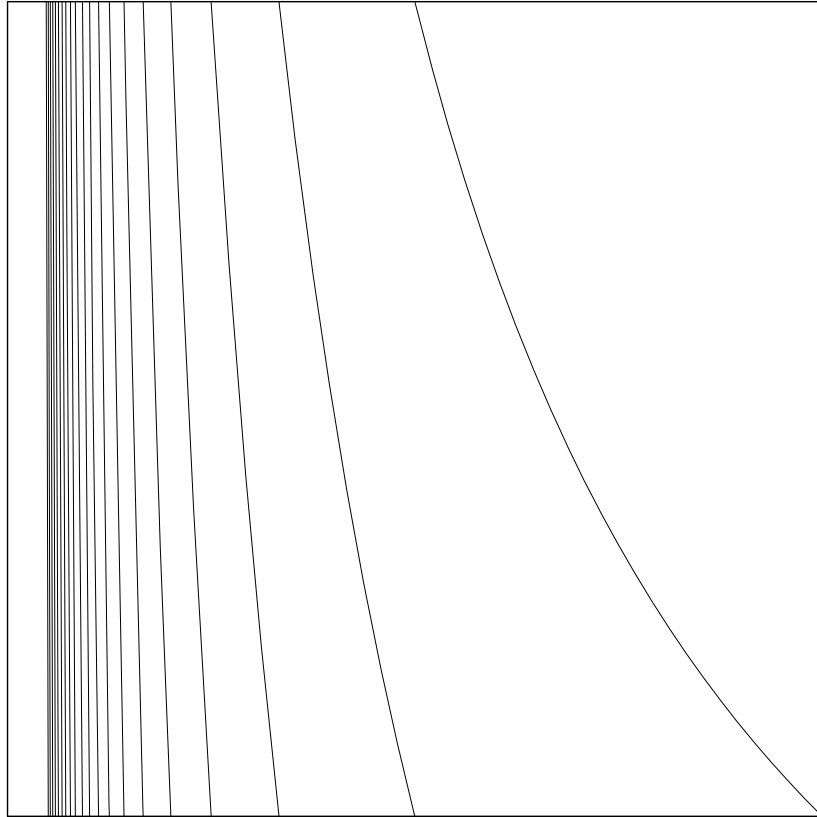
Replace the integer pair  $(u_i, u_{i-1})$  by the rational  $x_i := \frac{u_i}{u_{i-1}}$ .

The division  $u_{i-1} = m_i u_i + u_{i+1}$  is then written as

$$x_{i+1} = \frac{1}{x_i} - \left[ \frac{1}{x_i} \right] \quad \text{or} \quad x_{i+1} = T(x_i), \quad \text{where}$$

$$T : [0, 1] \longrightarrow [0, 1], \quad T(x) := \frac{1}{x} - \left[ \frac{1}{x} \right] \quad \text{for } x \neq 0, \quad T(0) = 0$$

An execution of the Euclidean Algorithm  
= A rational trajectory of the Dynamical System  $([0, 1], T)$   
= a trajectory that reaches 0.



## The Euclidean dynamical System (II).

A dynamical system with a denumerable system of branches  $(T_{[m]})_{m \geq 1}$ ,

$$T_{[m]} := ]\frac{1}{m+1}, \frac{1}{m}[ \longrightarrow ]0, 1[, \quad T_{[m]}(x) := \frac{1}{x} - m$$

The set  $\mathcal{H}$  of the inverse branches of  $T$  is

$$\mathcal{H} := \left\{ h_{[m]} := ]0, 1[ \longrightarrow ]\frac{1}{m+1}, \frac{1}{m}[; \quad h_{[m]}(x) := \frac{1}{m+x} \right\}$$

The set  $\mathcal{H}$  builds **one step** of the CF's.

The set  $\mathcal{H}^n$  is the set of the **inverse branches of  $T^n$** ;

it builds CF's of **depth  $n$** .

The set  $\mathcal{H}^* := \bigcup \mathcal{H}^n$  builds **all the** (finite) CF's.

## Dynamical Systems relative to MSB Algorithms.

A **continuous** dynamical system extends each **discrete** division:

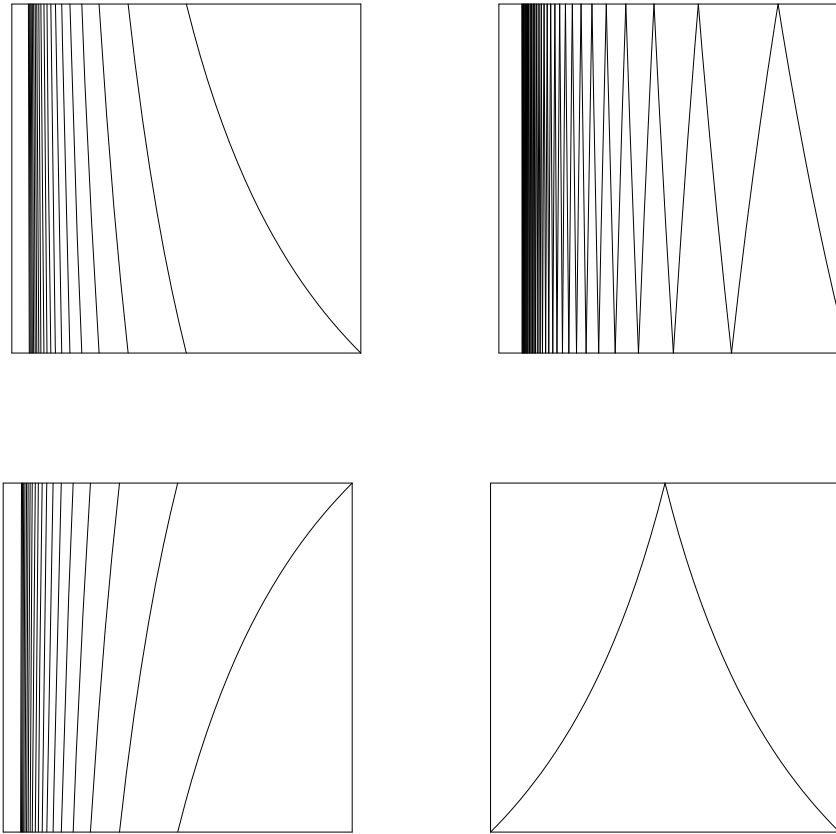
Replace the rational  $u/v$  by a generic **real**  $x$  [for the MSB Group]

Key Property : Expansiveness of branches

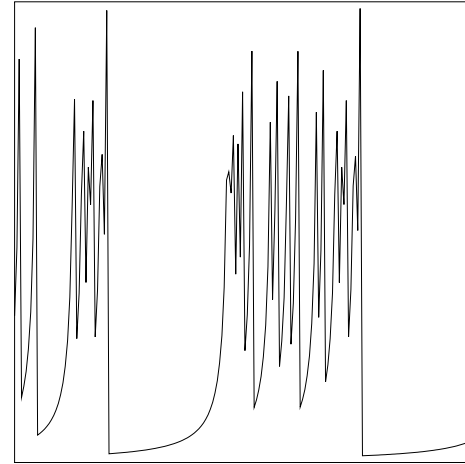
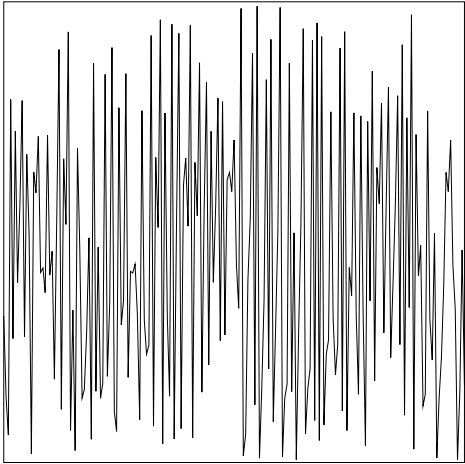
$$|T'(x)| \geq \rho > 1 \text{ for all } x \text{ in } \mathcal{I}$$

When **true**, this implies a **chaotic** behaviour for trajectories. The associated algorithms are **Fast** and belong to the **Good Class**

When this condition is **violated at only one fixed point**, this leads to **intermittency phenomena**. The associated algorithms are **Slow**.



Dynamical systems for MSB Algorithms; above, Standard and Centered; on the bottom, By-Excess and Subtractive. There exists an indifferent point for systems on the bottom line :  $\eta = 1$  for  $(M)$  –  $\eta = 0$  for  $(T)$ .



**Chaotic** Orbit [Fast Class], **Intermittent** Orbit [SlowClass].

### Induction Method.

For a DS with a “**slow**” branch relative to a **slow** interval  $J$ ,  
**contract** each part of the trajectory which **belongs** to  $J$  into **one** step.  
 This (often) permits to **transform** the slow DS into a fast one

While  $x \in J$  do  $x := T(x)$ ;  
 $S(x) := T(x)$ ;

The **Induced DS of the Subtractive** Alg = the DS of the **Standard** Alg.

## The DS relative to the Binary Algorithm:

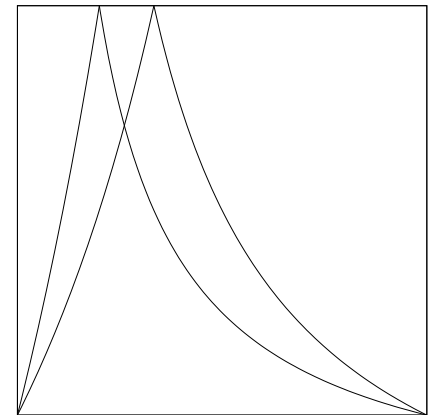
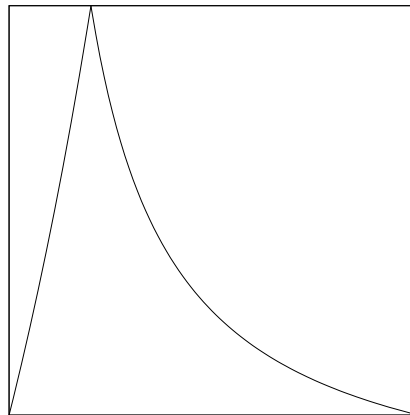
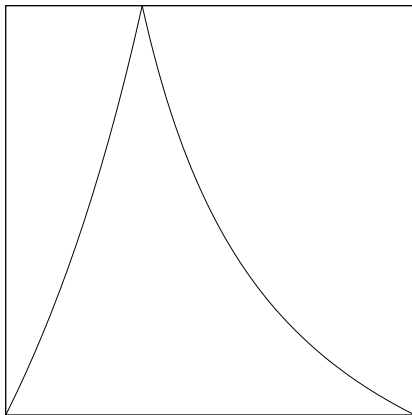
An instance of a probabilistic dynamic system.

The 2-adic valuation  $\nu$  (only defined on rationals) is extended to a real random variable  $\nu$  with

$$\mathbb{P}[\nu = k] = 1/2^k \quad \text{for } k \geq 1.$$

## The Binary Dynamical system

for  $k = 1$ ,  $k = 2$ ; then with  $k = 1$  and  $k = 2$  on the same figure.



## The induced DS of the Binary Algorithm:

A dynamical system where the set of inverse branches is

$$\mathcal{H} := \{h_{a,k} : x \mapsto \frac{1}{2^k x + a}, \quad k \geq 1, a \text{ odd}, a < 2^k\}.$$

For a real  $x \in ]2^{-k_0}, 2^{-(k_0-1)}]$ , the set of possible branches is

$$\{h_{a,k}; \quad k \geq k_0, a \text{ odd}, a < 2^{k_0}\},$$

where each  $h_{a,k}$  is chosen with probability  $2^{-k}$ .

The **probabilistic** choice **depends** on the **position** of real  $x$ .



## The LSB Dynamical System.

The remainders are **not decreasing**,

so that the rationals  $x = u/v$  may belong to **the whole  $\mathbb{R}$**  .

Using the **conjugaison** with the **tangent** map leads to work inside the **compact interval**  $J = [-\pi/2, +\pi/2]$  and to deal with inverse branches

$$h_{a,k} : x \mapsto \arctan \left( \frac{2^k}{2^k \tan x + a} \right)$$

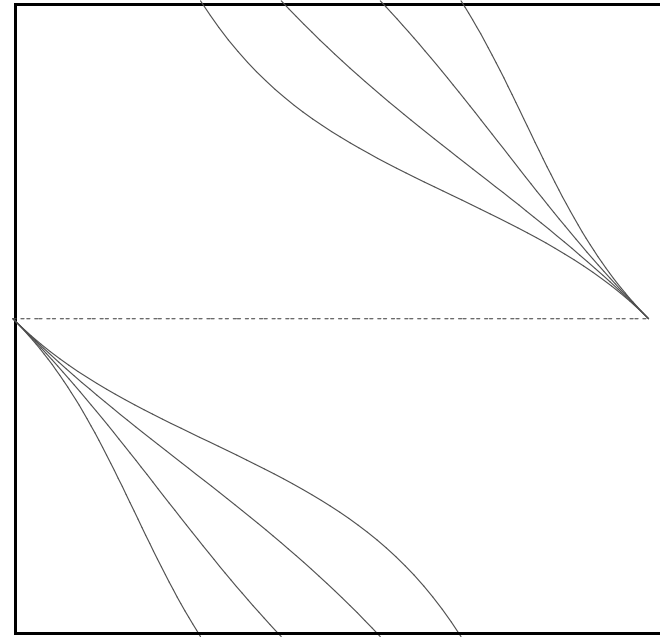
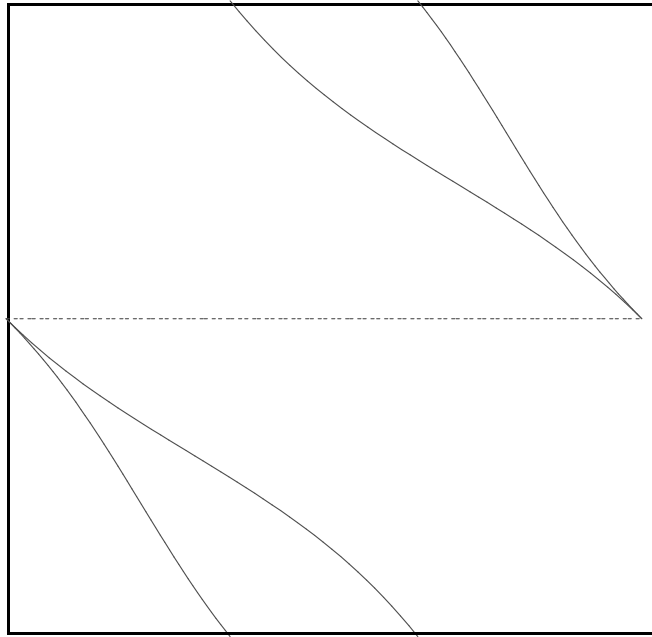
At **each** step of the process, **each** branch

$$h_{a,k}, \text{ with } k \geq 1, \quad a \text{ odd, } |a| < 2^k$$

is chosen with probability  $1/2^{2k}$ .

The **probabilistic** choice does **not** depend on the **position** of  $x$ .

This defines a **system of iterated functions**.



The DS relative to the LSB Alg.

On the left, for  $k = 1[a = \pm 1]$  – On the right, for  $k = 2[a = \pm 1, \pm 3]$ .

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## General principles of Dynamical Analysis.

Two objects:

The (discrete) Algorithm, the (continuous) Dynamical System

Two tools:

The generating function, The transfer operator

And their relations:

Geometric properties of the Dynamical System



Spectral properties for the Transfer Operator  
in a convenient functional space.



Analytical properties of the Dirichlet series



Asymptotic Analysis of the Algorithm

The density transformer  $\mathbf{H}$  expresses the new density  $f_1$  as a function of the old density  $f_0$ , as  $f_1 = \mathbf{H}[f_0]$ . It involves the set  $\mathcal{H}$

$$\mathbf{H}[f](x) := \sum_{h \in \mathcal{H}} \delta_h \cdot |h'(x)| \cdot f \circ h(x) \quad [\text{here, } \delta_h = \mathbb{P}[h]]$$

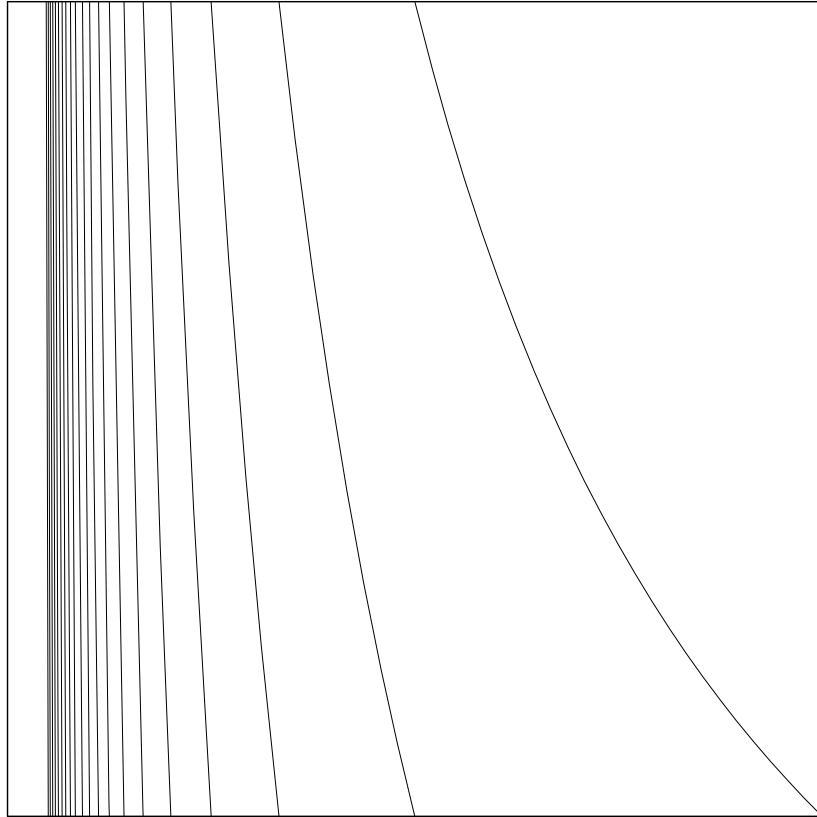
With a cost  $c : \mathcal{H} \rightarrow \mathbf{R}^+$  defined by  $c(h_{[d]}) := c(d)$  and extended to  $\mathcal{H}^*$  by additivity, it gives rise to the weighted transfer operator

$$\mathbf{H}_{s,w}[f](x) := \sum_{h \in \mathcal{H}} \delta_h^s \cdot \exp[wc(h)] \cdot |h'(x)|^s \cdot f \circ h(x)$$

$\left\{ \begin{array}{l} \text{Multiplicative properties of the derivatives and probabilities} \\ \text{Additive properties of the cost} \end{array} \right\} \implies$

$$\mathbf{H}_{s,w}^k[f](x) := \sum_{h \in \mathcal{H}^k} \delta_h^s \cdot \exp[wc(h)] \cdot |h'(x)|^s \cdot f \circ h(x)$$

$\mathbf{H}_{s,w}^k$  generates all the weighted trajectories truncated at length  $k$ .  
 $(I - \mathbf{H}_{s,w})^{-1} = \sum_{n \geq 0} \mathbf{H}_{s,w}^n$  generates all the finite trajectories



## The Dirichlet series.

If  $\Omega$  is the whole set of inputs, the Dirichlet generating function

$$S_C(s) = \sum_{(u,v) \in \Omega} \frac{C(u,v)}{|(u,v)|^{2s}} = \sum_{m \geq 1} \frac{c_m}{m^{2s}} \quad \text{with } c_m := \sum_{\substack{(u,v) \in \Omega \\ |(u,v)|=m}} C(u,v)$$

is used for expressing the mean value  $\mathbb{E}_n[C]$  of  $C$  on  $\Omega_n$ , since

$$\mathbb{E}_n[C] = \frac{1}{|\Omega_n|} \sum_{m=2^{n-1}}^{2^n-1} c_m$$

The relation  $S_C(s) = \frac{\partial}{\partial w} S_C(s, w)$ ,

$$\text{with } S_C(s, w) := \sum_{(u,v) \in \Omega} \frac{1}{|(u,v)|^{2s}} \exp[wC(u,v)],$$

proves that it is sufficient to deal with  $S_C(s, w)$  and obtain an alternative expression for it.

## Relation between the transfer operator and the Dirichlet series.

We need generating both the cost  $C(u, v)$  and the length  $|(u, v)|$ . Any Euclid Algorithm provides a unique decomposition

$$u/v = h(\eta), \quad \text{with} \quad \eta = 0 \quad \text{or} \quad 1 \quad \text{and} \quad h \in \mathcal{H}^*.$$

Then,  $C(u, v) = c(h)$ . There are two main choices for  $|(u, v)|$ ,

$$|(u, v)| := \max(u, v) = v \quad \text{or} \quad |(u, v)|^2 = u^2 + v^2$$

according as the remainders are decreasing or not. In all the cases, (with the use of the tangent-conjugaison in the LSB case),

$$\frac{1}{|(u, v)|^{2s}} = \delta_h^s \cdot |h'(\eta)|^s,$$

Then, there is a nice relation between  $S_C(s, w)$  and  $\mathbf{H}_{s, w}$ ,

$$S_C(s, w) = (I - \mathbf{H}_{s, w})^{-1}[1](\eta),$$

and finally 
$$S_C(s) = (I - \mathbf{H}_s)^{-1} \circ \mathbf{H}_s^{[c]} \circ (I - \mathbf{H}_s)^{-1}[1](\eta)$$



Finally, an alternative expression of the Dirichlet series

$$S_C(s) := \sum_{(u,v) \in \Omega} \frac{C(u,v)}{|(u,v)|^{2s}} = (I - \mathbf{H}_s)^{-1} \circ \mathbf{H}_s^{[c]} \circ (I - \mathbf{H}_s)^{-1}[1](\eta)$$

as a function of two operators

$$\mathbf{H}_s^{[c]}[f](x) = \sum_{h \in \mathcal{H}} \delta_h^s \cdot c(h) \cdot |h'(x)|^s \cdot f \circ h(x)$$

and the quasi-inverse  $(I - \mathbf{H}_s)^{-1}$  of the transfer operator  $\mathbf{H}_s$ ,

$$\mathbf{H}_s[f](x) := \sum_{h \in \mathcal{H}} \delta_h^s \cdot |h'(x)|^s \cdot f \circ h(x).$$

The quasi-inverse plays the leading rôle (except for Slow Alg): For **extracting coefficients** via Tauberian Theorems, we need to know the **position** and the **nature** of the (dominant) **singularity** of the quasi-inverse  $(I - \mathbf{H}_s)^{-1}$ .

Analytical properties of  $(I - \mathbf{H}_s)^{-1}$  for applying Tauberian Theorems.

On the line  $\Re s = 1$ ,  
 $s \neq 1$ ,  $\mathbf{H}_s$  is analytic.



$s=1$

$s = 1$  is a (simple) pole

$$(I - \mathbf{H}_s)^{-1} \sim \frac{a}{s - 1}$$

Half-plane of convergence  $\Re s > 1$

No hypothesis needed  
on the half-plane  $\Re s < 1$ .

Since  $\mathbf{H}_1$  is the density transformer,  $s = 1$  is the dominant singularity.

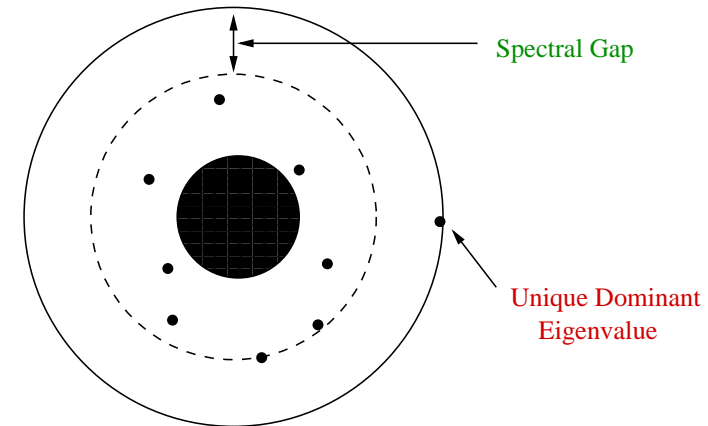
## Expected spectral properties of $\mathbf{H}_s$

(i) *UDE* and *SG* for  $s$  near 1:

*UDE* – Unique dominant eigenvalue

*SG* – Existence of a spectral gap

(ii) *Aperiodicity*: for  $\Re s = 1, s \neq 1$ ,  
the spectral radius of  $\mathbf{H}_s$  is  $< 1$



On **which** functional space?

The answer **depends** on the DS,  
and thus **on the algorithm**....

The functional space where the triple  $UDE + SG + \text{Aperiodicity}$  holds.

Algs	Geometry of branches	Convenient Functional space
Good Class (Standard, Centered)	Contracting	$\mathcal{C}^1(\mathcal{I})$
Binary	Not contracting	The Hardy space $\mathcal{H}(\mathcal{D})$
LSB	Contracting on average	Various spaces: $\mathcal{C}^0(J), \mathcal{C}^1(J)$ Hölder $\mathbb{H}_\alpha(J)$
Slow Class (Subtractive, By-Excess)	An indifferent point	Induction + $\mathcal{C}^1(\mathcal{I})$

The aperiodicity holds since the branches have not all the same form.

## Conclusion.

Here, the average-case of a class of Euclid Algorithms....  
which uses the transfer operator  $\mathbf{H}_s$  of the underlying DS, with  $\Re s \geq 1$ .

## Possible extensions.

Talks of Loick LHOTE and Antonio VERA

- Distributional analysis of the Euclid algorithms  
or Analysis of **Fast variants** of the Euclid Algorithms [**Loick**]  
Use the **same** transfer operator  $\mathbf{H}_s$ , with its behaviour for  $\Re s < 1$
- Study of the **Gauss** algorithm (for **reducing** lattices) [**Antonio**]  
Use of an **extension** of the transfer operator  $\mathbf{H}_s$ ,  
which operates on functions of two variables, for  $s \sim 2$