

An instance of analysis of algorithms:
The plain GCD algorithm

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Based on a joint work
with Valérie Berthé, Jean Creusefond and Loïck Lhote.

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Computing the gcd of ℓ inputs

For $\ell = 2$: the “classical” Euclid algorithm :

a sequence of Euclidean divisions

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On the input $(x_1, x_2, \dots, x_\ell)$, it computes

– first: $y_2 := \gcd(x_1, x_2)$

– then, for $k \in [3..\ell]$: $y_k := \gcd(x_k, y_{k-1}) = \gcd(x_1, x_2, \dots, x_k)$.

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but **not yet analyzed** for $\ell > 2$ (Problem HM 48)

Which behavior can be expected?

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The **same phenomena** occur for the **size** of the **partial gcd**.

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III. Analysis of the **integer** case.

Based on **Dynamical** Combinatorics and **Dirichlet** Generating Functions

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IV. An **unified** point of view for the two analyses?

A **dynamical** point of view.

I. Probabilistic analysis of algorithms

Based on Analytic Combinatorics
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 - study the **probabilistic behavior of L** on each Ω_n
 - estimate its **mean**, its **variance**, its **distribution**,
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$$\mathbb{E}_n[L] \underset{n \rightarrow \infty}{\sim} a_n, \quad \mathbb{V}_n[L] \underset{n \rightarrow \infty}{\sim} b_n, \quad \mathbb{P}_n \left[\frac{L - a_n}{\sqrt{b_n}} \in [x, x + dx] \right] \underset{n \rightarrow \infty}{\sim} f(x) dx$$

Examples of limit laws (discrete or continuous)

x -axis: possible values of the cost $L(\omega)$

y -axis: probability density $x \mapsto f(x)$

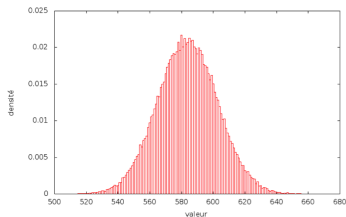
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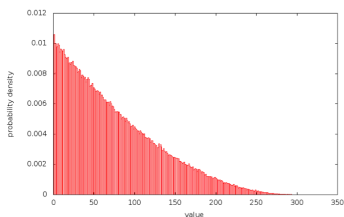
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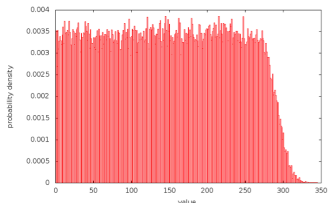
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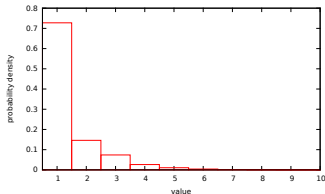
Gaussian law $f(t) \asymp e^{-t^2/2}$



Beta law (a, b) $f(t) \asymp t^{a-1}(1-t)^{b-1}$



Uniform law $f(t) \asymp 1$



A discrete law: Geometric law $f(n) \asymp a^n$

Usual analysis for the Euclid Algorithm $\ell = 2$.

The usual size of $\underline{x} = (x_1, x_2)$ is the **maximum** of the sizes of the inputs:

$$d(\underline{x}) := \max(d(x_1), d(x_2)).$$

The two costs of interest are :

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Its distribution is asymptotically **geometric**.

This (usual) size is **not adapted** for a number of inputs $\ell > 2$.

For $\ell \geq 2$, we consider the **total size** and this choice modifies the results

\implies Irruption of the (unexpected) beta law

Analysis of algorithms and generating functions.

Two main cases here, depending on

- the **type** of the input $\underline{x} = (x_1, x_2, \dots, x_\ell)$, $x_i \in \mathbb{N}$ or $x_i \in \mathbb{F}_q[X]$
- the natural **size** on the input : the **size of** $\prod x_i$.

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– for polynomials: the total degree $d(\underline{x}) = d(\prod x_i) = \sum d(x_i)$

$$\Omega_n := \{\underline{x} = (x_1, x_2, \dots, x_\ell); \quad d(\prod x_i) = n\}$$

Power generating functions for polynomials

$$S(z) := \sum_{\underline{x} \in \Omega} z^{d(\prod x_i)} \quad \implies \quad |\Omega_n| = [z^n]S(z)$$

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– for integers: the total length $d(\underline{x}) = \lfloor \log \prod x_i \rfloor \sim \sum \lfloor \log x_i \rfloor$

$$\Omega_n := \{ \underline{x} = (x_1, x_2, \dots, x_\ell); \quad e^n \leq \prod x_i < e^{n+1} \}$$

Dirichlet generating functions for integers

$$S(s) := \sum_{\underline{x} \in \Omega} (\prod x_i)^{-s} \quad \implies \quad |\Omega_n| = \sum_{e^n \leq N < e^{n+1}} [N^{-s}]S(s)$$

Analysis of algorithms and generating functions (II).

$$\mathbb{P}_n[L = m] = \frac{|\Omega_{n,m}|}{|\Omega_n|} \quad \text{with} \quad \Omega_{n,m} = \{\underline{x} \in \Omega_n; L(\underline{x}) = m\}$$

Numerators are expressed with coefficients of bivariate generating functions.

For polynomials:

$$\Omega_{n,m} := \{\underline{x} = (x_1, x_2, \dots, x_\ell); \quad d(\prod x_i) = n, \quad L(\underline{x}) = m\}$$

$$S(z) := \sum_{\underline{x} \in \Omega} z^{d(\prod x_i)}, \quad S(z, u) := \sum_{\underline{x} \in \Omega} z^{d(\prod x_i)} u^{L(\underline{x})}$$

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Analysis of algorithms and generating functions (GF): main principles.

Distributional analysis is related

- to the asymptotic behaviour of coefficients of the BGF's
- obtained with two steps:

Combinatorial step.

Translate the **structure** of the algorithm on the generating functions:

- obtain an **algorithmic expression** of the generating function,
- from which the dominant singularity becomes apparent.

Analytic step.

For any GF, the **asymptotic** behaviour of **coefficients** is related to

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The same principles in the two analyses and two main differences.

- Dirichlet GF's more difficult to study than power GF's
- For integers, there are correlations due to carries.

II. Analysis of the polynomial case.

Based on Analytic Combinatorics
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Probabilistic analysis of the plain ℓ -GCD algorithm on $\mathbb{F}_q[X]$.

On the input $(x_1, x_2, \dots, x_\ell)$,

- the algorithm computes the total gcd $y_\ell := \gcd(x_1, x_2, \dots, x_\ell)$
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The set of inputs is $\Omega = \{(x_1, \dots, x_\ell); \quad x_i \text{ monic} \in \mathbb{F}_q[X]\}$

The size of an input : $d(x_1, \dots, x_\ell) = d(x_1) + \dots + d(x_\ell)$,

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Main costs of interest

– the number L_k of divisions during the k -th phase

i.e. on the input (x_k, y_{k-1})

– the degree D_k of the k -th gcd

(at the beginning of the k -th phase).

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$$\dots = \dots +$$

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– the sequence of quotients (m_1, m_2, \dots, m_r) ,

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- the sequence of quotients (m_1, m_2, \dots, m_r) , where
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The combinatorial bijection induced by the Euclid Algorithm [$\ell = 2$]

$\text{Euclid}(a_1, a_2),$	[case $d(a_1) \geq d(a_2)$].
$a_1 = m_1 a_2 + a_3$	$0 < d(a_3) < d(a_2)$
$a_2 = m_2 a_3 + a_4$	$0 < d(a_4) < d(a_3)$
$\dots = \dots +$	
$a_{r-1} = m_{r-1} a_r + a_{r+1}$	$0 < d(a_{r+1}) < d(a_r)$
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The last non zero remainder is the gcd y . Here $y = a_{r+1}$.	

The Euclid Algorithm is then extended to the case when $d(a_1) < d(a_2)$,
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Proof of the decomposition:

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Algorithmic expression of the GF of the ℓ -Euclid Algorithm

We have shown that the Euclid algorithm ($\ell = 2$) translates as a product

$$U(z_1)U(z_2) = T(z_1, z_2)U(z_1z_2), \quad \text{with} \quad T(z_1, z_2) = \frac{U(z_1) + U(z_2) - 1}{1 - G(z_1z_2)}$$

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Now, with $z = z_1 = \dots = z_\ell$,

the (plain) generating function $S(z)$ of \mathcal{U}^ℓ has the alternative expression

$$S(z) = U(z)^\ell = U(z^\ell) \prod_{k=1}^{\ell-1} T(z, z^k)$$

which is an exact translation of the ℓ -Euclid algorithm.

T is the “phase generating function”.

Bivariate Generating Functions relative to the ℓ -Euclid Algorithm

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- L_k (number of steps in the k -th phase)
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$$L_k(z, u) = U(z)^\ell \cdot \frac{T(z, z^k, u)}{T(z, z^k)}, \quad D_k(z, u) = U(z)^\ell \cdot \frac{U(z^k, u)}{U(z^k)},$$

With:

$$T(z, t, u) = u \frac{U(z) + U(t) - 1}{1 - uG(zt)}, \quad U(t, u) = \frac{1}{1 - gut}.$$

Towards the distributional analysis of L_k and D_k .

$$\mathbb{P}_n[L_k > m] = \sum_{j>m} \frac{[u^j z^n] L_k(z, u)}{[z^n] S(z)} = \frac{[z^n] \sum_{j>m} [u^j] L_k(z, u)}{[z^n] S(z)} = \frac{[z^n] \widehat{L}_k^{[m]}(z)}{[z^n] S(z)}$$

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The generating functions “of the numerators” are

$$\widehat{L}_k^{[m]}(z) = \frac{1}{(1 - qz)^\ell} \cdot G(z^{k+1})^m, \quad \widehat{D}_k^{[m]}(z) = \frac{1}{(1 - qz)^\ell} \cdot (qz^k)^m,$$

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The asymptotics depends on the value $a := A_k(1/q)$ at the pole $z = 1/q$

- For the first phase $k = 1$, one has $a = 1$
- For the subsequent phases $k \geq 2$, one has $a < 1$

A general result for the asymptotics of coefficients

Consider the function $F^{[m]}(z) = \frac{1}{(1-z)^\ell} A(z)^m$, with $\ell \geq 2$

where (i) $A(z)$ is analytic on the disk $|z| \leq \rho$ with $\rho > 1$,

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(iii) for $|z|$ close enough to 1, $|A(z)| \leq A(|z|)$.

Then, for $n \rightarrow \infty$ and $m/n \in [0, a/b[$,

$$[z^n]F^{[m]}(z) = \frac{n^{\ell-1}}{(\ell-1)!} a^m \left(1 - \frac{b m}{a n}\right)^{\ell-1} \left[1 + O\left(\frac{1}{n}\right)\right].$$

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Application to the present situation.

For the first phase: $a = 1 \implies$ A "beta" behavior $(1, \ell - 1)$

For the subsequent phases: $a < 1 \implies$ A geometric behavior of ratio a

Main result for the number of divisions L_k – First phase ($k = 1$)

The number of divisions L_1 performed during the first phase

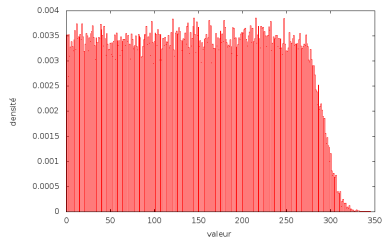
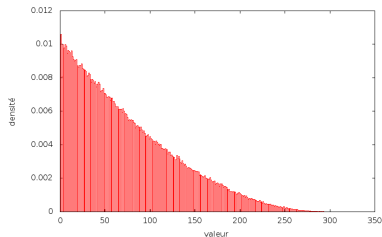
– has a **mean value** of **linear** order

$$\mathbb{E}_n[L_1] = \frac{q-1}{2q} \frac{n}{\ell} + \frac{3q+1}{4q} + O\left(\frac{1}{n}\right). \quad \frac{2q}{q-1} = \text{entropy}$$

– follows an asymptotic **beta law** of parameter $(1, \ell - 1)$.

Its distribution satisfies when $n \rightarrow \infty$, and $m/n \in [0, (q-1)/(2q)]$

$$\mathbb{P}_n[L_1 > m] = \left(1 - \frac{2q}{q-1} \frac{m}{n}\right)^{\ell-1} + O\left(\frac{1}{n^\alpha}\right).$$



Main result for the number of divisions L_k – Subsequent phases (case $k \geq 2$)

For $k \geq 2$, the number of divisions performed during the k -th phase
– has a **mean value** of **constant** order

$$\mathbb{E}_n[L_k] = \left(1 + \frac{q-1}{q^k - q}\right) + O\left(\frac{1}{n}\right),$$

– follows an asymptotic **geometric law**, with ratio $\frac{q-1}{q^k - 1}$

For $n \rightarrow \infty$ and $m/n \in [0, 1/(k+1) \cdot (q^k - 1)/q^k]$

$$\mathbb{P}_n[L_k \geq m] = \left(\frac{q-1}{q^k - 1}\right)^m + O\left(\frac{\log n}{n}\right)$$

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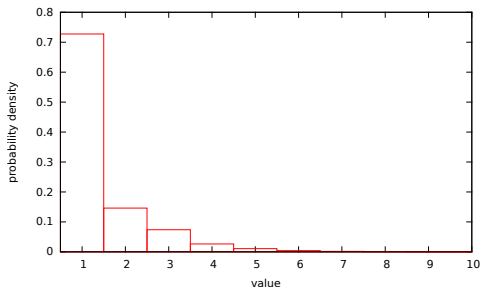
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III. Analysis on integers.

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We use Dirichlet generating functions
together with a dynamical point of view

Remind : The Euclid algorithm ($\ell = 2$) on polynomials
translates as a product of power generating functions

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We will prove that the Euclid algorithm ($\ell = 2$) on integers
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$$\zeta(s)\zeta(t) = \zeta(s+t) \cdot \left(\frac{1}{2}(I - \mathbf{G}_{s+t})^{-1} \circ (\mathbf{G}_s + \mathbf{G}_t) \right) [1](0)$$

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This involves – the Riemann Dirichlet series $\zeta(s) = \sum_{a \geq 1} a^{-s}$
– a functional operator \mathbf{G}_s (that depends on a parameter s)

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The operator \mathbf{G}_s “generates” the quotients; it is closely related
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This involves – the Riemann Dirichlet series $\zeta(s) = \sum_{a \geq 1} a^{-s}$
– a functional operator \mathbf{G}_s (that depends on a parameter s)

$$\mathbf{G}_s[f](t) = \sum_{m \geq 1} \left(\frac{1}{m+t} \right)^s f\left(\frac{1}{m+t} \right)$$

The operator \mathbf{G}_s “generates” the quotients; it is closely related
to the transfer operator of the underlying dynamical system...

An instance of a “dynamical” analysis....

More involved than the previous one, but provides the same type of results.

Similarities and differences between the two analyses

	Polynomials	Integers
GF	Power GF's	Dirichlet GF and operators
Basic tool	$G(z) = \sum_m z^{d(m)}$	$\mathbf{G}_s[f](t) = \sum_{m \geq 1} \left(\frac{1}{m+t}\right)^s f\left(\frac{1}{m+t}\right)$
Phase GF	$\frac{U(z^k) + U(z) - 1}{1 - G(z^{k+1})}$	$(I - \mathbf{G}_{(k+1)s})^{-1} \circ (\mathbf{G}_{ks} + \mathbf{G}_s)[1](0)$
Singularities	z s.t. $G(z^{k+1}) = 1$	s s.t. $\lambda((k+1)s) = 1$
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$\lambda(s)$ is the **dominant eigenvalue** of \mathbf{G}_s

$\lambda(2) = 1$; $\lambda'(2)$ closely related to the entropy

Results in the integer case.

NB: the integer size of the input \approx number of digits in base e

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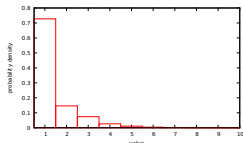
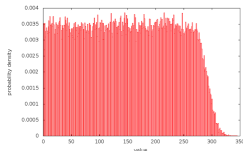
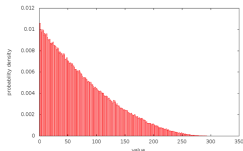
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- during **subsequent** phase:
 - it is **constant** on average
 - it asymptotically follows a **geometric** law

The **same phenomena** occur for the **size** of the **partial gcd**.



Probabilistic analysis of the plain ℓ -GCD algorithm on \mathbb{Z} .

On the input $(x_1, x_2, \dots, x_\ell)$,

- the algorithm computes the total gcd $y_\ell := \gcd(x_1, x_2, \dots, x_\ell)$
- with $\ell - 1$ phases.
- The k -th phase computes the k -th gcd,

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The set of inputs is $\Omega = \{\underline{x} := (x_1, \dots, x_\ell); \quad x_i \in \mathbb{N}\}$

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Main costs of interest

– the number L_k of divisions during the k -th phase

i.e. on the input (x_k, y_{k-1})

– the size D_k of the k -th gcd

(at the beginning of the k -th phase).

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For LFT's, the denominators are recovered with the derivatives

$$\text{if } h(x) = \frac{ax + b}{cx + d}, \quad \text{then } h'(x) = \frac{\det h}{(cx + d)^2}$$

Proof of the alternative expression for the Dirichlet generating function ($\ell = 2$)

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The Euclid algorithm ($\ell = 2$) on integers

translates as a product of Dirichlet generating functions

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Then, for any $\ell \geq 2$, the ℓ -Euclid algorithm translates as the product

$$\zeta(s_1) \cdot \dots \cdot \zeta(s_\ell) = \zeta(t_\ell) \prod_{k=1}^{\ell-1} T(s_{k+1}, t_k) \quad [t_k := s_1 + s_2 + \dots + s_k,]$$

The Euclid algorithm ($\ell = 2$) on integers

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$$\zeta(s_1) \zeta(s_2) = T(s_1, s_2) \zeta(s_1 + s_2),$$

$$\text{with } T(s_1, s_2) = \frac{1}{2} (1 - \mathbf{G}_{s_1+s_2})^{-1} \circ (\mathbf{G}_{s_1} + \mathbf{G}_{s_2}) [1](0)$$

Then, for any $\ell \geq 2$, the ℓ -Euclid algorithm translates as the product

$$\zeta(s_1) \cdots \zeta(s_\ell) = \zeta(\ell s) \prod_{k=1}^{\ell-1} T(s_{k+1}, t_k) \quad [t_k := s_1 + s_2 + \dots + s_k,]$$

Now, with $s = s_1 = \dots = s_\ell$,

the (plain) generating function $S(s)$ of \mathbb{N}^ℓ has an alternative expression

$$S(s) = \zeta(s)^\ell = \zeta(\ell s) \prod_{k=1}^{\ell-1} T(s, ks)$$

which is an exact translation of the ℓ -Euclid algorithm.

T is the “phase generating function”.

Bivariate Generating functions for the ℓ -Euclid Algorithm

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$$L_k(s, u) = \zeta(s)^\ell \cdot \frac{T(s, ks, u)}{T(s, ks)}, \quad D_k(s, u) = \zeta(s)^\ell \cdot \frac{Z(ks, u)}{\zeta(ks)},$$

$$T(s, t, u) = \frac{1}{2} u (1 - u \mathbf{G}_{s+t})^{-1} \circ (\mathbf{G}_s + \mathbf{G}_t)[1](0), \quad Z(s, u) = \sum_{n \geq 1} \frac{u^{d(n)}}{n^s}$$

Towards the distributional analysis of L_k and D_k .

The generating functions of the events $[L_k > m]$ and $[D_k > m]$

$$\widehat{L}_k^{[m]}(s) := \sum_{j>m} [u^j] L_k(s, u), \quad \widehat{D}_k^{[m]}(s) = \sum_{j>m} [u^j] D_k(s, u)$$

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The asymptotics depends on the value $a := A_k(1)$ at the pole $s = 1$ of $B(s)$

For $k = 1$, one has $a = 1$ – For $k \geq 2$, one has $a < 1$.

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The number of divisions L_1 performed during the first phase

– has a mean value of linear order

$$\mathbb{E}_n[L_1] = \frac{6 \log 2}{\pi^2} \frac{n}{\ell} \left[1 + O\left(\frac{1}{n}\right) \right] \quad \frac{\pi^2}{6 \log 2} = \text{entropy}$$

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Its distribution satisfies when $n \rightarrow \infty$, and $m/n \in [0, (6 \log 2)/\pi^2]$

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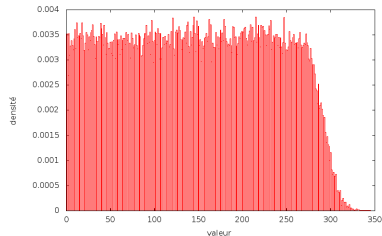
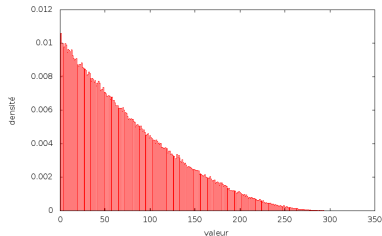
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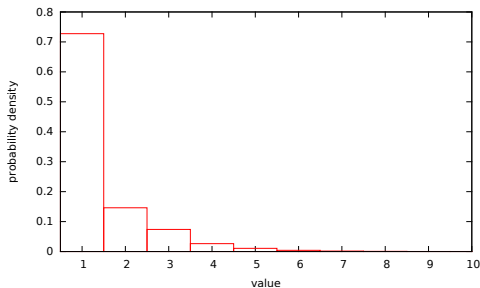
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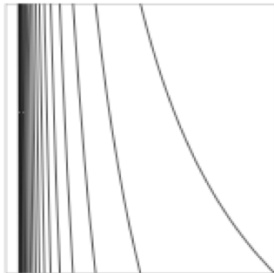
IV. An unified point of view for the two analyses?

A dynamical point of view....

The underlying dynamical system in the integer case

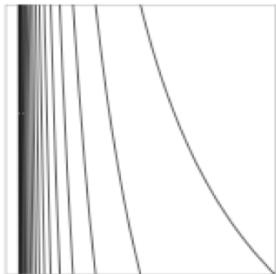
The underlying dynamical system in the integer case defined by the pair $([0, 1], V)$.

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



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The set of the inverse branches of V is

$$\left\{ h_m(x) = \frac{1}{m+x} \right\}$$

It builds the continued fraction expansions.

The operator \mathbf{G}_s is the transfer operator of the dynamical system.

It involves the derivatives of the inverse branches of V

$$\mathbf{G}_s[f](x) = \sum_{m \geq 1} \left(\frac{1}{m+x} \right)^s f \left(\frac{1}{m+x} \right)$$

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The exact ultrametric analog of the DS in the integer case.

$$\mathbb{F}_q[Z] \longleftrightarrow \mathbb{Z}, \quad \mathbb{F}_q(Z) \longleftrightarrow \mathbb{Q}, \quad \mathbb{F}_q((1/Z)) \longleftrightarrow \mathbb{R}$$

$\mathbb{F}_q((1/Z))$: the completion of $\mathbb{F}_q[Z]$ for the ultrametric norm $\|u\| := q^{-d(u)}$.

The integer part function $[\cdot]$ satisfies $[f] = 0$ iff $\|f\| \leq 1$

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Classical analysis for polynomials: a particular case of a dynamical analysis

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Methodologic point of view.

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 - Classical analysis versus dynamical analysis
 - Provides precise distributional results.
 - With an (unexpected) occurrence of the Beta law

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