

The EUCLID ALGORITHM is "TOTALLY" GAUSSIAN

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Study of Local Limit Theorems, with their speed of convergence.

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even in the simplest probabilistic framework.

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II- Distributional results which are already known

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Local limit theorems in the particular case of a lattice cost.

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$$u_0 := v; u_1 := u; u_0 \geq u_1$$

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

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$$\text{CFE of } \frac{u}{v}: \quad \frac{u}{v} = \frac{1}{m_1 + \frac{1}{m_2 + \frac{1}{\ddots + \frac{1}{m_p}}}},$$

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– the $\gcd(u, v)$ itself

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 - Extensively used in cryptography
- the Continued Fraction Expansion CFE (u/v)
 - Often used directly in computation over rationals.
 - The main object of interest here.

A basic algorithm ... Perhaps the fifth main operation?

The main costs of interest for the continued fraction expansion

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if $d = 1$, then $\widehat{D} :=$ the number of iterations

if $d = \mathbf{1}_{m_0}$, then $\widehat{D} :=$ the number of digits equal to m_0

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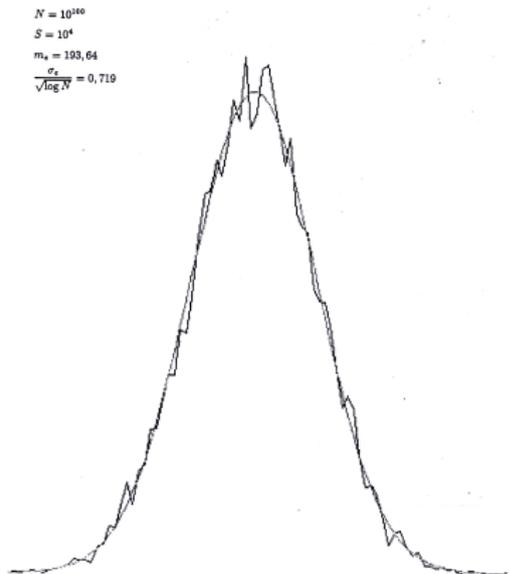
However, it is also interesting to study general digit costs,

They give rise to various observables on the Continued Fraction expansion

For instance $d(m) = \log m$, related to the Khinchine constant.

Main probabilistic questions on the Continued Fraction Expansion ... and its "total" cost \widehat{D}

$N = 10^{100}$
 $S = 10^4$
 $m_s = 193,64$
 $\frac{\sigma_s}{\sqrt{\log N}} = 0,719$

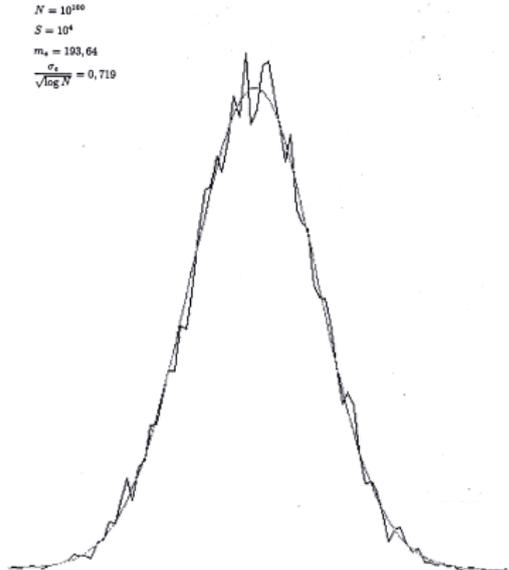


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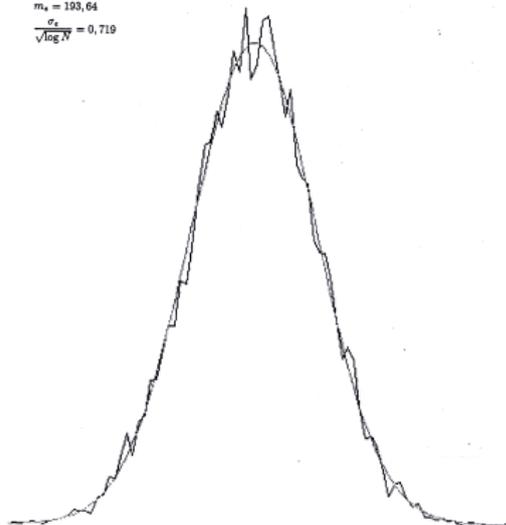
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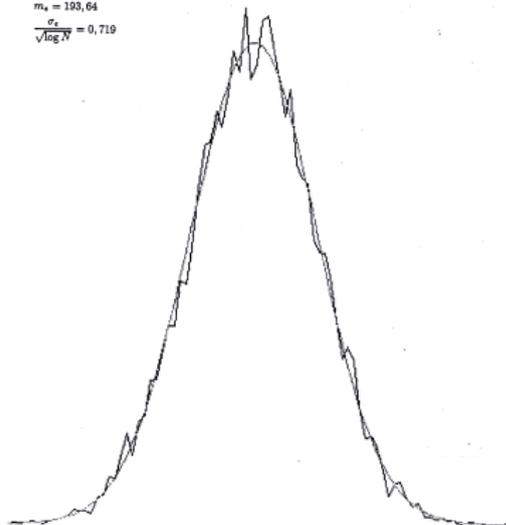
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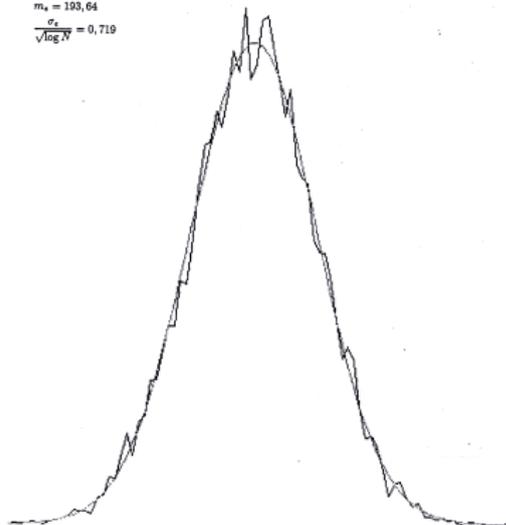
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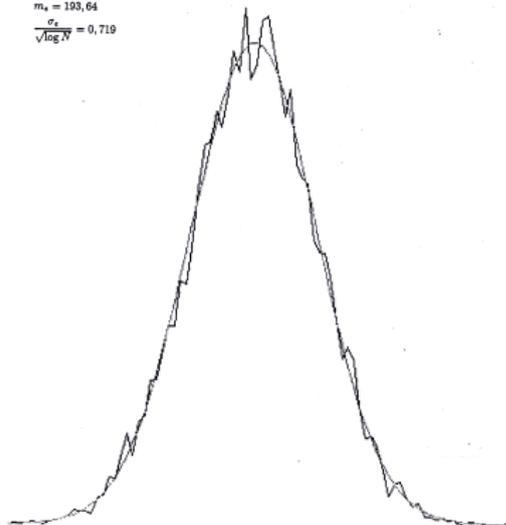
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Which speed of convergence?

The underlying dynamical system (I).

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

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

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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

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

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

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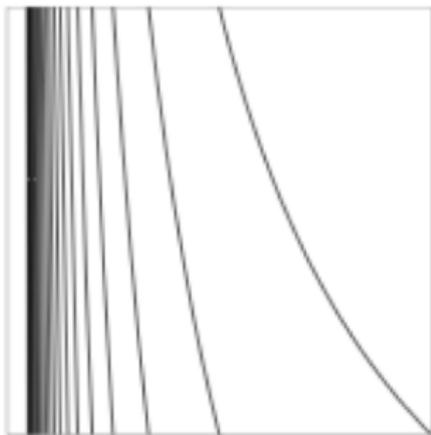
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An **execution** of the Euclidean Algorithm $(x, T(x), T^2(x), \dots, 0)$

= A **rational trajectory** of the Dynamical System $([0, 1], T)$

= a **trajectory** that reaches **0**.

The dynamical system is a continuous extension of the algorithm.



$$T(x) := \frac{1}{x} - \left\lfloor \frac{1}{x} \right\rfloor$$

$$T_{[m]} := \frac{1}{m+1}, \frac{1}{m} [\rightarrow) 0, 1[,$$

$$T_{[m]}(x) := \frac{1}{x} - m$$

$$h_{[m]} :=] 0, 1[\rightarrow \frac{1}{m+1}, \frac{1}{m} [$$

$$h_{[m]}(x) := \frac{1}{m+x}$$

$$\frac{u}{v} = \frac{1}{m_1 + \frac{1}{m_2 + \frac{1}{\ddots + \frac{1}{m_p}}}} = h_{[m_1]} \circ h_{[m_2]} \circ \dots \circ h_{[m_p]}(0)$$

The **discrete** algorithm is extended into a **continuous** process.

Two types of weighted trajectories and **two probabilistic** models:

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First model : Study of **truncated real** trajectories “at depth n ”

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We wish to **compare** these **two “observables”** .

Since the **discrete** data are of **zero measure** amongst the **continuous** data,
we need a “transfer from **continuous** to **discrete**”.

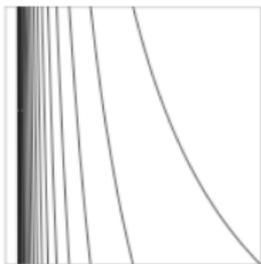
A main tool in **both probabilistic** models: The **transfer** operator.

The transfer operator

Density Transformer:

For a density f on $[0, 1]$, $\mathbf{H}[f]$ is the density on $[0, 1]$ after one iteration of the shift

$$\mathbf{H}[f](x) = \sum_{h \in \mathcal{H}} |h'(x)| f \circ h(x) = \sum_{m \in \mathbb{N}} \frac{1}{(m+x)^2} f\left(\frac{1}{m+x}\right).$$



$\mathcal{H} :=$ the set of
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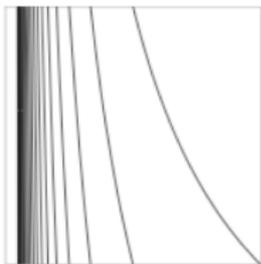
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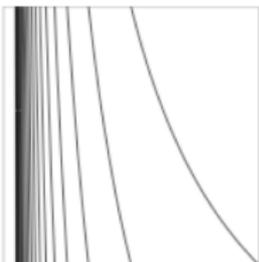
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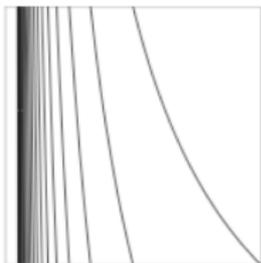
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The k -th iterate satisfies, with d extended in an additive way

$$\mathbf{H}_{s,w}^k[f](x) = \sum_{h \in \mathcal{H}^k} |h'(x)|^s e^{wd(h)} f \circ h(x)$$



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II- Distributional results for weighted trajectories

Transfer operator and distributional study of weighted trajectories

In distributional studies, the main tools are the **characteristic functions**

$$\mathbb{E}[\exp(wD_n)], \quad \mathbb{E}_N[\exp(w\hat{D})]$$

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Rational case :
$$\mathbb{E}_N[\exp(w\hat{D})] \text{ related to } [N^{-s}](I - \mathbf{H}_{s,w})^{-1}[1](0)$$

due to the relation between

Dirichlet generating functions and quasi-inverses of the transfer operator,

$$S_d(s, w) := \sum_{(u,v) \in \Omega} \frac{1}{v^{2s}} \exp[w\hat{D}(u, v)] = (I - \mathbf{H}_{s,w})^{-1}[1](0)$$

Distributional results for the continued fraction expansion

Already known results [Baladi-V (2003)]

In both cases, **Real** trajectories or **Rational** trajectories,

For a cost d of moderate growth $d(m) = O(\log m)$,

(a) **Central** Limit Theorems hold for D_n, \hat{D}_N

(b) Moreover, for a **lattice cost**, **Local** Limit Theorems hold for D_n, \hat{D}_N

$$\exists d_0, L \in \mathbb{R}, \quad \text{with } L > 0, \text{ such that } \forall m \quad \frac{d(m) - d_0}{L} \in \mathbb{Z}$$

(c) With **optimal speed** of convergence

$$O\left(\frac{1}{\sqrt{n}}\right), \quad O\left(\frac{1}{\sqrt{\log N}}\right)$$

Distributional results for the continued fraction expansion

They deal with the characteristic functions $\mathbb{E}[\exp(wD_n)]$, $\mathbb{E}_N[\exp(w\hat{D})]$
and thus with the transfer operator $\mathbf{H}_{s,w}$

Different cases of study for parameters s and w

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- Real trajectories: $s = 1$
- Rational trajectories $s = 1 + it$, with $t \in \mathbb{R}$

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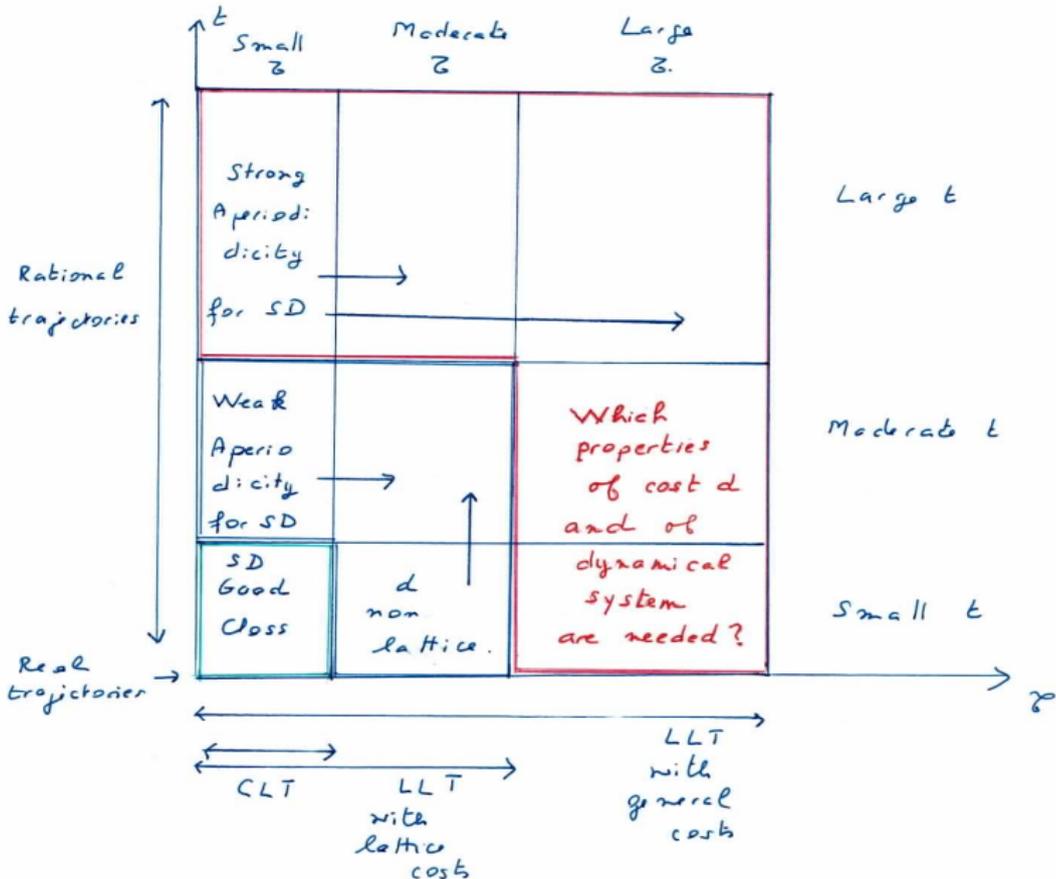
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For parameter w :

- Central Limit Theorems:
 $w \sim 0$
- Local Limit Theorems for a lattice cost :
 $w = i\tau$ with $\tau \in K$ compact $\subset \mathbb{R}$
- Local Limit Theorems for a non lattice cost :
 $w = i\tau$ with $\tau \in \mathbb{R}$

Properties of the dynamical system and cost needed in distributional studies

for dealing with the operator $\mathbf{H}_{1+it, i\tau}$ in each each domain (t, τ) .



III- Local limit theorems with speed of convergence in simpler cases
Memoryless case.

Let (X_i) be a **i.i.d sequence** with values in \mathbb{N} , and $p_m := \Pr[X_i = m]$.
A cost $d : \mathbb{N} \rightarrow \mathbb{R}^+$, Some technical conditions:

$$\sigma_0 := \inf\{\sigma; \sum_{i=1}^{\infty} p_m^\sigma < \infty\} < 1, \quad d(m) = O(|\log p_m|)$$

The mean $\mu[d]$ and the standard deviation $\sigma[d]$ exist. We assume **$\sigma[d] \neq 0$** .

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There is a **Central Limit Theorem (CLT)** for D_n

with a **speed of convergence** of order $O(1/\sqrt{n})$,

$$\Pr\left[\frac{D_n - n\mu[d]}{\sigma[d]\sqrt{n}} \leq y\right] - \frac{1}{\sqrt{2\pi}} \int_{-\infty}^y e^{-t^2/2} dt = O\left(\frac{1}{\sqrt{n}}\right).$$

A Local Limit Theorem (LLT)

- deals with $Q(x, n) := \mu[d]n + \delta[d]x\sqrt{n}$,
- evaluates the probability that $D_n - Q(x, n)$ belongs to some $J \subset \mathbb{R}$,
- compares it to $(|J|/\sqrt{2\pi n}) e^{-x^2/2}$.

A Local Limit Theorem (LLT) proves that

$$\sqrt{n} \Pr[D_n - Q(x, n) \in J] - |J| \frac{e^{-x^2/2}}{\delta(d)\sqrt{2\pi}} \rightarrow 0 \quad (n \rightarrow \infty).$$

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A Local Limit Theorem (LLT) proves that

$$\sqrt{n} \Pr[D_n - Q(x, n) \in J] - |J| \frac{e^{-x^2/2}}{\delta(d)\sqrt{2\pi}} \rightarrow 0 \quad (n \rightarrow \infty).$$

What about the **speed of convergence**?

It depends on **arithmetical** properties of cost d . Two main cases:
the **lattice** case, and the **non-lattice** case.

A cost d is lattice if

$$\exists d_0, L \in \mathbb{R}, \quad \text{with } L > 0, \text{ such that } \forall m \quad \frac{d(m) - d_0}{L} \in \mathbb{Z}$$

The smallest possible $L > 0$ is called the span of the lattice cost.

If $d_0 = 0$, the cost is called “plain lattice”.

In the **lattice case**, the **optimal speed**, of order $O(1/\sqrt{n})$ is attained.

More precisely, for a plain lattice cost of span 1, one has

$$\sqrt{n} \Pr[D_n = P(x, n)] = \sqrt{2\pi} \frac{e^{-x^2/2}}{\delta(d)} + O\left(\frac{1}{\sqrt{n}}\right) \quad P(x, n) := \lfloor Q(x, n) \rfloor.$$

In this case, the **characteristic function** ϕ is **periodic**

$$\phi(\tau) := \int_{\mathbb{R}} \exp[i\tau x] dP_d(x) = \sum_{m \geq 1} p_m \exp[i\tau d(m)],$$

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In the **non-lattice** case, the **speed** in the LLT depends

- on the behaviour of the **characteristic function** ϕ of cost d , when $\tau \rightarrow \infty$
- on **arithmetic** properties of cost d
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Important fact: There is a **relation** between these two properties.

Proposition (classical and easy). The conditions are equivalent :

(i) The cost d is **lattice**

(ii) There exists $\tau_0 \neq 0$ for which ϕ_d satisfies $|\phi_d(\tau_0)| = 1$.

Moreover, Condition (ii) entails Condition (iii)

(iii) For any $h, k, \ell \in \mathbb{N}$, the ratio $\frac{d(h) - d(k)}{d(h) - d(\ell)}$ is **rational**.

Reinforcements of negations of Conditions (ii) or (iii).

A cost d is of **characteristic exponent** χ if

$$\exists K, \tau_0 > 0, \quad |\phi_d(\tau)| \leq 1 - \frac{K}{|\tau|^\chi} \quad \text{for } |\tau| \geq \tau_0.$$

A cost d is of **diophantine exponent** μ if

$$\exists (h, k, \ell) \in \mathbb{N}^3, \quad \text{such that the ratio } \frac{d(h) - d(k)}{d(h) - d(\ell)} \text{ is Diop } (\mu)$$

A number x is **diophantine** of exponent μ if

$$\exists C > 0, \quad \forall (p, q) \in \mathbb{N}^2, \quad \text{one has: } \left| x - \frac{p}{q} \right| > \frac{C}{q^{2+\mu}}$$

First result (Breuillard)

The cost d is of **characteristic** exponent χ

\implies a **Local Limit Theorem** for D_n with speed $n^{1/\chi}$

For any ϵ with $\epsilon < 1/\chi$, for any compact interval $J \subset \mathbb{R}$,

there exists M_J , so that $\forall x \in \mathbb{R}, \forall n \geq 1$, one has:

$$\left| \sqrt{n} \Pr[D_n(u) - \mathbf{Q}(x, n) \in J] - |J| \frac{e^{-x^2/2}}{\delta(d)\sqrt{2\pi}} \right| \leq \frac{M_J}{n^\epsilon}$$

Second result (Breuillard)

The cost d is of **diophantine** exponent μ ,

$\implies d$ of **characteristic** exponent χ for any $\chi > 2(\mu + 1)$.

Conclusion:

The cost d is of **diophantine** exponent μ ,

\implies a **Local Limit Theorem** for D_n with speed $n^{1/2(\mu+1)}$.

IV- Local limit theorems with speed of convergence
Trajectories of dynamical systems.

And now if the X_i are generated by a **dynamical system**?
For instance the digits of the continued fraction expansion
(they are **no longer independent**)

Case of **real** trajectories

Definition: d is of **characteristic** exponent χ (wrt to the DS), if,

$$\|\mathbf{H}_{1,i\tau}^{n(\tau)}\| \leq 1 - \frac{1}{|\tau|^\chi}, \quad \text{for any } \tau \text{ with } |\tau| \geq \tau_0 \quad n(\tau) := \Theta(\log |\tau|).$$

Two properties:

The cost d is of **characteristic** exponent χ wrt to the DS

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The cost d is of **diophantine** exponent μ ,

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K depends on the DS.

A **good** generalization of the **memoryless** case.

Case of **rational** trajectories.

Definition: d is of **uniform characteristic** exponent χ

$$\|\mathbf{H}_{1+it, i\tau}^{n(\tau)}\| \leq 1 - \frac{1}{|\tau|^\chi}, \quad \text{for any } (t, \tau) \text{ with } |t| \leq a \text{ and } |\tau| \geq \tau_0.$$

NOW: (Baladi-Hachemi)

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Baladi and Hachemi proposed an **intertwined diophantine** condition involving the **branches** of the dynamical system **AND** the **cost** d

Our result:

A set of two conditions NOT intertwined

- The diophantine condition (D) on the cost d
- A (new) condition (C) on the branches of the DS
 - a "diophantine" version of the aperiodicity condition on the DS.

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a "**diophantine**" version of the **aperiodicity** condition on the DS.

The **Aperiodicity Condition** says :

“The branches of the system do not have all the same shape”.

If h^* is the fixed point of branch h ,

This implies that the cost $c(h) := \log |h'(h^*)|$ is **strongly non** additive,
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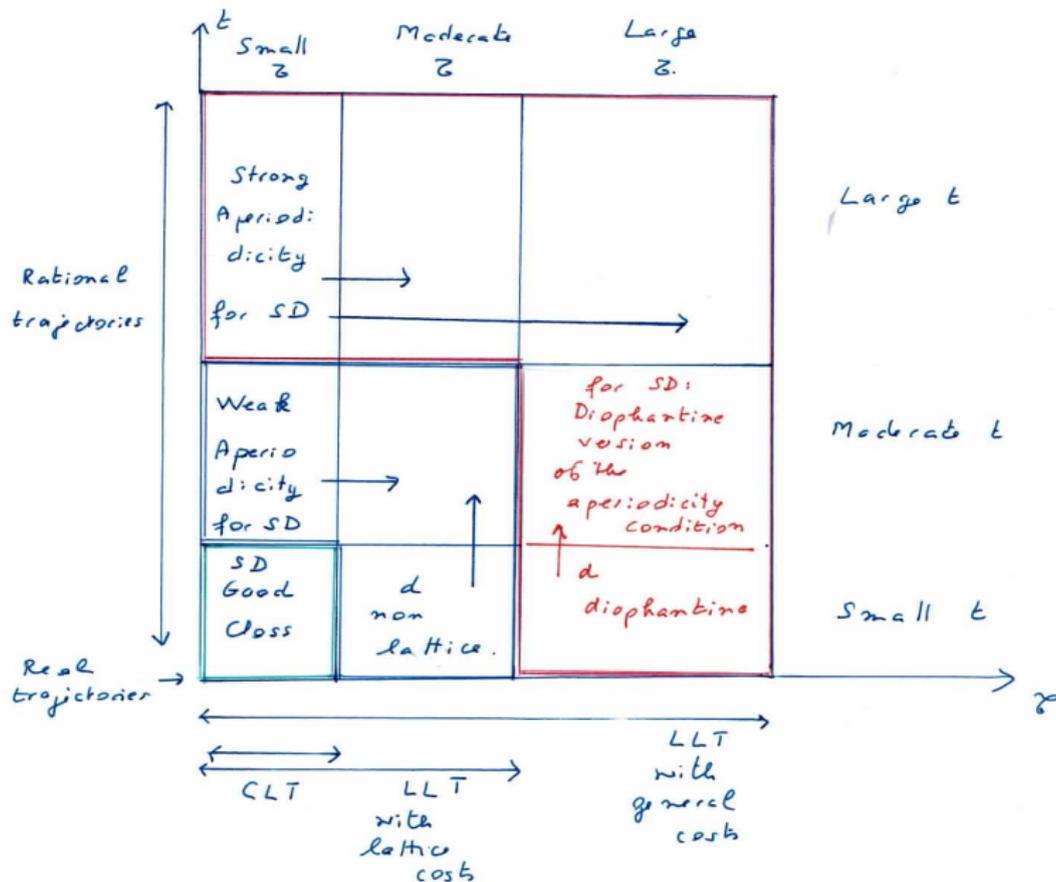
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Our condition (C). There exist three branches h, k, ℓ for which

$$\Gamma(h, k) \neq 0, \quad \Gamma(h, \ell) \neq 0, \quad \text{and} \quad \frac{\Gamma(h, k)}{\Gamma(h, \ell)} \quad \text{is diophantine.}$$

Properties of the dynamical system and cost needed in distributional studies

for dealing with the operator $\mathbf{H}_{1+it, i\tau}$ in each each domain (t, τ) .



Return to the Euclid dynamical system.

In this case, the **condition (C)** is **always satisfied**.

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$$\alpha(h, k) \text{ and } \alpha(h, \ell) \text{ be algebraically independent.}$$
- Baker's theorem proves that the ratio $\Gamma(h, k)/\Gamma(h, \ell)$ is diophantine.

The final result,
for the total costs of a continued fraction relative to some cost d .

$$\widehat{D}_N(x) := \sum_{i=1}^{P(x)} d(m_i(x)) \quad \text{on } \Omega_N := \{x = p/q; \quad q \leq N\}$$
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