

# Gaussian behavior of quadratics irrationals

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Alea EnAmSud and Dyna3s, June 2017

Joint work with Brigitte Vallée

## Reduced quadratic irrationals: basic facts and notations

Every real number  $x \in [0, 1[$  can be written as a **continued fraction (CFE)**

$$x = \frac{1}{\underset{\textcolor{red}{m}_1}{\textcolor{red}{m}_1} + \frac{1}{\ddots + \frac{1}{\underset{\textcolor{red}{m}_k}{\textcolor{red}{m}_k} + \frac{1}{\ddots}}}} = [m_1, \dots, m_k, \dots]$$

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We are interested in combinatorial properties of the set

$$\mathcal{P} = \{x \in \mathcal{I} \mid x \text{ is a rqi number} \}.$$

# Additive costs

## Definition

A **digit-cost** is a nonnegative and nonzero map on the natural numbers:

$$c : \mathbb{Z}_{>0} \mapsto \mathbb{R}_{\geq 0}$$

A digit-cost defines a **total additive cost**  $C$  on  $\mathcal{P}$ .

For a rqi  $x = \overline{m_1, m_2, \dots, m_p}$ ,

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- ▶  $\ell(m) = \lfloor \log_2 m \rfloor + 1$ ;  
total cost  $C(x)$ : number of bits to store the CFE of  $x$ .

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## Remark

$\epsilon$  is not additive and it is not multiplicative.

# Our goals

The set  $\mathcal{P}$  of rqi's is endowed with

- ▶ An additive cost  $C$
- ▶ A notion of size  $\epsilon$ .

The set of rqi with **size at most  $N$** :

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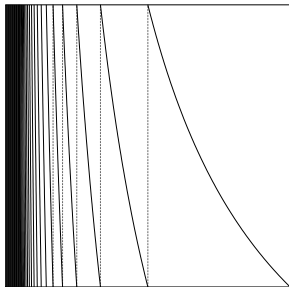
$$\mathbb{P}_N \left[ x \mid \frac{C(x) - \mu(c) \log N}{\sqrt{\nu(c) \log N}} \leq t \right] = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^t e^{-u^2/2} du + O \left( \frac{1}{\sqrt{\log N}} \right) .$$

The constants  $\mu(c)$  and  $\nu(c)$  are computable.

## Our results: computation of constants $\mu(c)$ and $\nu(c)$

The Gauss map:

$$T : [0, 1] \mapsto [0, 1] \quad T(x) = \left\{ \frac{1}{x} \right\} \text{ if } x \neq 0 \quad \text{and} \quad T(0) = 0$$



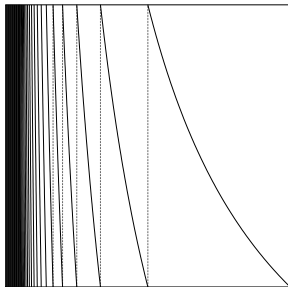
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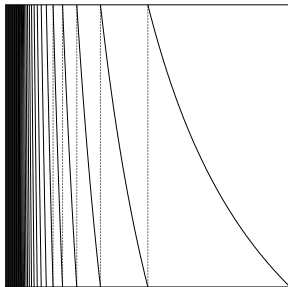
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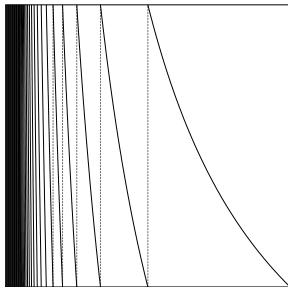
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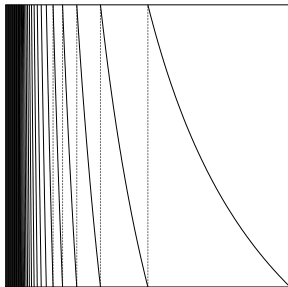
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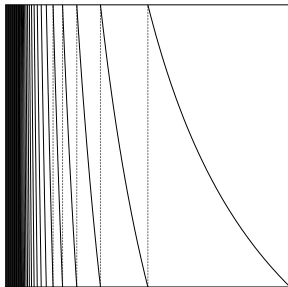
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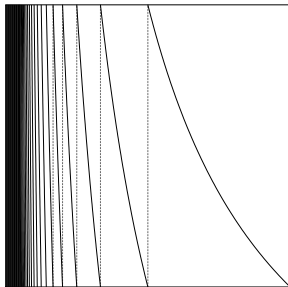
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Remark:  $\mathbf{H} = \mathbf{H}_{1,0}$



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- ▶  $\lambda'_s(1, 0)$  equals the opposite of the **entropy of the Gauss map**  $\mathcal{E}$ :

$$\lambda'_s(1, 0) = -\frac{\pi^2}{6 \log 2}.$$

It can be computed as  $\lambda'_s(1, 0) = -\int_0^1 \log |T'(x)| \psi(x) dx$ .

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- ▶  $\lambda'_s(1, 0)$  equals the opposite of the **entropy of the Gauss map**  $\mathcal{E}$ :

$$\lambda'_s(1, 0) = -\frac{\pi^2}{6 \log 2}.$$

It can be computed as  $\lambda'_s(1, 0) = -\int_0^1 \log |T'(x)| \psi(x) dx$ .

- ▶  $\lambda'_w(1, 0)$  equals the **weighted average**  $\mathbb{E}[c]$  of the digit-cost  $c$  with respect to the Gauss density, i.e.

$$\lambda'_w(1, 0) = \sum_{m=1}^{\infty} c(m) \int_{1/(m+1)}^{1/m} \psi(x) dx.$$

## Constants $\mu(c)$ and $\nu(c)$

### Theorem

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the following holds:

$$\mathbb{E}_N[C] = \mu(c) \log N + O(1), \quad \mathbb{V}_N[C] = \nu(c) \log N + O(1) .$$

with  $\mu(c)$  and  $\nu(c)$  positive.

### Theorem

Let  $\mathcal{E}$  be the entropy of the Gauss map and let  $\mathbb{E}[c]$  be the weighted average:

$$\mu(c) = \frac{2\lambda'_w(1,0)}{|\lambda'_s(1,0)|} = \frac{2}{\mathcal{E}} \mathbb{E}[c] \quad \text{and}$$

$$\mathbb{E}[c] = 1, \quad \mathbb{E}[\chi_n] = \frac{1}{\log 2} \log \left[ \frac{(n+1)^2}{n(n+2)} \right], \quad \mathbb{E}[\ell] = \frac{1}{\log 2} \prod_{i=1}^{\infty} \log \left( 1 + \frac{1}{2^k} \right) .$$

## Previous results

Rqi framework: Pollicott, Faivre, Vallée

### Previous works: rationals trajectories

Let  $\Omega = \{(u, v) \in \mathbb{Z}^2 : 1 \leq u \leq v, \gcd(u, v) = 1\}$

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### Remarks

- ▶ The constants  $\mu(c)$  and  $\nu(c)$  in the rational case are the same as for the rqi case.
- ▶ The constants hidden in the  $O(1)$  are expressed in terms of the transfer operator in both cases.

They do not coincide !

- ▶ Our methods are inspired in those of Baladi and Vallée, 2005

# Real trajectories

A classical result:

## Theorem

*The mean of a digit-cost  $c$  satisfies:*

$$\frac{1}{N} \sum_{1 \leq k \leq N} c(m_k) \rightarrow \mathbb{E}[c] \quad N \rightarrow \infty$$

*with the exception of a set of zero measure.*

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## Remark

Rationals and rqi's are set of zero measure.

## A “combinatorial” version

$$\frac{1}{p(x)} \sum_{1 \leq k \leq p(x)} c(m_k(x)) \rightarrow \mathbb{E}[c] \quad N \rightarrow \infty.$$

Here  $p(x)$  is the length of the smallest period of  $x$ .

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$$\frac{\mathbb{E}_N[C]}{\mathbb{E}_N[p]} = \sum_{x \in \mathcal{P}_N} \frac{1}{\sum_{x \in \mathcal{P}_N} p(x)} \sum_{1 \leq k \leq p(x)} c(m_k(x)) \rightarrow \mathbb{E}[c] \quad N \rightarrow \infty.$$

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## A comment about “moderate growth”

### Remark

Our results hold under the assumption that the digit-cost  $c$  is of moderate growth:  $c(m) = O(\log m)$ .

### Explanation

The weighted transfer operator depends of two complex parameters  $(s, t)$ :

$$f \mapsto \mathbf{H}_{s,t}[f](x) = \sum_{m \geq 1} \frac{e^{t c(m)}}{(m+x)^{2s}} f\left(\frac{1}{m+x}\right)$$

For  $c(m) = \Theta(\log^a(m))$  with  $a > 1$ ,  $\mathbf{H}_{1,t}$  is not convergent for  $\Re t > 0$ .

Spectral properties of the transfer operator are well-known around  $(1, 0)$ .

Our methods rely on the fact that  $\mathbf{H}_{1,t}$  is convergent for  $t$  in a complex neighborhood of  $t = 0$ .

# Methods: The role of Dirichlet series

## “Combinatorial class”

The set  $\mathcal{P}$  of rqi's is endowed with

- ▶ an additive cost  $C$ ,
- ▶ a notion of size  $\epsilon$ .

Our main object of analysis: the bivariate Dirichlet series  $P(s, t)$

$$P(\textcolor{red}{s}, \textcolor{blue}{t}) = \sum_{x \in \mathcal{P}} e^{tC(x)} \epsilon(x)^{-s}$$

where  $s$  and  $t$  are complex parameters.

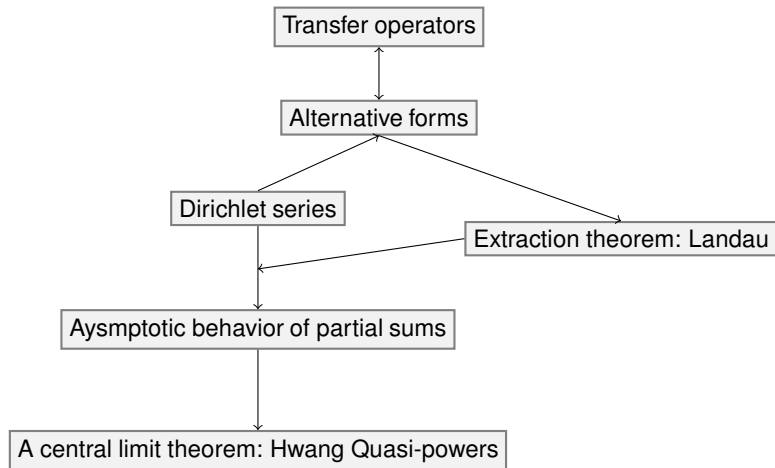
## Extraction theorems

Extraction theorems gives information about the partial sums of coefficients

$$S_N^{[c]}(t) = \sum_{\epsilon(x) \leq N} e^{tC(x)}$$

if we have information about the analytic behavior of  $s \mapsto P(s, t)$ .

## Interaction between the main elements of the proof



## Alternative forms for $P(s, t)$ : replace rqi's by homographies

### Homographies

Recall the set of inverse branches of the Gauss map  $T$ :

$$\mathcal{H} := \{h_m; \quad h : x \mapsto \frac{1}{m+x}, \quad m \geq 1\}$$

Composition of inverse branches give rise to the two sets:

$$\mathcal{H}^k = \{h = h_{m_1} \circ h_{m_2} \circ \cdots \circ h_{m_k}\} \quad \mathcal{H}^+ = \bigcup_{k \geq 1} \mathcal{H}^k$$

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- The number  $x$  is a rqi iff there exists  $h \in \mathcal{H}^+$  so that  $x$  is its fixed point:  
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### Important relation

Each rqi  $x$  is the fixed point of only one primitive  $h \in \mathcal{H}^+$ .

## Size: replace rqj's by homographies

Recall:  $x \longrightarrow \mathbb{Q}(\sqrt{\Delta}) \longrightarrow \epsilon(x) =$  fundamental unit of  $\mathbb{Q}(\sqrt{\Delta})$

### Definition

Let  $x$  be a rqj number and define

$$\alpha(x) = \prod_{i=0}^{p(x)-1} T^i(x).$$

Here  $T$  is the Gauss map and  $p(x)$  is the smallest period length of  $x$ .

### Theorem

- *The size  $\epsilon(x)$  and  $\alpha(x)$  are related by*

$$\epsilon(x) = \alpha(x)^{-r(x)}, \quad r(x) = 1 \text{ for even } p(x), \quad r(x) = 2 \text{ for odd } p(x).$$

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- ▶ *If  $x$  is the fixed point of the primitive homography  $h$ , then*

$$\alpha(x) = |h'(x)|^{1/2}$$

## Ready to extend cost and size to $\mathcal{H}^k$ . A new Dirichlet series

### Definition

For  $h \in \mathcal{H}^k$  (primitive or not) with a fixed point  $x_h$ , let

$$\alpha(h) := |h'(x_h)|^{1/2}.$$

The size of the homography is defined by

$$\epsilon(h) = \alpha(h)^{-r(h)}, \quad r(h) = 1 \text{ for even } k, \quad r(h) = 2 \text{ for odd } k.$$

### Definition

The total cost  $C$  is naturally extended to  $\mathcal{H}^+$ .

If  $h = h_{m_1} \circ h_{m_2} \circ \dots \circ h_{m_k}$ ,  $C(h) = c(m_1) + c(m_2) + \dots + c(m_k)$ .

### Definition

A new Dirichlet series: Replace the set  $\mathcal{P}$  of rqi's by  $\mathcal{H}^+$  and  $\epsilon$  by  $\alpha$  in the definition of  $P(s, t)$ :

$$Y(s, t) = \sum_{h \in \mathcal{H}^+} e^{tC(h)} \alpha(h)^{2s}$$

with  $\alpha(h) = |h'(x_h)|^{1/2}$ .

## Alternative form for $Y(s, t)$ and transfer operators

Dirichlet series

$$Y(s, t) = \sum_{h \in \mathcal{H}^+} e^{tC(h)} |h'(x_h)|^s$$

Transfer operator

$$\mathbf{H}_{s,t}(I - \mathbf{H}_{s,t})^{-1}[f](x) = \sum_{h \in \mathcal{H}^+} e^{tC(h)} |h'(x)|^s f(h(x))$$

Relation

$$Y(s, t) \approx \mathbf{H}_{s,t}(I - \mathbf{H}_{s,t})^{-1}$$

Remark the difference between the evaluation points

# Relating fixed points and traces of transfer operators

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  - ▶ If  $f'(x_h) = 0$  ...
- 
- ▶ The trace is the sum of all the eigenvalues:  $h'(x_h)^n$

$$\text{Tr}(f \mapsto |h'(x_h)| f \circ h) = \frac{\alpha(h)^2}{1 - (-1)^{|h|} \alpha(h)^2} \quad \text{with } \alpha(h) = |h'(x_h)|^{1/2}.$$

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- ▶ The trace is the sum of all the eigenvalues:  $h'(x_h)^n$

$$\text{Tr}(f \mapsto |h'(x_h)| f \circ h) = \frac{\alpha(h)^2}{1 - (-1)^{|h|} \alpha(h)^2} \quad \text{with } \alpha(h) = |h'(x_h)|^{1/2}.$$

# Relating fixed points and traces of transfer operators

## Theorem

$$Y(s, t) = \text{Trace}(\mathbf{H}_{s,t}(I - \mathbf{H}_{s,t})^{-1})$$

*Sketch of the proof:* Consider the composition operator  $f \mapsto f \circ h$ .

- ▶  $\lambda$  is an eigenvalue iff  $f \circ h(x) = \lambda f(x)$  for any  $x$ .
- ▶ If  $x = x_h$ , the previous equality becomes  $f(x_h) = \lambda f(x_h)$ .

Two cases:  $f(x_h) \neq 0$  or  $f(x_h) = 0$ .

- ▶ If  $f(x_h) \neq 0$ , then  $\lambda = 1$ .
- ▶ If  $f(x_h) = 0$ , differentiation of the equality

$$f \circ h(x) = \lambda f(x)$$

yields

$$f'(x_h)h'(x_h) = \lambda f'(x_h).$$

Again two cases according to  $f'(x_h)$  is zero or not.

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The statement is obtained adding over all  $h \in \mathcal{H}^+$ .

## Analytic properties of $Y(s, t)$

For  $s$  near the real axis, the transfer decomposes

$$\mathbf{H}_{s,t} = \lambda(s, t)\mathbf{P}_{s,t} + \mathbf{N}_{s,t}$$

where

- ▶  $\lambda(s, t)$  is the dominant eigenvalue,
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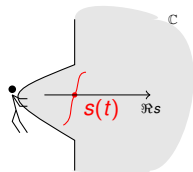
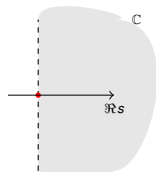
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## Theorem

$s \mapsto Y(s, t)$  extends meromorphically to a complex neighborhood of  $s = 1$  (uniformly for  $t \sim 0$ ):

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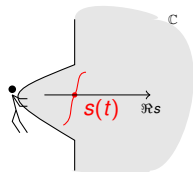
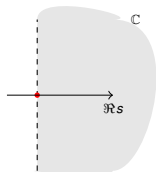
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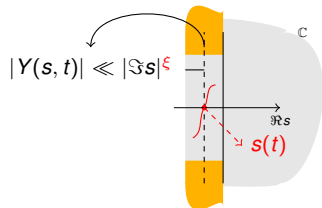
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*The solutions  $s(t)$  of  $1 - \lambda(s, t) = 0$  are the only singularities of  $s \mapsto Y(s, t)$ .*

# Extensions of Dolgopyat-Baladi-Vallée results

- Dolgopyat, Baladi and Vallée for the “true quasi-inverse  $(I - \mathbf{H}_{s,t})^{-1}$ ”

The following holds for  $s \mapsto Y(s, t)$



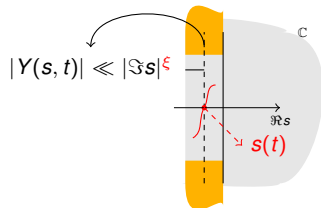
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The original series  $P(s, t)$  and the series  $Y(s, t)$  has the same analytic behavior.

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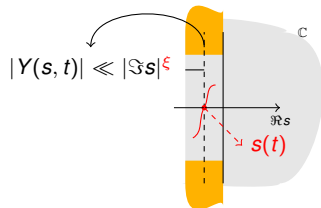
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- ▶ This work: periodic points of the Gauss map (infinite number of branches).

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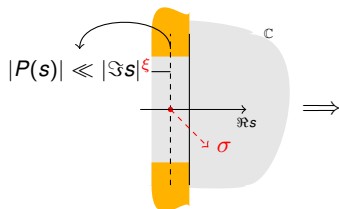
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# Dirichlet series and Landau Theorem

Dirichlet series:  $P(s) = \sum_{x \in \mathcal{P}} \epsilon(x)^{-s}$

If



$$S_N = |\mathcal{P}_N| = \sum_{\epsilon(x) \leq N} 1$$

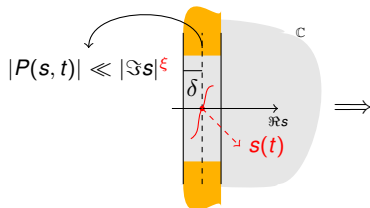
$$S_N = KN^\sigma \left(1 + O(N^{-\beta})\right)$$

$$\beta = \beta(\xi, \delta) > 0$$

Here  $\delta$  is the distance between  $s(t)$  and the left straight line.

Bivariate Dirichlet series:  $P(s, t) = \sum_{x \in \mathcal{P}} e^{tC(x)} \epsilon(x)^{-s}$

If



$$S_N^{[c]}(t) = \sum_{\epsilon(x) \leq N} e^{tC(x)}$$

$$S_N^{[c]}(t) = v(t)N^{s(t)} \left(1 + O(N^{-\beta})\right)$$

$$\beta = \beta(\xi, \delta) > 0$$

## A central limit theorem. The final element of the proof.

### Definition

The moment generating function  $\mathbb{E}_N[e^{tC}]$  satisfies

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Constants are computed from  $s(t)$  using the relation  $\lambda(s, t) = 1$ .

# Extensions and conclusions

- ▶  $x \mapsto \log \epsilon(x)$  is close to the Lévy constant. It is **not** an additive cost.
  - ▶  $\mathbb{E}_N[\log \epsilon] = 2 \log N + O(1)$ .
  - ▶ The variance is of constant order.
- ▶ The set of  $\mathcal{P}[M] = \{\text{rqi with partial quotients bounded by } M\}$ 
  - ▶ Similar Gaussian laws for additive parameters and  $M$  large enough.
  - ▶ Based on properties of the constrained transfer operator

$$\mathbf{H}_{M,s}[f](x) = \sum_{m \leq M} \frac{1}{(m+x)^{2s}} f\left(\frac{1}{m+x}\right).$$

- ▶ For small  $M$ , Dolgopyat like results are not proved to hold.